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DEVELOPMENT OF HYDRAULIC RELATIONSHIPS FOR ESTIMATING IN-BANK RIVER DISCHARGE USING REMOTELY SENSED DATA

ΒY

David Michael Bjerklie BS, University of Maine, 1977 MS, University of New Hampshire, 1980 MSCE, University of Alaska, 1987

DISSERTATION

Submitted to the University of New Hampshire

in Partial Fulfillment of

the Requirements for the Degree of

Doctor of Philosophy

In

Earth Sciences

May, 2004

UMI Number: 3132783

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This dissertation has been examined and approved.

S. Lawrence Dingman, Professor of Hydrology and Water Resources

Charles J. Worosmarty, Director Water Systems Analysis Group, Institute for the Study of Earth, Oceans and

Space

Carl H. Bolster, Assistant Professor of Hydrology

Russell G. Congalton, Professor of Remote Sensing and GIS

G. Robert Brakenridge, Research Associate and Professor of Geography, Dartmouth College

22 January 2014 Date

ACKNOWLEDGEMENTS

First and foremost I would like to acknowledge my wife and family who have encouraged and patiently supported me through this process. My advisor, S. Lawrence Dingman, Department of Earth Science, provided both the foundation and the continued encouragment that I needed to compelete this research. I would also like to acknowledge the valuable input, guidance and review comments from my graduate committee: Charles J. Vorosmarty, Complex Systems Research Center, University of New Hampshire; Carl H. Bolster and Russell G. Congalton Department of Natural Resources, University of New Hampshire; and G. Robert Brakenridge, Department of Geography, Dartmouth College. I also received valuable comments from William Kirby of the United States Geological Survey. In addition, I am indebted to Balazs Fekete, Research Scientist at the Institute for the Study of Earth, Oceans and Space at the University of New Hampshire for providing insight, ideas and data which contributed to much of this work; Ellen Douglas and all of the staff and colleagues in the Water Systems Analysis Group at the Complex Systems Research Center, University of New Hampshire for providing input, comments and support; and Delwyn Moller at JPL-NASA and Laurence Smith at UCLA for valuable input and comments to this work. Elements of this research were funded by NASA grant numbers NAG5 – 7601 and NAG5 – 8683. My interaction with NASA researchers as a partcipant in the NASA grant program has been both rewarding and professionally encouraging.

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ABSTRACT

DEVELOPMENT OF HYDRAULIC RELATIONSHIPS FOR ESTIMATING IN-BANK RIVER DISCHARGE USING REMOTELY SENSED DATA

by

David M. Bjerklie

University of New Hampshire, May, 2004

An evaluation river hydraulic data currently or potentially available from satellite and other remote platforms was completed, and a set of discharge estimation models proposed that can use the remotely sensed information to estimate discharge with reasonable accuracy. Reasonable accuracy is defined as within +/- 20% of the observed on average for a large number of estimates. The proposed estimation models are based on the Manning and Chezy flow resistance equations, and utilize combinations of potentially observable variables including watersurface width, maximum-channel (or bankfull) width, mean water depth, mean maximumchannel depth, mean water velocity, and channel slope. Both stastistically and rationally derived prediction models are presented, developed and calibrated on a data base of river discharge measurements and a quasi-theoretical data base of synthetic data. It was found that the channel slope can be used in lieu of a measured water surface slope with very little reduction in prediction accuracy when considering many estimates. Notably absent from this list is a resistance variable, which is included in both the Manning and Chezy equations, because this variable cannot be observed or directly measured. One of the key outcomes of the research is that an exponent of 0.33 on the slope explains much of the variablity in the resistance variable, and provides better predictive qualities than the traditional value of 0.5. A dimensionally homogeneous form of the Manning equation was developed which derives the slope exponent of 0.33 based on stable-bed

grain size considerations. The prediction models were tested on two data sets of remotely sensed hydraulic information that included width, maximum channel width, and channel slope. Predictions were also made from a single radar image that also included remotely sensed surface velocity, demonstrating the potential for greatly improved accuracy with this additional information. Additionally, the prediction models were tested with channel slope information derived from a digital elevation model, and used to define river channel geometry for a continental scale runoff model.

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CHAPTER I

INTRODUCTION

Currently, less than 60% of the runoff from the continents is monitored at the point of inflow to the oceans (Fekete, 1999). The distribution of runoff within the continents is even less well monitored. Despite the importance of river discharge information, a comprehensive global river-monitoring network faces numerous technological, economic, and institutional obstacles. As a result, gaging stations and access to river-discharge information have been declining since the 1980s (Vorosmarty et al. 1999; IAHS, 2001). Hydrographic data obtained from satellites and other remote sources offer the possibility of broad and potentially frequent global coverage of river-discharge estimates (Barrett, 1998). Thus, a method that uses remotely sensed data to estimate river discharge would provide a means to maintain or even increase the global streamflow-monitoring network and may, in the long run, be a cost-effective method to obtain needed river-discharge data on a global scale.

Remotely sensed information can be appled to the science of estimating river discharge in two fundamental ways: 1) by providing data necessary to the watershed-runoff modeling process such as soil type, land cover, precipitation, topography, air temperature, solar radiation and wind speed such that runoff can be estimated and the discharge inferred from a routing scheme; or 2) by directly observing the hydraulic variables of flow in a river channel and estimating discharge from hydraulic functions that use this information. Although watershed modeling can provide estimates of river discharge, the discharge estimate is itself a by-product of a set of modeling assumptions and simplifications and cannot be said to be directly measured or estimated.

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Additionally, it would not be independent of spatial and temporal complexities that are subject to various scaling and model-input limitations. Without ground-based discharge calibration data for a specific watershed, discharge estimates made from the first approach may or may not be accurate. The spatial and temporal complexities of the watershed runoff process and the modeling of that process suggest that a general approach to estimating discharge in this way would be inherently unreliable without watershed-specific calibration. In general, the number of variables required to track the variability and describe the mechanics of discharge in a river system is much less than those necessary to understand and track the variability of the watershedrunoff process. It is because of these issues that estimating discharge with the second approach is preferred and is the focus of this study.

The goal of this dissertation is to develop, demonstrate, and evaluate the accuracy and application of methods suitable for estimating the discharge in rivers from remotely sensed river channel information. The specific objectives of the research are to:

- Document the type and quality of river channel information that can be potentially observed from remote platforms;
- Evaluate the potential application and accuracy of the observed data to estimate discharge using hydraulic relationships developed from ground-based river-discharge data;
- Develop suitable hydraulic relationships from general hydraulic principles;
- Develop and test a method derived from the hydraulic analysis to estimate discharge from currently available information;

• Evaluate the hydraulic methods within the context of general discharge modeling application that maximizes the use of remotely obtained or modeled river discharge and channel variables.

The use of remotely sensed information, including water-surface elevation, water-surface velocity and water-surface area, to track changes in river discharge has been shown to be feasible and potentially useful where ground-based data are difficult to obtain (Kuprianov, 1973; Koblinsky et al., 1993: Birkett, 1998; Brakenridge et al., 1994;and Brakenridge et al., 1998, Horritt et al., 2001, Jasinski et al., 2001). These studies suggest that remotely sensed river hydraulic data could be used to directly estimate the discharge at a specific location, if ground-based discharge measurements are used to develop discharge ratings in conjunction with the remotely observed variable(s). This strategy, however, does not capitalize on the advantage of using remotely sensed data because ground measurements of discharge are still a fundamental aspect of the approach.

If remotely sensed river hydraulic data were used to directly estimate the discharge without the need for ground-based calibration data, then remote observation platforms could be used to estimate discharge over large areas in many rivers. A bankside system that remotely obtains the cross-sectional area of flow and surface velocity of rivers has been demonstrated by Costa et al. (2000), however, this system would still require ground-based installation and maintenance. Thus, if satellite or aerial platforms could be used to obtain sufficient amount of information to estimate discharge, the need for ground-based measurements could be eliminated and would enable the potential of remote observation systems to obtain information over large geographic areas, including those areas that are difficult to access, to be realized.

Estimating discharge in rivers from hydraulic information obtained solely from aerial and satellite platforms has been explored and summarized by Smith et al. (1996 and 1997). The water-surface width (estimated from water-surface area), channel slope and mean channel width (estimated from channel surface area) can all be obtained from existing remote sources. The surface velocity of rivers can also be observed remotely using various forms of Doppler radar or lidar (Vörösmarty et al., 1999; Emmitt, 2000 personal communication; Moller, 2003 personal communication).

This study further explores the potential for, and the accuracy of, estimating discharge from remote observations of the river channel. Hydraulic relationships and a reasonably accurate methodology are developed for this purpose. The relationships are applied to a set of aerial photos and SAR images and hydraulic modeling and mapping applications also explored.

CHAPTER II

THE POTENTIAL FOR ESTIMATING RIVER DISCHARGE AND MEASURING HYDRAULIC VARIABLES REMOTELY

The measurement of river discharge from space will fundamentally require a knowledge of the hydraulic relationship between river characteristics that can be observed from space-based platforms and river discharge. This chapter reviews the types of river hydraulic information that can potentially be observed from space-based platforms and develops several general relationships that can use this information to estimate discharge. Hydraulic data from more than 1,000 flow measurements in a wide range of rivers are used to develop and validate the relationships. An analysis of the impact of measurement error on prediction accuarcy is also undertaken. The approaches reviewed here are based on fundamental in-stream hydraulic relationships that are independent of watershed or basin predictor variables. Thus, the prediction methods are independent of regional and temporal climatic and physiographic variability and can be considered to be generally applicable to fluvial environments.

Estimating River Discharge from Hydraulic Variables

For most rivers, discharge (Q) cannot be measured directly, but rather must be calculated from measurements of the pertinent hydraulic elements of the flow. Discharge at a river crosssection, from continuity, is the volumetric flow rate through that cross-section and is given by

$$Q = VWY = VA \tag{2-1}$$

where V is the average velocity, W is the water-surface width, Y the average water depth, and A the cross-sectional area perpendicular to the flow. Traditionally, Q is measured at selected cross-

sections in a river by measurement of the velocity, depth, and width at incremental vertical stations across the channel, and the incremental flow estimates are summed to obtain the discharge through the cross-section. These periodic velocity-area measurements of discharge are then correlated with measured water-surface elevation (stage) to develop a stage-discharge "rating" for the cross-section. The stage-discharge rating equation takes the general form (Rantz et al., 1982; Herschy, 1998)

$$Q = a(Z-e)^m$$
(2-2)

where Z is the stage and the coefficients a and m are characteristic of the specific channel crosssection, and e is the elevation of zero flow.

For the periods between measurements of Q, the stage (Z) is recorded and Q is inferred from the rating curve. Since the value of e represents the elevation of zero flow, the term (Z-e) may be viewed as equivalent to the effective flow depth (Y) and thus the rating provides an estimate of discharge from the hydraulic flow depth. A rating equation such as equation (2-2) is developed for a particular river channel or cross-section, and would not be expected to be applicable to any other river location (Rantz et al., 1982). This is because change in stage (or depth) is used as an index to change in width and velocity, and is specific to the channel characteristics of the reach being measured. Thus, single variate discharge ratings cannot be generalized without a substantial loss in accuracy. Inclusion of additional hydraulic infromation into the rating model would improve the accuracy of the rating by accounting for more of the variability at any specific location.

Recently, Jasinski et al. (2001) used river stage obtained from satellite (TOPEX/Poseiden) altimetry data to develop discharge ratings for several locations in the Amazon basin by comparing the altimetry data with stage and discharge measured at existing gaging stations. The accuracy of the ratings varied depending on distance between the altimetry

observation and the ground-measured discharge, and on the topography and the width of the river. This study demonstrated the feasibility of using satellite altimetry as a source of remote riverstage information. However, ground-based discharge data were required to develop the rating, and the derived ratings could not be extrapolated to other rivers or reaches of the Amazon. While such a system might have advantages in some situations, it does not solve the problems imposed by the costs of establishing and periodically measuring discharge on-the-ground, and would not offer the prospect of expanding the global coverage of discharge observations. Thus, a general rating that can estimate discharge from remotely obtained hydraulic data without ground-based measurements of discharge provides the best opportunity to capitaltize on satellite and other remote data sources.

A more general depth-discharge rating equation can be developed from the Manning equation which is widely viewed as generally applicable to natural rivers (Chow, 1959). Assuming a wide (W > 10Y) rectangular channel, the depth-discharge rating defined from the Manning equation is

$$Q = aY^{1.67}$$
 (2-3)

with

$$a = WS^{0.5}/n.$$
 (2-4)

where S is the friction slope (slope of the total energy grade line but equivalent to the water surface or bed slope assuming uniform flow conditions) and n is the Manning resistance coefficient. In equation (2-3), the average depth is the dynamic predictive variable and the coefficient a can be directly calculated from channel properties and is comprised of a geometric component defined by W and a channel component defined by $S^{0.5}/n$ (which represents the balance between the gravitational energy supplied to the reach, S and the flow resistance, n). In a rectangular channel, W is constant and thus if S and n are constant, the coefficient a is constant. To the extent that S and n vary with depth, the exponent of equation (2-3) may also vary.

If a parabolic shape is assumed for the channel cross-section, a common assumption for natural channels (Chow, 1959), the width is related to the depth by $W^x=aY$ where x is the parabolic order. The derived depth-discharge rating from this assumption is

$$Q = aY^{(1/x + 1.67)}$$
(2-5)

with

$$a = (W_m/Y_m^{1/x})(S^{0.5}/n)$$
(2-6)

The variable W_m is the maximum or bank-full width and Y_m the maximum or bank-full average depth.

A similar equation can also be developed that uses width as the rating variable:

$$O = aW^{(1.67x+1)}$$
(2-7)

with

$$a = (Y_m^{-1.67} / W_m^{-1.67x})(S^{0.5} / n)$$
(2-8)

Equations (2-5) and (2-7) can be regarded as generally applicable discharge ratings for withinbank flow to the extent that the Manning equation is generally applicable, under the assumption of a parabolic cross-section shape.

The channel resistance cannot be measured directly but is usually inferred from specific channel conditions including bed and bank material, channel irregularity (both in cross-section and planform shape) and other factors. In practice, the channel resistance is difficult to estimate with accuracy (Dingman and Sharma, 1997) and often varies considerably with discharge (Dingman, 1984). However, statistical studies by Riggs (1976), Jarrett (1984) and Dingman and Sharma (1997) have shown that reasonably accurate estimates of Q for within-bank flows can be obtained without resistance as an input variable, because the resistance varies with the channel geometry. Assuming that the hydraulic radius of the cross-section is equivalent to the mean depth

(which would be expected for a wide channel), Dingman and Sharma (1997) show using multiple regression analysis that for a wide range of rivers discharge can be estimated as:

$$O = 4.62 W^{1.17} Y^{1.57} S^{0.34}$$
(2-9)

with all variables in SI units. Equation (2-9) was calibrated with over 500 flow measurements in 128 rivers and provides estimate accuracies, on average, in the range of 20% or better. This relationship can be considered a generally applicable multi-variate discharge rating because it includes the fundamental elements of uniform flow including the width, depth and slope. Additionally, since resistance is not an input variable, all of the necessary data can be measured either directly or remotely. However, equation (2-9) is fundamentally limited by the data used to develop it and therefore cannot be said to be generally applicable in all situations. In addition, because of this limitation, specific knowledge of the variation in the coefficient or exponents of the equation as they may relate to known channel conditions cannot be incorported into the model.

An equation similar to (2-9), which assumes that resistance is a function of the channel slope and geometry, can also be developed for situations where depth cannot be effectively measured, but velocity could be, such as in channels where there is substantial bed movement or bottom debris. The equation is developed by equating (2-1) with a general uniform flow equation such as equation (2-9), solving for the depth in terms of W, V and S, and then substituing this back into (2-1). Carrying through these operations yields an equation of the form:

 $Q = cW^{b}V^{f}S^{g}$ (2-10)

In many situations it is difficult to establish the hydraulically meaningful channel slope that should be used in a theoretically or statistically based equation. Davidian (1984) suggests that a hydraulically meaningful slope should be measured over a reach length on the order of 75 times the water depth. However, the water-surface slope in a channel reach may vary spatially and temporally due to unsteady and non-uniform flow conditions (Davidian, 1984), and because of this, the reach length and timing associated with the slope measurement can alter the "true" uniform hydraulic slope associated with a particular discharge and channel geometry. Thus, consistent definitions of channel and water-surface slope will be important in attempting to apply equations involving those quantities. Given the potential difficulties of consistently measuring a water-surface slope that is hydraulically meaningful, a slope index may be used that considers the slope to be a constant rather than a variable. Such an index could be the topographic slope of the channel and thus might be related to channel morphology.

Alternatively, a relationship between discharge and an index velocity can be developed (Rantz, et al., 1982) which eliminates the slope variable. Since the average velocity in a channel is generally considered to be proportional to the square root of slope and 2/3 power of the depth via the Manning equation (or to the square root of slope and depth via the Chezy equation), the mean velocity could be substituted for the depth and slope to obtain a width-velocity relationship that avoids the need to measure depth and slope but that still provides estimates over a wide range of flow conditions. The form of this equation would be

$$Q = cW^{h}V^{i}$$
(2-11)

where c is a coefficient, and the exponents h and i reflect the relationships between depth and both width and velocity.

Measurement of Hydraulic Variables from Remote Platforms

Few studies have attempted to estimate river discharge entirely from satellite and/or other remotely obtained information, although the potential has been pointed out (Koblinsky, et al., 1993). Estimating the discharge in rivers via equations (1-1), (1-5), (1-7), (1-9), (1-10) and (1-11) requires a measure of the water-surface width, depth and water velocity, and/or river channel

information including the water-surface slope, bank-full width and bank-full depth. The channel resistance is not a directly measurable quantity in the sense that it cannot be measured using an instrument, however it is related to the other geometric variables of the channel (Leopold et al., 1964; Bray, 1978; Dingman and Sharma, 1997) or can be evaluated by comparison with channel charcteristics where resistance values are known.

Satellite-based sensors and other remote data sources can be used to determine channel and water-surface width and water-surface area, water-surface elevation, channel slope and channel morphology (Table 2.1). In addition, there is a possibility that surface velocity can be measured at discrete locations across the river channel (Vorosmarty et al., 1999; Emmitt, personal communication, 2001). The key hydrographic variables that cannot be directly measured from satellite information or other remote data sources are average depth and average cross-sectional velocity. Thus, average depth and average cross-sectional velocity will need to be related, at least implicitly, to stage and surface velocity, respectively, if these variables are used for estimating discharge. Recently, Costa et al. (2000) have demonstrated that surface velocity measurements can effectively be used to estimate the mean velocity in a channel section.

Numerous studies have employed satellite-based imagery to estimate flood inundation area (Smith, 1997). However, few have used satellite derived data to track variability in river and flood stage elevations, and even fewer have attempted to quantitatively estimate river discharge. Landsat 7 multi-Spectral Scanner (MSS), Thematic Mapper (TM), and other visible/infrared spectrum sensors, and synthetic-aperture radar (SAR) imagery from satellites have proven to be useful in tracking changes in water-surface area (and widths) in floodplains and large rivers (Smith,1997). Sippel et al. (1994) determined the inundation area of the Amazon River floodplain using a scanning multi-channel microwave radiometer (SMMR) mounted on the Nimbus 7 satellite. The SMMR sensor measures the microwave emission of the earth's surface,

Table 2.1 Satellite and Remote Data Sources

Data Source	Hydrographic Data Type	Resolution	Relative Cost	Limitations/Advantages	Coverage		
Aerial Photography	Surface features including width and channel shape. Stereoscopic Pairs can also provide surface roughness and slope.	High Resolution Depending on Scale	Low	Limited by inability to see through cloud cover Can provide high resolution and detail and provides direct Interpretation and yields from a range of spectral bands	Spatial - Depends on Scale Temporal - Infrequent coverage		
Visible Spectrum Digital Imagery	Surface features including width, channel shape and coupled with a DEM surface roughness and slope.	Depends on sensor, platform and orbital characteristics Aerial Imagery such as Emerge(1) Photography can be 1m or less Satellite Based Imagery such as Landsat 7 typically 10m or less	Moderate to high depending on coverage (large areas are expensive)	Limited by inability to see through cloud cover Provides direct interpretation and yields information from a range of spectral bands	Spatial - Can be large depending on desired resolution Temporal - Depends on orbital period and weather		
Radar Imagery	Surface features including width, channel and roughness and used with interferometric methods can provide slope and possibly surface velocity using SAR with interferometry.	Space based 10m to 30m SAR surface velocity (not verified). Higher resolution with aerial	High.	Interpretation may be difficult Can see through cloud cover and yields information from a range of spectral bands	Spatial - can be 150 km X 150 km or less depending on desired resolution. Temporal - depends on orbital period.		
Radar Altimetry	Elevation at discrete points which can be used to determine water surface heights (stage) and and possibly slope.	Space based elevations typically 0.5 m but possible to 10 cm Higher resolution with aerial	High	Limited range of information Can see through cloud cover	Provides discrete point data with coverage that depends on the orbital period (frequency of repeat orbits).		
Lidar	Surface velocity, water surface slope and stage.	Possible to 10cm/s for velocity (not verified) 5 cm elevation	Not evaluated	Limited by cloud cover and range of information is limited Interpretation of return may be simpler than radar	Provides discrete point data with coverage that depends on the orbital period (frequency of repeat orbits).		
Topographic Maps and GIS	Static channel shape and slope and other static surface features.	Depends on Scale	Low	Temporally limited because it is a static data base Interpretation is direct	Spatial - depends on scale Temporal - static.		
EPA Reach data base and other comparable data bases	Potentially reach lengths, channel types and other channel feaatures	Depends on data	Low	Temporally limited because it is a static data base interpretation is direct	Dependent on available data		

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which can be correlated to ground saturation and open water at the surface; however the resolution is low, on the order of 25 km, and the signal is attentuated by atmospheric moisture. Vörösmarty et al. (1996) correlated SMMR signals with discharge in the Amazon River, thus developing a discharge rating based on general moisture conditions within the basin. Brakenridge et al. (1994) used SAR images from ERS-1 to delineate flood inundation area coupled with topographic information to determine water-surface elevations during the 1993 Mississippi floods. Horritt (2000), Bates and DeRoo (2000) and Horritt et al. (2001) have used SAR imagery to delineate flood boundaries and calibrate river hydraulic models.

A method to estimate river discharge from aircraft has been developed that couples ground-based channel geometry information with surface velocity measurements made by photographing floats or other tracking substance introduced into the river by aerial drop (Kuprianov, 1978). The mean velocity is estimated from the surface velocity using an assumed vertical velocity distribution, and the channel geometry is measured on the ground and assumed to be constant thereafter. This method has a reported accuracy of 10% or better where winds are moderate (2 to 3 m/s) and water-surface velocity is in the range of 1 to 2 m/s. Although this method relies on instruments introduced into the streamflow (the floats) to measure velocity, the measurement is made entirely from a remote platform (the aircraft) once the appropriate ground measurements are made.

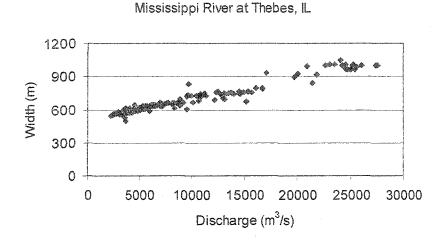
Smith et al. (1996) estimated the discharge in three braided glacial rivers using reachaveraged water-surface area obtained from RADARSAT SAR imagery. That study correlated the water-surface area in braided reaches (lengths on the order of 10 km) with discharge obtained from existing ground-based gaging stations to derive power-function discharge ratings that use effective width (water-surface area divided by the reach length) as the predictor variable. The accuracy of the ratings varied in each river, ranging from 1.5% (for 11 estimated values) to 54%

(19 estimated values). A single best-fit power function was also developed as a general rating for all of the rivers. Error associated with this function was much larger, providing accuracy only within a factor of 2 (100% error).

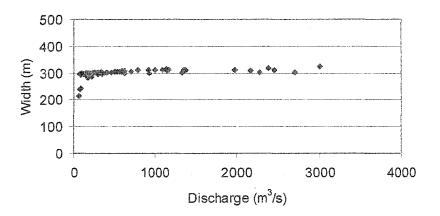
Smith et al. (1996) also pointed out that the total sinuosity was an important discriminator between the rivers studied. To test the predictive power of sinuosity, we used the data from Smith et al. (1996) to develop a general multi-parameter power function with reach-averaged width as the dynamic variable and the average sinuosity as a channel constant to predict discharge in all the three braided rivers (data not shown). This relation reduced the standard error of the estimate by 30% and improved the slope of the regression compared to the width-only relationship reported by Smith et al. (1996). These results suggest that morphologic features of a river channel that can be observed remotely and that are related to the energy-dissipation process may be useful for remote- discharge estimation. These features may include, in addition to channel sinuosity, meander wavelength, meander radius of curvature, bankfull width, width/depth ratio, and others (Leliavsky, 1966; Dury, 1976; Osterkamp et al., 1983; Rosgen, 1994).

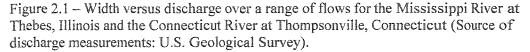
In principle, it would seem that a width-discharge rating might be developed for a wide range of rivers, because width generally increases with increasing discharge. However, in nearly rectangular channels, or channels with highly irregular cross-sectional shape, width may change very little or in a highly non-linear way with discharge. This is illustrated in Figure 2.1, which shows changes in width with discharge over a range of flows for the Mississippi River at Thebes, Illinois, and the Connecticut River at Thompsonville, Connecticut (USGS, 2001). The graphs demonstrate that in the Mississippi at Thebes width does indeed change linearly with discharge and could be used as an index to flow variation, whereas in the Connecticut at Thompsonville it changes non-linearly with very little change at higher discharges. Similarly, width changes very little with discharge in the Amazon River narrows at Obidos (Oltman, 1968). This condition may

be common at many locations in larger rivers, and suggests that multi-variate discharge ratings that



Connecticut River at Thompsonville, CT





reflect general hydraulic relationships would be more universally applicable than relations based only on width. This also suggests that the best locations for evaluating river discharge from space, where width is the most readily observed hydraulic variable, are those channel reaches where width variation with dischage is most pronounced (Smith et al., 1997).

Measurement of Width

Both channel width and water-surface width (and also the water-surface area) can be measured from a variety of sensors and imagers mounted on satellites and aircraft (Table 2.2), including panchromatic and infrared imagers, digital photography, and synthetic aperture radar (SAR) (Barrett, 1998; University of Wisconsin Environmental Remote Sensing Center, 2001). The resolution of satellite-mounted digital panchromatic sensors is within the same range as aircraft- mounted sensors, indicating that satellite observation of width, because of the larger coverage, may be the preferred method to obtain this type of data. SAR is the only sensor that can measure the width in any atmospheric condition (Smith, 1997).

Panchromatic imagers have spatial resolution as fine as 1 or 2 m and SAR imagers as fine as 10 m (University of Wisconsin Environmental Remote Sensing Center, 2001). However, the accuacy of a sensor to observe surface-area or width change is not limited solely by the resolution. Improved measurement accuracy can be obtained by averaging resolution errors over the observed reach, such that relatively coarse resolution imagery may provide measurement accuracy significantly better than the resolution may imply. In addition, the ability to use different sensor bands to observe the surface area, each with its own observation qualities, can be used to complement each other and achieve potentially greater measurement accuracy.

The key objective of measuring surface area and width, as for any dynamic variable, is to detect change from one scene to another. Change detection is not necessarily restricted to resolution because identification of a pixel as either water or not-water depends on sub-pixel size qualities that are also detected. It is difficult to evaluate the true "error" that might be associated with the measurement of width and surface area from remote platforms, especially considering

Table 2.2	Examples of River Hydraulic Variables Observable from Existing Space Based Sensors
-----------	--

Observed Variable	Satellite/Sensor	Data Type	Data Resolution	Repeat Observation Frequency	Observational Issues
Water Surface Width and Channel Morphology	TERRA/ASTER	Visible Infrared Thermal Infrared Shortwave Infrared	15m	1-2 days	Cannot detect through clouds Banks may be obscured by vegetation and shadows
	EROS A & B	Visible to Infrared	1.5m	daily (with a constellation of satellites)	Cannot detect through clouds Banks may be obscured by vegetation and shadows
	ERS2	SAR	12-26m	6 days	Banks may be obscured by vegetation and wet soils
	SPOT 4	Panchromatic visible	10m	26 days	Cannot detect through clouds Banks may be obscured by vegetation and shadows
	LANDSAT 7	Panchromatic visible	15-60m	16 days	Cannot detect through clouds Banks may be obscured by vegetation and shadows
	IKONOS	Panchromatic visible	1-4 m		Cannot detect through clouds Banks may be obscured by vegetation and shadows
	RADARSAT	SAR	8-30m	1-6 days	Banks may be obscured by vegetation and wet soils
Water Surface Stage and slope	TOPEX/Posiedon	Radar Altimeter	10cm	10 days	Repeat observations limited to large rivers Using Interferometry coupled with Altimetry (unproven)
	RADARSAT	JSAR	1cm	1-6 days	
Water Surface Velocity		Lidar Radar SAR	NA NA NA		Signal obscured by surface wind and waves Sensors have not been tested in rivers

that the error would be a function of many factors including the observed reach length, the resolution, the spectral bands used for pixel identification and processing technique. Based on these considerations, we could easily assume that a "typical" measurement error for a reach-averaged width measurement could be on the order of 10 m or less.

Width estimates using any imager would be subject to errors associated with vegetation obscuring the water's edge and the bank and, in the case of SAR, wet ground, vegetation, wind roughening and rocks can also obscure the edge of water. With a combination of SAR imagery (to observe through cloud cover) and digital panchromatic imagery, it is conceivable that width could be observed with near global coverage on a repeat cycle of nearly one week.

Measurement of Stage and Depth

Radar altimetry has been successfully used to track water-level elevations (stage) in large rivers, lakes and floodplains. Koblinsky et al. (1993) were able to use Geosat altimeter data to track elevation changes at several locations in the Amazon River basin with an accuracy on the order of 0.7 m. The altimeter footprint ranges from 0.2 to 2 km so target must be at least this wide to obtain a return unique to the water body. More recently, Birkett (1998) and Birkett et al. (2002) measured water-surface elevation changes in several rivers (including rivers in the Amazon Basin, the Okavango River, the Indus River and the Congo River) using water-surface elevation data obtained from the TOPEX/Posiedon (T/P) altimeter and reported an accuracy ranging from 11 to 60 cm.

With the currently deployed T/P altimeter, the theoretical minimum river width that can be observed ranges from 0.58 to 1.16 km (Birkett et al. 2002) with accuracies ranging from 10 cm to 1 m. However, it is possible that the altimeter can obtain accurate water-surface elevation

measurements on rivers with widths as low as 50 meters by altering the signal-filtering algorithms (Ernesto Rodriquez, personal communication, 2001). The accuracy of the T/P altimeter (and altimeters in general) is strongly dependent on the surface conditions being observed (Birkett et al., 2002). Laser altimeters (lidar) such as GLAS (NASA, 1997), which will be deployed on ICESAT, can track elevation changes to within 15 cm, and thus may provide significantly higher accuracies in river environments than possible with currently deployed radar altimeters.

Depth cannot be measured directly from remote data (Table 2.1 and 2.2). Thus, this variable will need to be estimated, at least in part, from measurements of stage coupled with other observable characteristics of the channel. Depth could be derived from repeated observations of stage over a wide range of flow conditions provided accurate topographic data or altimetric measurements of sufficient accuracy were available to determine the exposed bank elevation at each observed water level. However, in large rivers low flow depths may never be observed, necessitating the estimation of the bank-full depth or other depth reference so that stage measurements can be converted to average water depths.

An estimate of the depth can be developed from a time series of stage measurements provide it is long enough to identify the bottom (zero flow) elevation, or the elevation of the top of bank. If the zero flow elevation (Z_0) is known, then computation of depth from observations of stage can be made directly. If the top of bank elevation is known, then Z_0 could be estimated if width observations are also available, by statistically relating the width and stage observations through linear regression, with the intercept being equal to Z_0 . Another approach would be to assume a specific cross-section shape (e.g. a parabola) and then solve for Z_0 given stage and width observations that include the top of bank elevation and bankfull width.

Measurement of Water-Surface Slope

Water-surface slopes on the Amazon River and some of its larger tributaries have been estimated by Mertes et al. (1996) and Dunne et al. (1998) using SEASAT and Birkett et al. (2002) using T/P. All of these estimates have been made from sea-level (the mouth of the Amazon) to an inland point hundreds or thousands of kilometers upstream. The long reaches that were evaluated minimized the impact of altimeter accuracy on the estimates. Birkett et al. (2001) were also able to observe temporal changes in water-surface slope in the mainstem of the Amazon over long reaches.

In the Amazon River at Obidos, the water depth is on the order of 40 to 50 m and hydraulically meaningful water-surface gradients are on the order of 1 cm/km (Oltman, 1968). Thus, given an optimistic altimeter error of 10 to 20 cm, a reach of 5 to 10 km could conceivably result in slope estimates ranging from negative values to 8 times the actual value. This suggests that slope information obtained from the current generation of altimeters would not provide sufficient spatial resolution to be hydraulically meaningful. Averaging the slope obtained from a large sampling of slope measurements may be the most meaningful slope information that can be considered reliable.

One approach to obtaining more accurate water-surface slope measurements could be through the use of interferometric SAR. With this technique, water-surface elevation changes on the order of 1 cm can be detected in large rivers and flooded areas (Alsdorf et al. 2000, Alsdorf et al. 2001) and, when coupled with high resolution topographic information, could be used to estimate water-surface slopes across a flooded area as well as within a river. Laser altimeters may also provide a means to accurately measure hydraulically meaningful water-surface slopes because the altimeter could simultaneously measure the elevation at two points in a channel reach.

Measurement of Water-Surface Velocity

Surface velocity in rivers is potentially measurable from satellites with doppler lidar or radar. However, surface winds and waves on the water body could significantly interfere with the measurement (Vorosmarty et al. 1999), although observing limitations have not been fully evaluated. Theoretical (e.g. the Prandtl-von Karman velocity profile) or empirical relations would be required to translate surface velocity to average velocity; however, surface velocity could potentially provide an index of average flow velocity and hence be directly useful in predicting discharge.

Based on information supplied by Emmitt (personal communication, 2001), a satellite mounted doppler lidar sensor that could observe surface velocity would have a footprint of approximately 10 m with 75 m between observations along a track, and have a measurement accuracy on the order of 0.1 m/s. Given these specifications, the lidar could observe two to three surface-velocity "points" across a 200 meter wide river reach. There is no guarantee that the satellite track would cross the river reach perpendicular to the flow, thus the point measurements may be skewed across the channel. This should not be a problem provided the distance to each bank can be evaluated from another source (e.g. a concurrent image of the channel and knowledge of the satellite track) and the correction made. Despite the potential limitations, if surface velocity were measured and can be used to infer average velocity, there is the potential for measuring all elements of equation (2-1) simultaneously and thus enabling direct calculation of discharge.

Observation of Channel Morphology

Valley and channel features such as the channel sinuosity, channel slope, meander length, and meander radius of curvature can be observed from a variety of data sources, including visible- and infrared-spectrum images, SAR images, DEMs, and topographic map information. Because these features are considered relatively stable over short time frames, the frequency and timing of observations is not a limiting factor, and therefore high-resolution panchromatic images could be used to measure them when weather conditions permit.

Estimating River Discharge

Based on the above discussion, there is a possibility that the hydraulic elements of equation (2-1) can all be measured simultaneously from satellites. If so, discharge could be calculated directly, with an accuracy dependent on the accuracy and precision of the individual measurements of water-surface width, surface velocity, and stage and of the estimations of mean velocity and mean depth from observations of surface velocity and stage. Because there is a potential that stage or surface velocity will not be observed with confidence (e.g. under strong winds or where topography obscures the signal) there will be many situations when all three of the key variables cannot be observed at the same time. In these situations statistically based relationships such as described by equations (2-9), (2-10) and (2-11) may be useful.

Statistically Based Estimation Methods

To explore the predictive characteristics of different combinations of potentially observable (or estimated) river-hydraulic variables, a set of generally applicable river-discharge estimation equations (models) were developed based on equations (2-5), (2-7), (2-9), (2-10) and

Table 2.3	Range of H	lydraulic I	Parameters	in Calibra	tion and Va	lidation Da	ata Sets
Parameter	Symbol	<u>Units</u>	Mean	<u>Stdev</u>	<u>Coeff. Var.</u>	<u>Maximum</u>	Minimum
Calibration Data N = 506							
Discharge	Q	m³/s	1585	12260	7.74	216000	0.01
Top Width Average Depth (Hyd. Radius)	W Y	m m	146 2.48	206 3.56	1.41 1.44	2290 48.03	2.90 0.18
Average Velocity Water Surface Slope (average)	V S	m/s m/m	1.12 0.00278	0.66 0.00572	0.59	5.10 0.04	0.02 0.0000007
		111/111	0.00210	0.00012	.2.00	0.04	0.0000007
Valdiation Data N = 506							
Discharge	Q	m³/s	1666	13184	7.91	283170	0.05
Top Width Average Depth (Hyd. Radius)	W Y	m m	158 2.73	211 3.53	1.34 1.29	2300 50,33	5.40 0.14
Average Velocity	V	m/s	1.13	0.61	0.54	3.61	0.07
Water Surface Slope (average)	S	m/m	0.00243	0.00474	1.95	0.04	0.0000007

(2-11). The models were derived using multiple-regression analysis of hydraulic data from 1,012 discharge measurements in 102 rivers in the United States and New Zealand, including 4 measurements from the Amazon River at Obidos. The data include a wide range of river conditions (Table 2.3) and was randomly divided into a calibration data set and a validation data set each containing 506 measurements. The 4 Amazon River measurements were equally divided between the calibration and validation data sets.

The data base includes 569 discharge measurements with reach averaged (generally three or more cross-sections representing a reach length 5 or more times the width) values of watersurface width, average water-surface depth, average velocity, and water-surface slope measured concurrently with the discharge. These data were obtained from Barnes (1967), Hicks and Mason (1992) and Coon (1998). Because these data are reach-averaged, the hydraulic-geometry and velocity values are representative of the energy and resistance relationships within the channel, and less a reflection of conditions at a single cross-section. In addition, the reported width approximates the water-surface area divided by the reach length, consistent with Smith et al. (1996). To this extent, the data are consistent with what might be obtained from remote imagery capable of providing reach averaged width, channel slope, and surface velocity.

The reach-averaged data include only two discharge measurements greater than 10,000 m³/s. In order to include more large flows in the data base, 443 additional measurements representative of the larger rivers of North America were obtained from the USGS (2001) and data from four measurements for the Amazon River at Obidos, Brazil were also included (Oltman, 1968; Dury, 1976). These large discharge measurements are not reach averaged, and therefore have a certain incompatibility with the rest of the data in the data base. However, it is anticipated that hydraulic variability between the measurement section and the reach as a whole is not large, and that the number of observations will average out the variability. In addition,

inspection of the channel characteristics at the measurement sections for these rivers do not indicate any channel constraints from bridges or other structures (however, some of the river hydraulics may be affected by bedrock outcrops and canyons).

The discharge data are all in-bank and do not represent overbank-flow conditions. In general, the data were obtained from relatively straight single-thread channel sections, and therefore do not necessarily reflect the hydraulic conditions in more complex or less constrained channel patterns. Because of this, the derived regression coefficients may be biased towards these types of channels, reflecting typical relationships between width and depth, depth and resistance, and velocity and depth that would be found in straight channels. However, because the models are based on, and include, the fundamental hydraulic variables of uniform flow, the resultant regression equations are considered to remain generally representative of uniform-flow relationships for any defined channel.

Similar to Dingman and Sharma (1997), the predictive models were assumed to be multiplicative. The form of the prediction equations (models) that were developed are based on Equations (2-5), (2-7), (2-9), (2-10) and (2-11) as follows:

Model 1 (Equation 2-9):	$Q = c_1 W^a Y^b S^d$	(2-12)
Model 2 (Equation 2-11):	$Q = c_2 W^e V^f S^g$	(2-13)
Model 3 (Equation 2-10):	$Q = c_3 W^e V^f$	(2-14)
Model 4 (Equation 2-5):	$Q = c_4 W_m{}^g Y_m{}^h S^i Y^j$	(2-15)
Model 5 (Equation 2-7):	$Q = c_5 W_m^{\ k} Y_m^{\ l} S^m W^n$	(2-16)

Models 1, 2, 4 and 5 use the water-surface slope as a prediction variable. However, the USGS discharge measurement data base does not include slope as a measured parameter. Therefore, a channel slope for these river stations was measured manually from 1:24,000 scale USGS topographic maps over one contour interval. This results in a constant slope value for all of the

flow measurements at a particular river station, implying slope as a geomorphic characteristic of the river.

The implication of using a constant slope is explored by comparing two realizations of Model 1 developed from the reach averaged data base, that includes a unique measured slope for all discharge measurements (minimum of five) at each river station (excluding the Barnes (1967) data, which includes only one flow measurement at each station). The first model uses slope as a dynamic variable and the second uses a slope obtained by averaging all of the measured slopes over the entire discharge range at each river station. The comparison shows nearly identical regression models (Table 2.4). Based on this comparison, we conclude that using an average slope, or a channel slope obtained from topographic information that is a constant for a river reach, can be used in lieu of a measured slope, thus obviating the need to track water surface slope as a dynamic prediction variable.

These results also indicate that the USGS flow measurement data, which includes width, average depth, average velocity, and discharge (but not slope) can be combined with the reach averaged data base (which includes a measured slope) using a slope measured from 1:24,000 scale USGS topographic maps for each station. In the remainder of this paper, all of the regression models and all discussion of slope as a prediction variable assume a constant slope for each river station, developed either as an average of many measured values, or obtained from topography.

Using the entire calibration data set (N=506), the following regression models are developed (in SI units):

Model 1:	$Q = 7.22 W^{1.02} Y^{1.74} S^{0.35}$		(2-17)
Model 2:	$Q = 0.09 W^{1.21} V^{1.53} S^{-0.30}$	26	(2-18)

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COMPARIS		EGRESSION 	NODELS	USING CON	STANT AND V	ARIABLE	SLOPE					
Regression	1 Statistic	ģ										
		Std. Error		Relative Residual	Log Residual	Actual Residual	Coefficients					
Z	'N	Regression		(0* - 0)(0	(logQ* - logQ)	(<u>0* - 0</u>)		<u>Value</u>	Std. Error	Upper 95%	Lower 95%	đ
545	0.95	0.19	Mean	0 12	0 001	(m²/s) 14.6	Intercent	4.57	1 1.4	5 Q7	3 50	<0.0001
) - 1		2	Stdev	0.69	0.189	113.8	M	1.18	0.03	1.24	1.12	<0.0001
							≻	1.74	0.04	1.82	1.67	<0.0001
							ა	0.35	0.02	0.38	0.32	<0.0001
545	0.95	0.20	Mean	0.14	-0.001	11.4	Intercept	4.60	1.16	6.21	3.42	<0.0001
			Stdev	0.87	0.202	113.0	N	1.17	0.03	1.23	1.10	<0.0001
							≻	1.78	0.04	1.86	1.69	<0.0001
							ა	0.35	0.02	0.39	0.31	<0.0001
	Regression 545 545	Regression Statistic R^2 R^2 545 0.95 545 0.95 545 0.95	Regression Statistics Statistics N R ² Stat. Error 545 0.95 0.19 545 0.95 0.20	Regression Statistics N R ² Std. Error 545 0.95 0.19 Mean 545 0.95 0.20 Mean 545 0.95 0.20 Mean	Regression StatisticsNRdStd. ErrorRelativeNR ² Regression(0*- (0)/0)5450.950.19Mean0.125450.950.20Mean0.145450.950.20Mean0.14	Regression StatisticsN \mathbb{R}^2 Std. Error Std. ErrorRelative ResidualLog Residual545 0.95 0.19 Mean 0.12 0.001 545 0.95 0.19 Mean 0.12 0.001 545 0.95 0.19 Mean 0.12 0.001 545 0.95 0.20 Mean 0.14 -0.001 545 0.95 0.20 Mean 0.14 0.001	RelativeLogActualNRd. ErrorStd. ErrorResidualResidualNR 2 Regression(Ω^{\star} -Q)/Q($IogQ^{\star}$ -IogQ)(Ω^{\star} -Q)5450.950.19Mean0.120.00114.65450.950.19Mean0.120.00114.65450.950.20Mean0.14-0.00111.35450.950.20Mean0.14-0.00111.4	Std. Error Residual Residual Residual Residual (<u>0.19</u> Mean 0.12 0.001 Stdev 0.69 0.189 0.20 Mean 0.14 -0.001 Stdev 0.87 0.202	Std. ErrorRelative ResidualLog ResidualActual ResidualCoefficientsRegression $(\Omega^* - Q)/Q$ $(log Q^* - log Q)$ $(\Omega^* - Q)$ 0.19Mean0.120.00114.60.19Mean0.120.189113.80.19Mean0.120.00114.60.10Stdev0.690.189113.80.20Mean0.14-0.00111.40.20Stdev0.870.202113.00.20Stdev0.870.202113.0	Std. ErrorRelative ResidualLog ResidualActual ResidualCoefficients Value3.19Mean 0.12 0.001 14.6 Intercept 4.57 0.19Mean 0.12 0.001 14.6 Intercept 4.57 0.20Mean 0.12 0.001 11.3 W 1.18 0.20Mean 0.189 0.189 113.8 W 1.74 0.20Stdev 0.69 0.202 113.0 W 1.78 0.20Stdev 0.87 0.202 113.0 W 1.17 0.20Stdev 0.87 0.202 113.0 W 1.17 0.20Stdev 0.87 0.202 113.0 W 1.17 0.35Stdev 0.87 0.202 113.0 W 1.17	Std. ErrorRelative ResidualLog ResidualActual ResidualCoefficients Value3.19Mean 0.12 0.001 14.6 Intercept 4.57 0.19Mean 0.12 0.001 14.6 Intercept 4.57 0.20Mean 0.12 0.001 11.3 W 1.18 0.20Mean 0.189 0.189 113.8 W 1.74 0.20Stdev 0.69 0.202 113.0 W 1.78 0.20Stdev 0.87 0.202 113.0 W 1.17 0.20Stdev 0.87 0.202 113.0 W 1.17 0.20Stdev 0.87 0.202 113.0 W 1.17 0.35Stdev 0.87 0.202 113.0 W 1.17	Relative std. Error Log Residual Actual Residual Actual Residual Actual Residual Relative Residual Log (0.1-0)(Q. (log(Q.*-log(Q)) Coefficients 0.19 Mean 0.12 0.001 14.6 Intercept 4.57 1.14 5.97 0.19 Mean 0.12 0.189 113.8 W 1.18 0.03 1.24 0.19 Mean 0.12 0.189 113.8 W 1.18 0.03 1.24 0.10 14.6 Mercept 4.57 1.14 5.97 0.35 0.02 0.38 0.20 Mean 0.18 0.183 Y 1.74 0.04 1.82 0.20 Mean 0.14 -0.001 11.4 Intercept 4.60 1.16 6.21 0.20 Stdev 0.87 0.202 113.0 Y 1.78 0.35 0.35 0.39

SON OF REGRESSION MODELS USING CONSTANT AND VARIABLE SLOP

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Model 3:	$Q = 0.23 W^{1.46} V^{1.39}$	(2-19)
Model 4:	$Q = 3.55 W_m^{-1.19} Y_m^{-0.84} S^{0.30} Y^{2.17}$	(2-20)
Model 5:	$Q = 1.07 W_{m}^{-1.61} Y_{m}^{-1.11} S^{0.08} W^{2.65}$	(2-21)

The values for Y_m and W_m used in the regression analysis are obtained as the maximum value for all of the flow measurements at each river station, and thus are constant for each station. The possiblity that the Amazon River measurements skewed the regression results was evaluated by removing them from the calibration data set and re-running the regression analysis. It was found that the Amazon data did not significantly impact the regression results.

The four regression models varied in their ability to describe the observed data. Comparative statistics between the models are shown on Table 2.5 and indicate that Models 1, 2 and 3 perform comparably well, and that Model 4 does not perform as well as Models 1, 2 and 3 but is better than Model 5. The intercept and coefficient of the slope for Model 5 are not significantly different than zero at the 95% confidence level. Since the form of the model is based on the Manning equation, slope would be expected to be a significant predictor variable as it is in Model 1. The reason for this outcome may be due to the fact that width by itself is not an especially good predictor variable at many specific river stations (as indicated by Figure 2.1), and thus a constant slope at each river station does not contribute to explaining at-a-station variation. The standard error of the estimate (standard deviation of the log residuals) for Model 5 is nearly twice as large as the standard errors for Model 1, 2 and 3, and indicates that 67% of the predictions using this model fall within a wide margin (factor of 2.75). Because of the relatively poor performance of Model 5 it is not evaluated further.

For comparative purposes, Table 2.5 also lists regression results for three single-variate models that use each element of equation (2-1) (W, Y and V) to predict Q. These models indicate that depth, by itself, predicts discharge better than width and has a lower standard error than

TABLE 2.5	REGRESS	ION MODEL	REGRESSION MODEL COMPARISON	N					
Model	Regressio	Regression Statistics							
Model	R ²	Regression	Regression Coefficients	<u>Value</u>	Std. Error	Std. Error upper 95% lower 95%	<u>lower 95%</u>	t stat	a
would 1 logG = 0.86 + 1.02logW + 1.74logY + 0.35logS (G = 7.02W142×174_c035,	0.95	0.23	Intercept	0.86	0.06	0.98	0.73	13.40	< 0.0001
			≤ ≻	1.02	0.03 0.04	1.07	0.96 1.66	35.16 42.07	<0.0001 <0.0001
Model 2			S	0.35	0.02	0.40	0.31	15.27	<0.0001
logQ = -1.06 + 1.21logW + 1.53logV - 0.30logS	0.97	0.19	Intercept	-1.06	0.04	-0.98	-1.14	-26.22	<0.0001
$(0 = 0.09W^{1.2}V^{1.2}S^{0.2}V^{0.2})$			32	1.21	0.02	1.25	1.16	53.65	<0.0001
			> w	-0.30	0.02	-0.26	-0.33	51.85 -15.29	<0.0001
woorei 3 logQ = -0.63 + 1.46logW + 1.39logV	0.95	0.23	Intercept	-0.63	0.04	-0.56	-0.70	-17.93	<0.0001
$(Q = 0.23W^{1.50}V^{1.50})$			≥>	1.46 1.39	0.03	1.45	1.32	40.98 80.57	<0.0001
Model 4			•	-	4	2		0.00	20000
logQ = 0.55 + 1.19logWm - 0.84logYm + 0.3logS + 2.17logY	0.93	0.28	Intercept	0.55	0.11	0.76	0.34	5.19	< 0.0001
$(Q = 3.55 W_m^{1.10} Y_m^{2.000} X_{2.00}^{2.00} Y_{2.00})$			W "	1.19	0.06	1.29	1.08	21.42	<0.0001
			۲	-0.84	0.11	-0.63	-1.06	-7.64	<0.0001
			s :	0.30	0.03	0.37	0.24	9.48	<0.0001
Mindal S			٢	71.7	0.05	17.7	2.06	42.11	<0.0001
logQ = 0.03 - 1.61logWm + 1.11logYm + 0.08logS + 2.65logY	0.83	0.44	Intercept	0.03	0.17	0.36	-0.30	0.17	0.86
$(Q = 1.07W_{m}^{-1.61}Y_{m}^{-1.11}S^{0.08}W^{2.65})$			۳ ۳	-1.61	0.16	-1.30	-1.92	-10.24	< 0.0001
			۲ ۳	1.11	0.16	1.42	0.79	6.86	< 0.0001
			s V	0.08 2.65	0.05 0.13	0.18 2.90	-0.02 2.39	1.53 20.00	0.13 <0.0001
<u>Comparative Single-Variate Models</u>									
u-w logQ = -0.98 + 1.62logW	0.79	0.49	Intercept	-0.97	0.07	-0.84	-1.11	-13.74	<0.0001
$(Q = 0.11W^{1.62})$			M	1.62	0.04	1.69	1.55	43.85	<0.0001
Q-Y 1040 1 40 + 0 46102V	7 8 V	0 43	Intercent	1 49	0 0	1 54	1 45	70.33	<0.0001
$(Q = 30.90Y^{2.46})$	5		Y	2.46	0.05	2.55	2.37	51.64	<0.0001
0-r.									
logQ = 2.07 + 1.96logV (Q = 117.49V ^{1.96})	0.33	0.87	Intercept V	2.07 1.95	0.04	2.15 2.20	2.00	52.92 15.90	<0.0001

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Model 5. The standard error is the standard deviation of the log residual, and its antilog is representative of the standard deviation of the fractional errors between the predicted and observed values on a log scale, and can be used as an approximation of the percent expected error provided there are not too many extreme values in the residual distribution. Using this approach (for comparative purposes), the depth by itself would be expected to predict discharge to within a factor of 2.7 67% of the time, width by itself would be expected to predict discharge within a factor of 3 67 % of the time, and velocity by itself would predict discharge within a factor of 7.4 67% of the time. Relative to depth and width, velocity by itself is a poor indicator of discharge.

The validation statistics for Models 1, 2, 3 and 4 and the Dingman and Sharma Model (Equation 2-9) are compared in Table 2.6. Comparative statistics include the mean and standard deviation of the following quantities:

Relative Residual = $(Q' - Q)/Q$	(2-16)
Log Residual = log(Q') - log(Q)	(2-17)
Actual Residual = $Q' - Q$	(2-18)

In addition, the number of predictions within a specified percent-error interval (percent different than the observed) are also tabulated for 20%, 50% and 100% error. Figure 2.2 shows the predicted discharge (Q') plotted against the observed discharge (Q) for each model, along with an upper- and lower-envelope curve defined by the +/- 50% error in the observed value.

The log and actual residuals indicate that Model 1 and the Dingman and Sharma model tend to over-predict discharge and Models 2, 3 and 4 tend to under-predict discharge (Table 2.6 and Figure 2.2). Model 1 shows the least overall prediction bias, and the Dingman and Sharma model has the highest. The mean relative error indicates the average percent error of the predictions. Model 2 performs the best in this regard, with an average relative error of 10%. Average relative error for Model 1, 3 and 4 are less than 20%. The antilog of the mean of the log

REGRESSION MODEL VALIDATION STATISTICS

Model

TABLE 2.6

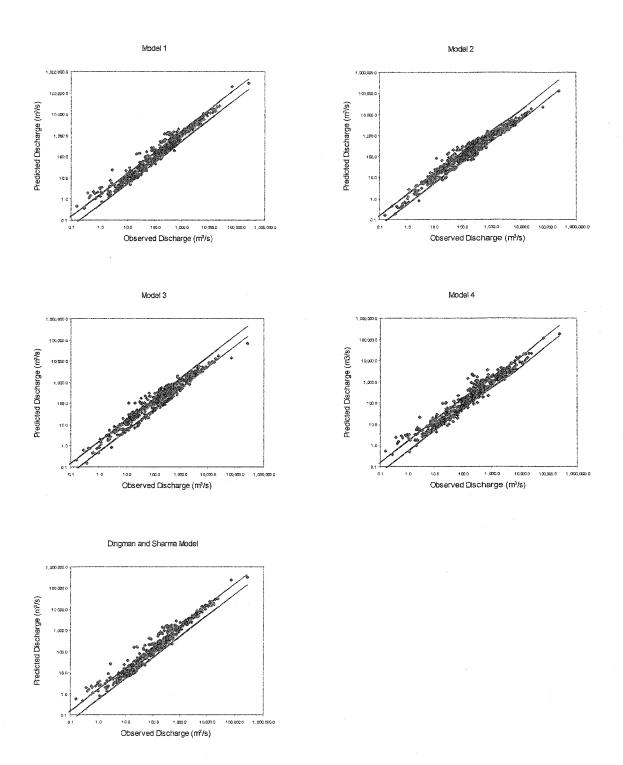
Validation Statistics

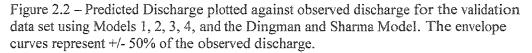
INICACI	a minerio i	I DECENDENCO			Percent of	predictions	
		Relative Residual	Log Residual	Actual Residual	within 20, 5	0 and 100%	6 of the observed
		<u>(Q* - Q)/Q</u>	<u>(logQ* - logQ)</u>	<u>(Q* - Q)</u>	<u>20%</u>	<u>50%</u>	<u>100%</u>
Model 1				(m ³ /s)			
$Q = 7.22W^{1.02} * Y^{1.74} * S^{0.35}$	Mean	0.16	0.004	243	39%	82%	90%
	Stdev	0.81	0.207	5059			
Model 2							
$Q = 0.09W^{1.21}V^{1.53}S^{-0.3}$	Mean	0.07	-0.017	-615	37%	79%	94%
	Stdev	0.58	0.195	7129			*****
Model 3							
$Q = 0.23W^{1.46}V^{1.39}$	Mean	0.10	-0.024	-790	32%	71%	93%
	Stdev	0.71	0.231	9946			
Model 4							
$Q = 3.55W_{m}^{1.19}Y_{m}^{-0.84}S^{0.30}Y^{2.17}$	Mean	0.17	-0.016	-119	28%	73%	89%
	Stdev	0.99	0.243	5333			
Dingman and Sharma Model Q = 4.62W ^{1.17} Y ^{1.57} S ^{0.34}	2.5	A 4A	0.000	700	440/	"7 A O/	000/
$Q = 4.62VV Y S^{-1}$	Mean	0.43	0.092	763	41%	74%	86%
	Stdev	1.01	0.215	7644			

residuals indicates the fractional error between the predicted and observed discharge (the log residual can also be expressed as log(Q'/Q) such that the antilog is the ratio Q^*/Q), which can be regarded as a correction factor. This measure of error shows that Model 1 has the highest mean accuracy (with a ratio of less than 1%), and that Models 2, 3 and 4 all show mean accuracy within 5%. Model 2 shows the least overall prediction error variability, as indicated by the standard deviation of the relative error and the log residual. The error percentiles indicate that Models 1, 2 and the Dingman and Sharma model are comparable.

Inspection of Figure 2.2 indicates that the predictive characteristics of the models vary for different ranges of discharge. These differences are evaluated by comparing the distribution of the relative residual with observed discharge. To facilitate comparison, the mean and standard deviation of the relative residuals have been averaged within four categories of discharge range $(0-10, 10-100, 100-1,000 \text{ and }>1,000 \text{ m}^3/\text{s})$. Models 1, 4 and the Dingman Sharma model tend to over-predict primarily in the low discharge range $(0-10 \text{ m}^3/\text{s})$. This suggests that these models will have the best results in medium to large rivers where discharge typically ranges above 10 m³/s.

The reason for this may be that the relationship between resistance and the channel geometry cannot be fully represented by a single regression intercept (model coefficient). Models 2 and 3 also tends to over-predict discharge in the low range (0-10 and 10-100 m³/s) but also under-predicts in the high discharge range (>1,000 m³/s). This result indicates that Models 2 and 3 would do better if the coefficients varied with discharge, i.e. different model coefficients were calculated for different flow ranges. The Dingman and Sharma Model shows a consistent over-prediction for all flow ranges, which may result because it was developed from a data set with fewer large rivers (also suggesting that statistical models such as these would be improved if they were developed for specific flow ranges).





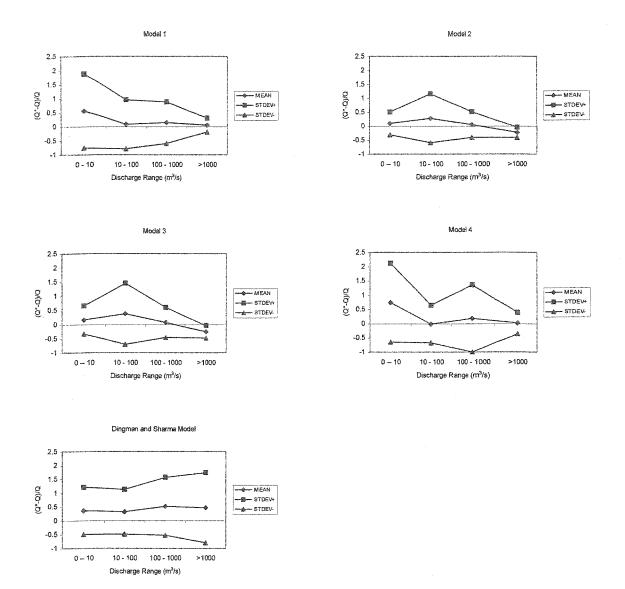


Figure 2.3 – Variation of the mean and standard deviation of the relative residuals averaged within ranges of observed discharge. The upper and lower lines are +/- one standard deviation from the mean. Multiplying the relative residual by 100 gives the percent error. The number of observations in each range are 71, 132, 209, and 94 from lowest to highest.

Prediction variability, as indicated by the upper and lower standard deviation of the relative residuals, is reduced in the highest discharge range for Models 1, 2, 3 and 4 (Figure 2.3). This indicates that model precision is improved for the larger rivers. The Dingman and Sharma

model does not follow this trend, which again may be due to the presence of fewer large rivers in the data base used to develop it. The validation statistics indicate that prediction models based on Models 1, 2, 3 and 4 could all be used as general discharge estimating models, with mean accuracy of less than 20% in all cases. The variability of the estimates would be expected to be within +/- 50% of the acutal value on the order of 2/3 of the time. The prediction accuracy would be improved for medium and large rivers.

As a comparison, under good measurement conditions, the accuracy of a discharge measurement made on the ground with standard techniques is assumed to be in the range of 2 to 4% of the actual value at least 2/3 of the time (Rantz et al., 1982, Herschy, 1998). The accuracy of measurements made using the slope-area method (usually for large discharges that could not be meausred using standard techniques), which is based on after-the-fact measurements of the flow width, depth, energy slope and flow resistance using the Manning or comparable uniform flow equation, are not explicitly known because it depends on field judgement and the quality of the measured data (Kirby, 1987). However it is often reported that good measurements have an accuracy between 10 and 20% (Herschy, 1998).

The development of the rating curve averages out some of the error associated with the discharge measurements, however interpolation from the rating curve may also introduce error, especially if the rating curve is subject to change over time. The accuracy of estimates made from the rating curve diminishes with extrapolation beyond the highest and lowest measured discharges because the nature of the "true" rating beyond the measured values is not known. Additionally, hysteresis effects may not be adequately reflected in the rating. Dickerson (1967) suggests that accuracy in estimating future (uncalibrated) discharge values from a rating curve may range from +13% and -11% at the 80% confidence level, and from + 21% to -17% at the 95% confidence level.

Measurement Uncertainty Analysis

Models 1, 2, 3 and 4, and equation (2-1) enable exploration of the impact that potential uncertainty (error) in measurement of the dynamic variables W, Y and V would have on the accuracy of discharge predictions. To do this, (the measured variables were assumed to be error free which is not really the case), typical measurement accuracies were assigned to each variable, and then varied randomly assuming a normal distribution such that the mean measurement uncertainty for the entire data base is zero and 95% of the uncertainties are within the assigned accuracy. The modified data were then used to re-estimate the discharge in the validation data base and then these values were compared via the relative residual to the estimates that assumed no uncertainity. A maximum and minimum measurement accuracy is assumed for each dynamic variable.

For W, the minimum assumed measurement uncertainty is 1 m and the maximum is 10 m, which would be consistent with the resolution of many of the current SAR and visible spectrum sensors (Table 2.2). Although, accuracy in width (surface area) measurement greater than 10 m may be routinely possible over longer reach lengths and by using complimentary observation bands, the range selected for the error analysis is not considered to be unreasonable for the purpose of this analysis. The minimum assumed measurement uncertainity in water-surface elevation (as a proxy for Y) is 0.1 m and the maximum is 0.5 m, consistent with the range associated with current satellite altimeters (Birkett, 1998, Birkett et al., 2002). The minimum measurement uncertainty in V is assumed to be 0.1 m/s, which is the low end of the anticipated accuracy of a surface velocity measurement (Emmitt, personal communication, 2001), and the maximum was arbitrarily chosen to be 0.5 m/s (since the measurement of surface velocity from satellites has not been tested).

This analysis does not consider the potential uncertainity in estimating channel slope or the other characteristic channel values W_m and Y_m . These variables could be determined by a number of methods, including: 1) estimation from topographic mapping and geomorphologic considerations; 2) measurement from repeated satellite observations; and 3) measurement via field surveys. The magnitude of uncertainty associated with determining the channel characteristics will depend in large part on the accuracy of available topographic information and availability of channel survey data. The analysis also does not consider the uncertainity associated with estimating the average velocity from the surface velocity measurements or the uncertainty associated with converting stage to average depth. However, Costa et al. (2000) has shown that the surface velocity can be used to estimate the mean velocity in a single cross-section with good overall results by using a simple correction factor of 0.85 (Rantz et al., 1982).

The assumed measurement uncertainties are distributed with a mean of zero such that the mean value of the prediction residuals would not change. Because of this, the measure used to evaluate the effect of the measurement uncertainty on the predictions is the standard deviation of the relative residuals. The standard deviation of the relative residuals as a function of discharge category for the maximum assumed uncertainty (error), the minimum assumed uncertainty, and the case with no uncertainity are shown on Figure 2.4. The least variability is associated with using equation (2-1) because there is no associated statistical error. All of the plots in Figure 2.4 show that the impact of maximum measurement uncertainty on prediction variability, relative to the no uncertainty for discharge above 10 m³/s is greatest for equation (2-1) and Models 2 and 3. This result shows the effect of compounding errors in the case of equation (2-1), which includes uncertainty in all three dynamic variables, and indicates that uncertainty in V has a larger impact on prediction variability than does uncertainty in Y (comparing Model 1 and 2).

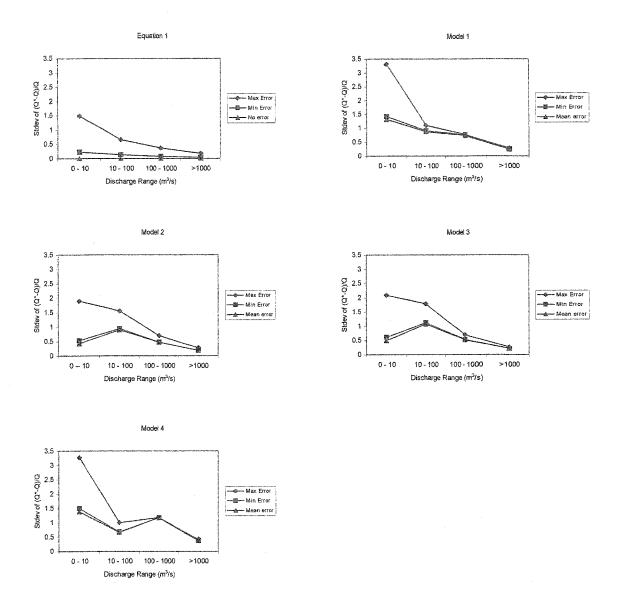


Figure 2.4 – Variation of the standard deviation of the relative residuals assuming a high (maximum) and low (minimum) measurement error in the dynamic variables as compared to no (mean) assumed measurement error. The dynamic variables are W, Y, and V. Ninety five percent of the assumed maximum errors are within +/-10 m for W, +/-0.5 m for Y, and +/-0.5 m/s for V. Ninety five percent of the minimum errors are within +/-1 m for W, +/-0.1 m for Y, and +/-0.1 m/s for V.

The impact of minimum uncertainty is not large within any discharge category, although as in the maximum uncertainty case it is most pronounced for discharge below 10 m³/s. However, if the minimum measurement uncertainty is achieved for all dynamic variables,

predicting discharge with equation (2-1) would result in a standard deviation in the relative residual (percent error) of less than 25% for discharges less than 10 m³/s, less than 15% for discharge in the range 10 - 100 m³/s and less than 10% for discharge greater than 100 m³/s. The impact of minimum measurement uncertainty using Models 1, 2, 3 or 4 is less than 15% for discharge less than 10 m³/s, and less than 10% for all other discharge categories. The plots in Figure 2.4 show that if the minimum measurement uncertainty can be achieved, uncertainty in the estimated discharge using the statistically based models is well below the uncertainty associated with the model itself (no error case).

As suggested by comparing the plots for Model 1 and Models 2 and 3 in Figure 2.4, there appears to be a different error response between Y and V. The differences in measurement uncertainty impact associated with the three dynamic variables were evaluated by introducing error into one variable at a time, and then comparing the standard devaition of the relative residuals. The results of this analysis are shown in Figure 2.5 for equation (2-1), and Models 1 and 3. The plot for equation (2-1) shows that error in V has greater impact on the discharge estimate than does error in Y, and that error in W has the least impact. Comparing Models 1 and 3 shows that error in Y has the largest impact relative to W and V at low discharge (less than 10 m³/s), and that error in V has a greater impact than error in Y for discharge greater than 10 m³/s.

Discussion

The advantage of a satellite-based river-discharge-monitoring system is that it has the potential to fill in gaps where there is little or no information and obtain data over large areas simultaneously. Another advantage that satellite (or aerial) based measurement of hydraulic variables (particularly width) could provide is the ability to observe variation over a reach, thus enabling a reach-averaged value to be derived and minimizing the local variability that is specific

to single cross sections. Development of a general method to estimate river discharge using riverchannel hydraulic information observed from existing space or aerial platforms can be accomplished with statistical relationships developed from river data bases. If the water surface velocity of a river can be observed with Doppler lidar and used to estimate the average crosssectional velocity and if the water-surface elevation can be used to estimate the average depth, all elements of equation (2-1) can be obtained remotely and the discharge in the river can be directly calculated.

The use of equation (2-1) is the preferred method to estimate discharge because it does not rely on a statistical derivation, shows the least overall prediction variance, and is applicable to any river under any flow conditions. However, it is likely that not all elements of equation (2-1) can be observed at the same time with confidence, thus in these situation statistically based models such as described by Models 1, 2, 3 and 4 can be used with reasonable accuracy, averaging +/- 20% or less, with accuracy within approximately +/- 50% 2/3 of the time. This level of accuracy compares favorably with estimates derived from extrapolation of ground-based ratings and slope-area measurements of discharge. Measurement error analysis indicates that with anticipated maximum uncertainty in the values of the observed variables, the variability of discharge estimates is increased substantially for discharges less than 100 m³/s, however assuming, a reasonable minimum measurement uncertainty (0.1 m accuracy in depth, 1 m accuracy in width and 0.1 m/s) prediction error variability is only slightly increased over the noerror case.

Models that use width and surface-velocity only to estimate discharge (Model 3) can be used in situations where slope cannot be measured, or where anthropogenic control of slope



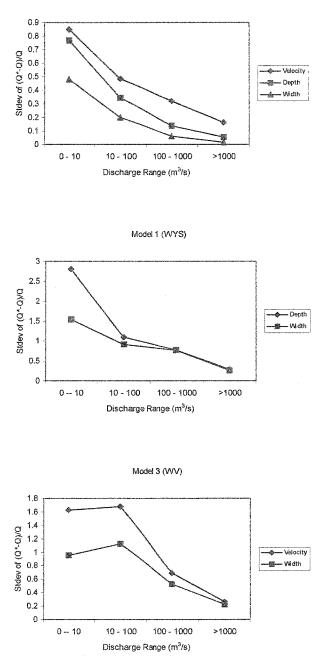


Figure 2.5 – Variation in the standard deviation of the realtive residuals for prediction methods with more than one dynamic variable assuming error in only one variable at a time, showing the relative impact that error in the different dynamic variables has on prediction variability.

violates the hydraulic assumptions inherent in Models 1 and 2, assuming surface velocity can be effectively measured. However, width-velocity models appear to have a bias trend across a wide range of discharge.

The predictive models described above are applicable to within-bank discharge only, because these models did not include over-bank flow in the data base used to develop and evaluate them. However, estimating over-bank discharge would require the same information, i.e. the width of flow, the average depth of flow and the average velocity of flow. Alsdorf et.al. (2000) has shown the feasibility of using interferometric SAR to map the surface relief of an inundated region of the Amazon, thus demonstrating that mapping flow paths within a flooded area is possible. With this information, the discharge within the flooded area could be estimated and resolved in the downstream direction using floodplain topography and water-surface elevation to estimate the flow depths across the inundated area. As shown by Brakenridge et al. (1998), Bates and DeRoo (2000), and Horritt (2000), this information could also be used in conjunction with a hydraulic model to estimate the discharge within a flooded area (including depth and areal extent) which were then used to track the flood wave and estimate flood discharge using the HEC-2 river hydraulic model.

The successful use of equation (2-1) and Models 2 and 3 will depend on the ability to measure surface velocity from space. To this end, development and verification of this technology will greatly enhance the potential ability to measure river discharge from space. Additionally, use of equation (2-1) and Models 1, 2, 3 and 4 all depend on the ability to translate surface measurements of stage and/or velocity into average values for the channel section under observation. Thus, techniques to estimate the average water depth in a channel section based on

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observation of water-surface elevation and techniques to estimate the average velocity in channel section based on measurements of surface velocity need to be developed and verified. Another issue of concern is that currently deployed altimeters cannot accurrately obtain water-surface elevations on rivers less than several hundred meters wide. However, there is an indication that these same altimeters can observe much smaller rivers with similar accuracy by effecting a change in the on-board signal processing (personal communication, Ernesto Rodriguez). Also, laser altimeters may provide much greater accuracy with reduced observation size limitations relative to radar altimeters. The potential improvements in river-stage measurement indicated by these developments need to be evaluated.

CHAPTER III

DEVELOPMENT OF GENERALLY APPLICABLE EQUATIONS FOR ESTIMATING RIVER DISCHARGE

Because width is the most readily and reliably measured hydraulic element of the channel from remote sources, relationships that use width with either of the other two elements of the continuity equation (depth and velocity) along with characteristic channel variables such as maximum width and channel slope, would provide the most flexibility in measuring discharge from remote platforms or sources. Estimating discharge from relationships that use width as the only variable do not provide sufficient information to characterize the range of river discharge variability with reasonable accuracy (e.g. a mean prediction accuracy within 20% or less of the expected value with, 67% of the predictions within 50% of the expected value), and consequently do not yield generally applicable models.

Additionally, models based on width only cannot be derived from hydraulic principles without over-simplification of in-channel hydraulic relationships, and are therefore limited to uniquely derived statistical relationships for a given reach. Thus, there is advantage to developing discharge-estimating equations that rely on measured width and geomorphic characteristics that can be readily observed and one of the other two dynamic variables of continuity, depth or velocity. Developing generally applicable models with the fewest possible independently measured variables that can provide reasonable estimation accuracy will minimize compounding error; and provide a way of estimating discharge when all of the elements of flow continuity cannot be observed or accurately estimated.

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Generally applicable statistical relationships developed from multiple-regression analysis, that use width, channel slope, and either mean depth or mean velocity to estimate river discharge, as described in Chapter 2, have shown that discharge can be estimated with reasonable accuracy. However, regression models do not allow for improvements in the estimates if there is better knowledge of the behavior of one river as opposed to another. The coefficients and exponents of the models are fixed by the errors and variability within the data set used to develop them, and even if the data set represented the entire population of flows, variability that is not explained by the regression cannot be reduced by inclusion of more specific knowledge that may be available for a specific river. For this reason, rationally based equations that are developed from physical principles would provide more general and adaptable models for estimating discharge in rivers. In addition, rationally based models can be calibrated to specific rivers where additional or better knowledge is available.

This chapter develops and evaluates the use of generally applicable river-discharge estimating equations that are based on width, channel slope and either mean depth or mean velocity. Both statistically and rationally derived equations are developed from a flow-measurement data base similar to that used in Chapter 2 and a synthetic discharge data base that is based on principles of river hydraulics. Comparison and applications of these relationships are discussed including their use with other types of hydrologic information.

Hydraulic Data

A large discharge-measurement data base was developed in order to derive, calibrate and compare statistically based discharge-prediction models with similar models developed from physical principles. The data base includes 1,037 flow measurements from 103 rivers in the United States and New Zealand. At each river station, from five to twenty in-bank discharge

measurements obtained for as wide a range of flow as possible, were incorporated into the data base. The data base includes a measured width and/or cross-sectional area, mean depth and/or hydraulic radius, a mean velocity for the measurement section or reach, and an average or topographic channel slope for the reach (see Chapter 2). In addition, the maximum depth and width for the set of measurements at each station were included as a separate channel-shape variable. Approximately half of the measurements consisted of values averaged for a given reach, and the remainder of the data were obtained from single measurement cross-sections.

The data were obtained from Barnes (1967), Hicks and Mason (1991), Coon (1998) and from the U.S. Geological Survey's on-line NWIS data base (USGS, 2001), and all represent single-thread channels. The data include rivers that do not exhibit any control on the slope (no back-water effects), and no large expansion or contraction of the flow within the reach where the data were collected (thus, rivers that were contracted by a bridge or natural feature such as a canyon or narrows were not included in this data). These data are referred to as the channel-control data base (Appendix 1). The channel-selection criteria were implemented so that the hydraulic variables could all be considered adjusted to the channel slope. The channel characteristics of each river in the data base were evaluated based on information available from the data sources, or from inspection of topographic maps of the channel at each station. For comparative purposes, the channel-control data base was randomly divided into a calibration data set (N = 680) and a validation data set (N = 357). The range of data in each sub-set is shown on Table 3.1.

Approximately 90% of the data in the channel control data base is the same as that used for the multiple regression analysis presented in Chapter 2. Eight large rivers, including the

Amazon, were excluded from this data base because the data were judged to be affected by channel constrictions (both natural and anthropogenic) or other controls on the water-surface slope. Nine additional rivers with no slope control were substituted for those that were excluded in order to maintain a similar size data set.

As discussed above, there are advantages to developing hydraulic models from physical principles rather than basing the relationships solely on the statistics of particular data sets. For this reason, a theoretically derived river-channel and discharge data set was generated from which various hydraulic relationships were statistically extracted and analyzed. The synthesized data set was developed from the Prandtl-von Karman universal velocity distribution law assuming a uniform channel with a parabolic channel cross-section shape. The following describes the steps taken in developing the data base.

The Prandtl von-Karman universal velocity distribution law states that the velocity (v) in a vertical profile varies with distance from the bottom (y) as a log-function of the vertical distance above an assumed roughness height. This relationship is given by:

$$v = 2.5V* \ln(y/k)$$
 (3-1)

where V* is the shear velocity $(V^{*=}(gYS)^{1/2}, g$ is the acceleration due to gravity, Y and S are mean depth and bed slope respectively) and k is a constant that is proportional to the surface roughness of the streambed, and is equal to 0.033 times the roughness height (k_s) (Chow, 1959). The roughness height is considered to be the effective height of surface irregularities that intrude

Table 3.1	Range of	Range of Hydraulic Parameters in Data Sets					
Parameter	Symbol	<u>Units</u>	<u>Mean</u>	<u>Stdev</u>	Coeff. Var	<u>. Maximum</u>	<u>Minimum</u>
Calibration Data N = 680							
Discharge Top Width	Q W	m ³ /s m	860 128	2434 159	2.83 1.24	27576 1009	0.01 2.9
Average Depth (Hyd. Radius) Average Velocity Water Surface Slope (average	Y V e) S	m m/s 1	2.38 1.15 0.0029	2.24 0.62 0.0056	0.94 0.54 1.93	12.39 5.1 0.04	0.1 0.02 0.000043
Validation Data N = 357	, -						
Discharge Top Width Average Depth (Hyd. Radius) Average Velocity Water Surface Slope (average	Q W Y V e) S	m ³ /s m m/s 1	717 126 2.33 1.11 0.0021	1960 146 2 0.59 0.0042	2.73 1.16 0.86 0.53 2.00	17837 765 12.7 3.53 0.04	0.02 3.1 0.18 0.02 0.000043
Synthetic Data Set N = 380							
Discharge Top Width Average Depth (Hyd. Radius) Average Velocity Water Surface Slope (average	Q W Y V S) S	m ³ /s m m/s 1	4985 337 3.43 1.4 0.0012	405 3.59 0.63	1.20 1.05 0.45	98233 2000 21.78 3 0.01	30 0.1 0.15

beyond the laminar sub-layer for hydraulically rough flow conditions (which is the case in most natural rivers)(Chow, 1959). The relation between k (used in the following theoretical development) and the roughness height (k_s) was developed from hydraulic experiments performed by J. Nikuradse in 1933 (Chow, 1959).

A general discharge equation can be derived in terms of velocity as:

$$Q = \iint v \, dy \, dx \tag{3-2}$$

where y is the mean distance above the bottom (depth) and x is the top width at y. Integrating v in equation (3-1) with respect to y

$$q = 2.5V^* \int \ln(y/k) \, dy$$
 (3-3)

gives the unit discharge in the vertical (the flow per unit distance along the cross section):

$$q = 2.5V^* y(\ln(y/k) - 1)$$
(3-4)

The unit discharge can now be integrated with respect to dx to obtain the total discharge in the cross-section assuming a regular geometric cross-sectional shape. To conduct the integration, a parabolic cross-section shape is assumed (Fekete, 2002; Chow, 1959). A parabolic shape is often used to represent self-formed river channel cross-sections (Chow, 1959), and in many cases are comparable to those obtained from assuming other regular geometric shapes commonly used such as a semi-ellipse, trapezoid, or higher order paraboloid. Additionally, hydraulically efficient stable channel cross-sections developed from theoretical considerations can be represented by cosine functions that are nearly equivalent to parabolic sections (Henderson, 1966; Ferguson, 1986). This indicates that the assumption of a parabolic shape as representing the "typical" self-formed channel is reasonable.

The integration of Equation (3-4) involves inverting the parabola to obtain the proper integration under the curve such that $y = y_m - c x^2$ where y_m is the maximum depth within the section bounded by x. The inversion is illustrated on Figure 3.1. The coefficient c is a geometric constant for the parabola defined at maximum width and depth, equal to $1/(W_m^2/Y_m)$. Substituting this into Equation (3-4) and setting $w_m/2 - \varepsilon = X$, where ε is an arbitrarily small number gives:

$$Q = 2.5V^* \int_0^x (y_m - c x^2) [\ln((y_m - c x^2)/k) - 1] dx$$
(3-5)

Integrating equation 3-5 gives (webMathematica, 2002):

$$Q/2 = 2.5V^{*}[-7y_{m}x/3 + 5cx^{3}/9 + (4y_{m}^{1.5}/3c^{0.5})(ArcTanh[c^{0.5}x/y_{m}^{0.5}]) + (y_{m}x - cx^{3})ln[(y_{m} - cx^{2})/k]$$
(3-6)

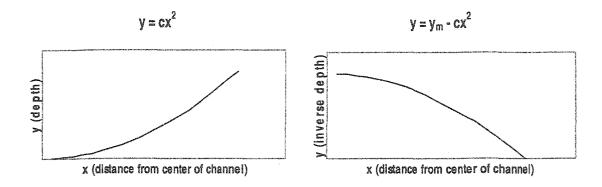


Figure 3.1 - Definition sketch: integration of the parabolic section

Numerous studies have shown that rivers exhibit general hydraulic relationships between depth, slope, width, velocity and resistance (Leopold et al, 1964; Henderson, 1966; Rosgen, 1996). In order to avoid inclusion of unrealistic channels in the synthesized data base, general rules for estimating the maximum depths and roughness heights were used. The rule for estimating the maximum depth was developed based on multiple-regression analysis of a bankfull hydraulic geometry data set compiled from various sources for 521 river reaches (Schumm, 1960; Barnes, 1967; Osterkamp et al., 1982; Church and Rood, 1983; and Dingman and Palaia, 1999) (Appendix 3). Based on the analysis of this data, it was found that maximum depth (Y_m) could be predicted from the maximum width (W_m) and the slope (S) according to the following relation:

$$Y_{m} = 0.08 W_{m}^{0.39} / S^{0.24}$$
(3-7)

In equation (3-7) all units are in meters. The width and slope in equation (3-7) explains approximately 73 % of the variation in maximum depth ($r^2 = 0.73$).

The roughness height was estimated directly from theoretical considerations based on an initial assumption that the Manning resistance coefficient (n) is a function of the slope (Bray, 1979) as follows:

$$n = 0.1S^{0.18} \tag{3-8}$$

Chow (1959) presents a dimensional relationship between the Chezy C and Manning n based on the hydraulic radius (in feet), and a theoretical relationship between the Chezy C and roughness height (in feet) as follows

$$C = 1.49R^{0.17}/n$$
(3-9)
C = 32.6 log(12.2R/k_s) (3-10)

Equations (3-8), (3-9) and (3-10) were used to compute the roughness height (k_s) for a given slope assuming that $R = Y_m$. The value of k in equation (3-6) is then computed as 0.033 times the roughness height. Thus, the channel dimensions and hydraulic characteristics of the synthesized data base are derived from maximum width and channel slope, and the assumption of a parabolic channel shape.

The range of slopes and widths used to develop the synthetic data base were within the same range of values included in the flow-measurement data base (Table 3.1). It is recognized that small streams may take on a large range of slopes, but typically larger rivers will only exhibit relatively flat slopes. However, no such behavioral rule between width and slope was invoked to generate the synthetic data. Instead, the slope and width were treated as independent variables, and the range of values were selected to be comparable to actual rivers.

The derived synthetic data base consists of 380 flows with associated values for width, mean depth, mean velocity and slope (a constant value for each synthetic river channel) in units of meters and seconds (Appendix 4). Comparison of the synthetic and measurement data bases was accomplished by analyzing the behavior of the dimensionless Froude number. In both the synthetic and flow measurement data bases, the Froude number was found to be predictable from a dimensionless velocity head index given by $V^2/(2gW)$. The velocity head index (VHI) is used here because it does not include a depth term, and therefore would be more useful in a predictive capacity (because depth is not readily available from remote data). The Froude-number-VHI relationship derived from the synthetic data is:

$$F = 2.20 [V^2/(2gW)]^{0.27}$$
(3-11)

The same relationship derived from the measurement data is:

$$\mathbf{F} = 2.32 \left[\mathbf{V}^2 / (2gW) \right]^{0.31} \tag{3-12}$$

Although not equivalent at the 95% confidence level, the similarity of these equations indicates the general comparability of the two data sets.

The Froude relationship for both the measurement data and the synthetic data are shown on Figures 3.2a and 3.2b. Further analysis of the relationship between the Froude number and

VHI for the synthetic data revealed that the variability in the general relation given by equation (3-11) can be reduced by using the maximum width and the slope to predict the coefficient and exponent of the equation. Multiple regression analysis of (W_m) and (S) on the coefficient (c) and exponent (m) of equation (3-11) gives the following:

$$c = 23.7 W_m^{0.125} S^{0.449}$$
(3-13)

and

$$m = 0.881 W_m^{-0.046} S^{0.148}$$
(3-14)

Figures 3.2c and 3.2d shows the predictive characteristics of the general relation between Froude number and VHI derived from the synthetic data base applied to both the synthetic and measurement data. Figures 3.2e and 3.2f show the improvement in predictability by using the relationships given by equation (3-13) and (3-14). The improvement in prediction for the synthetic data can be readily observed. The improvement in prediction for the measurement data was measured by computing the mean and standard deviation of the relative residual (predicted minus observed divided by the observed) and the log-residual (log of the predicted minus log of the observed) of the estimate. The mean relative residual and log-residual associated with Figure 3.2d is 36% and 29% respectively, and the mean relative residual and log-residual associated with Figure 3.2f is 23% and 17% respectively, indicating that knowledge of W_m and S can substantially reduce Froude-number estimating errors. The standard deviation of the errors were also reduced. Thus, the hydraulic characteristics of the synthetic data can be used to derive relationships that help explain variability within the measurement data, indicating that the theoretically derived data is a useful representation of real-world rivers.

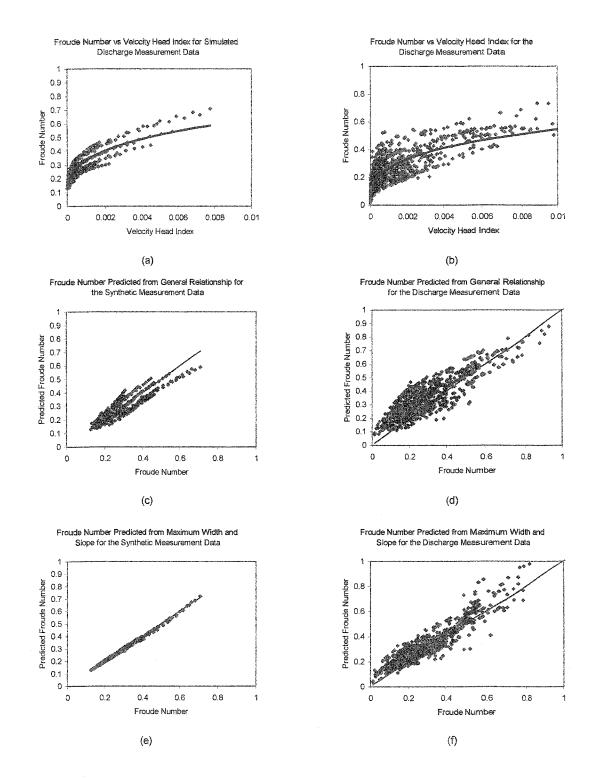


Figure 3.2 – Froude number plotted against the VHI for the synthetic (a) and measurement (b) data; actual Froude number plotted against the Froude number predicted from the general relation for the synthetic (c) and measurement (d) data; improvement in prediction using the width-slope correction for the synthetic (e) and measurement (f) data.

Statistically Based Discharge Estimating Models

For comparison, a set of regression models parallel to those proposed in Chapter 2 were developed from regression analysis of the channel-control calibration data. These models are:

		<u>r²</u>	std. error	
Model 1:	$Q = 4.24 W^{1.10} Y^{1.63} S^{0.33}$	0.97	0.19	(3-15)
Model 2:	$Q = 0.08 W^{1.16} V^{1.65} S^{-0.33}$	0.97	0.19	(3-16)
Model 3:	$Q = 0.23 W^{1.48} V^{1.45}$	0.95	0.24	(3-17)
Model 4:	$Q = 4.74 W_m^{1.04} Y_m^{-0.58} S^{0.29} Y^{2.11}$	0.94	0.27	(3-18)

The regression results differ from those presented in Chapter 2 because a somewhat different data base was used to develop them, although approximately 90% of the data in the channel-control data set are the same. Thus, even with a large amount of data in common, the statistically based models would, in general, provide different predictive results, indicating one a key limitation of statistically derived models. Comparing these models with those derived in Chapter 2 (Table 2.6), it is found that the magnitude of the slope exponent for Models 1 and 2 were the same at the 95% confidence level, whereas the values for the width, depth and velocity exponents were not always the same between these models. This suggests that slope is an effective discriminating variable even where the interaction of width and velocity may vary. Model 3 is similar for both data bases at the 95% confidence level.

The exponents of Model 4, with the exception of slope and depth, are also different, however an interesting aspect of this model is that the exponents on the maximum width and maximum depth are near the expected values (1.0 for W_m and -0.5 for Y_m) if the "typical"

channel shape were a parabola (2nd order paraboloid). This suggests that self-formed single thread channels tend towards parabolic shapes, further verifying the assumption underlying the development of the synthetic data base.

The exponents associated with Model 1 suggest a similarity to the Manning equation which is given as:

$$Q = (u/n)AR^{2/3}S^{1/2}$$
(3-19)

where u is a proportionality constant and n is a resistance coefficient. The exponent on the depth term is near the expected value of 1.67 and the exponent on the width is near the expected value of 1. However, the exponent on slope is closer to 0.33 rather than 0.5, as formulated by Manning (1895).

Comparable regression models were also developed for the Prandtl-von Karman synthetic data base. Model 4 was not developed from these data because the maximum depth was derived from the maximum width and the slope (and thus is perfectly correlated with maximum width and slope). The resultant regression models are:

		<u>r²</u>	std. error	
Model 1:	$Q = 8.42 W^{0.98} Y^{1.74} S^{0.31}$	1.00	0.04	(3-20)
Model 2:	$Q = 0.06 W^{1.07} V^{2.25} S^{-0.38}$	1.00	0.06	(3-21)
Model 3:	$Q = 0.12 W^{1.53} V^{1.85}$	0.99	0.13	(3-22)

Model 1 derived from the synthetic data shows nearly the same exponents as those derived from the measured data bases (although different at the 95% confidence level with the exception of the slope exponent), and are similar to the Manning equation except for the exponent on the slope,

which is also nearer to the cubed root rather than the square root. The exponents on Model 2 and 3 are different than those for the regression models derived from the measurement data.

The regression models from each of the data bases suggests that a general form of Model 1 could be represented by the Manning equation with an exponent of 0.33 (cubed root) on slope rather than the square root. However, the regression models do not immediately suggest any general form for Model 2 or 3. The consistent behavior of the regression statistics with regard to Model 1 suggests a robustness with regard to a general form, and its similarity to the Manning equation is encouraging. Additionally, the result that the slope exponent is always nearer 0.33 as opposed to 0.5 suggests that there is an underlying principle of natural rivers that relates the resistance to $S^{0.17}$, thus resulting in a slope exponent of 0.33 (as indicated by equation 3-8).

General Discharge Estimating Equations

Open-channel flows are often modeled as one-dimensional gradually varied steady or unsteady flows. Such flows satisfy three fundamental relations including: 1) continuity, requiring the conservation of mass; 2) the energy equation, characterizing the apportionment of mechanical energy and its spatial and temporal rates of change; and 3) a constitutive relation, characterizing the relation between energy gradient and flow rate. The constitutive relation is generally described as,

$$V = Kg^{1/2}R^{p}S^{q}$$
(3-23)

where K is a channel conductance and S in the general sense can be taken as the friction slope (slope of the energy grade line). Specification of the constitutive relation is not straightforward because there is uncertainty about the values of p and q (Manning, 1889; Golubstov, 1969) and the way in which the velocity conductivity coefficient, K, varies with flow and boundary characteristics.

The most widely used constitutive relation is the Manning equation. However, studies that have established a sound theoretical basis for this relation or have unequivocally demonstrated that it governs all uniform flows are not evident in the literature. A number of studies before and after publication of the papers on which wide acceptance of the Manning equation are based (Manning 1889, 1895) have discussed the appropriate values of p and q and the question of whether and how the conductance coefficient varies with flow and channel characteristics. As is well known, Manning himself felt that the constitutive equation should be dimensionally correct and was uncomfortable with the form of the equation that came to bear his name (Manning, 1895).

One major problem with the Manning equation – and of many other proposed forms of the constitutive relation - is that there is no universally accepted way of determining the appropriate value of the conductance/resistance parameter from measurable channel characteristics for *a priori* or *a posteriori* applications. In addition, the Manning equation violates at least two of the principles that should be satisfied by a constitutive relation (Bear 1972): (1) consistency with principles of momentum balance and (2) dimensional homogeneity. The Manning equation is an empirical modification of the Chezy equation,

$$V = C \cdot g^{1/2} \cdot R^{1/2} \cdot S_e^{1/2}$$
(3-24)

which can be derived from force-balance relations and is dimensionally homogeneous. However, the Chezy equation is based on the dimensionally-motivated assumption that resistance is proportional to V^2 . As Leopold et al. (1960) pointed out, that assumption may only be true if resistance associated with the flow boundary does not change with V, in other words resistance is constant for all velocities. This condition is generally true for pipe flow but not for open channels where the boundary changes substantially with discharge.

Manning (1889) himself cited empirical studies that showed various values for p and q, and subsequent empirical and even theoretical (Leopold et al. 1960) and quasi-theoretical (Henderson 1966) studies have found wide variation in both p and q. Several studies, including Golubtsov (1969), Riggs (1976), Jarrett (1984), and Dingman and Sharma (1997), have not only used statistical analysis to reveal different apparent values of p and q, but also to suggest that a very wide range of flows can be successfully modeled using a universal value for the velocity (conductance) coefficient (K). This latter point is especially important, because confirmation of this finding would free the modeler from the inherently subjective and highly uncertain (HEC, 1986) process of estimating the resistance. Lane's stable-channel analysis (Henderson, 1966) also suggests that it may be possible to model open-channel flows using a constant conductance coefficient for all channels, at least to the accuracy obtainable by the usual subjective methods for estimating reach-specific resistance/conductance. Thus, it is of interest to compare the variability of K over all flows in the data bases using different assumptions for the values of p and q.

Four discharge-estimating models with exponents selected *a priori* based on the Chezy, Manning and regression equations were used to evaluate the variability of K as a function of p and q. These are:

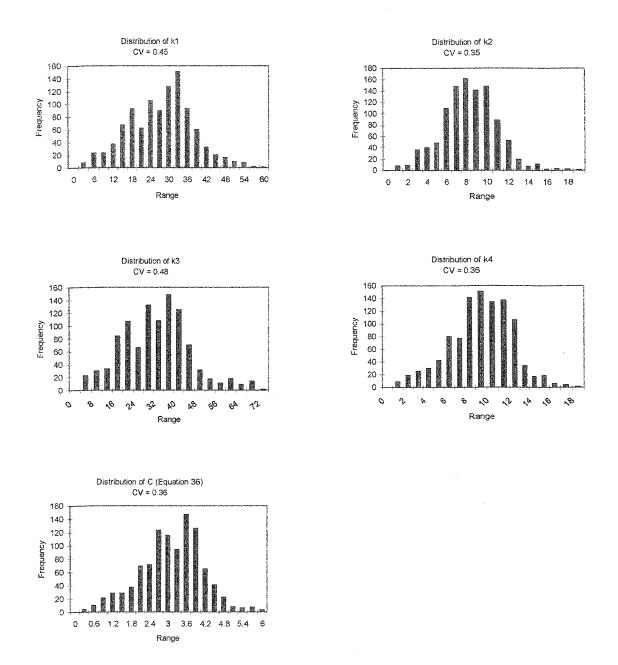
$Q = k_1 W Y^{1.67} S^{0.5}$	(3-25)
$Q = k_2 W Y^{1.67} S^{0.33}$	(3-26)
$Q = k_3 W Y^{1.5} S^{0.5}$	(3-27)
$Q = k_4 W Y^{1.5} S^{0.33}$	(3-28)

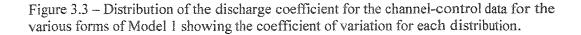
The discharge coefficients (k_1 through k_4) were determined for each flow measurement in the channel-control data base and their distributions plotted as histograms in Figure 3.3. The coefficient of variation for the distributions is also indicated on Figure 3.3. Comparison of the historgrams indicates that the discharge coefficient is bi-modally distributed when a slope exponent of 0.5 is used (k_1 and k_3), and there is significantly less overall variability in the conductance coefficient when an exponent of 0.33 is used (k_2 and k_4). This indicates that there would be less estimation error and greater accuracy when using constituitve equations that assume a slope exponent of 0.33 for natural rivers.

The improved predictive qualities of the models when using a slope exponent of 0.33 can be explained, in part, by assuming that the principal source of resistance is the boundary roughness, and that the boundary roughness is directly related to a characteristic stable grain size. The stable grain diameter is proportional to the maximum channel hydraulic radius (or depth) times the slope (Henderson, 1966), such that $D = \phi Y_m S$ with D equal to the stable grain diameter and ϕ is a coefficient that accounts for the Shields entrainment function and the specific gravity of the sediment (solid:fluid density ratio). Given that bed shear stress is related to the size of the bed material, the stable-bed resistance coefficient would also be related to grain size (Chow, 1959; Henderson, 1966). Resistance is also known to be a function of the depth of flow (Chow, 1959), consistent with the concept of relative roughness (Engelund, 1966;Limerinos, 1970; Hey, 1979; Arcement et al., 1989). Thus, an expression for the stable-bed resistance should include the maximum depth and slope (Y_mS) to account for the resistance associated with the size of the stable-bed material, and the flow depth (Y) to account for the relative roughness. With these assumptions a dimensionally homogeneous constitutive equation based on the Chezy equation would take the form:

$$Q = C^* g^{0.5} W Y^{1.5} S^{0.5} / (Y_m S / Y)^f$$
⁶⁰
⁶⁰

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where f is an exponent relating the stable grain size to resistance, and C^* is a constant of proportionality that may vary with flow conditions. The value of C^* , determined by minimizing the log-residual of error using the channel-control data, is estimated to be 2.74. This equation

maintains dimensionality by including the ratio Y_m/Y , which also accounts for the effect of relative roughness. Studies by Lacey (Bray 1979) have indicated that bankfull (or regime) resistance, as expressed in the Manning equation, in natural gravel bed channels is a function of slope to the $1/6^{th}$ power (i.e. resistance is proportional to S^{0.17}). This general relationship was further substantiated by Bray (1979). Accepting this relation (f = 0.17) would result in a slope exponent of 0.33 and a depth exponent of 1.67 for Model 1, which confirms the results from the regression analyses. The distribution of C^{*} for this model has a smaller range than the k values for the comparable models as shown on Figure 3.3.

Because the variables used in equation (3-29) are rationally developed and provide a more complete representation of the geometric contributions to resistance, this equation is considered to be a more physically complete formulation of Model 1 compared to equations (3-25) through (3-28). Deriving the form of equation (3-29) from regression analysis of the channel control data (N=1037) yields the following equation:

 $Q = 0.84g^{0.5}W^{1.25}Y^{1.70}S^{0.27}Y_{m}^{-0.43}$ $r^{2} = 0.97$ std error = 0.18 (3-30)

An interesting aspect of this equation is that it is very nearly dimensionally homogeneous. This suggests that the correct variables are included in the model, and thus also are included in equation (3-29). However, the magnitude of the exponents are significantly different than those proposed for equation (3-29), indicating that equation (3-29) is not a completely satisfactory physical representation of Model 1. Equation (3-30) suggests that width and other factors related to slope and maximum depth are important in defining the resistance.

Equation (3-29), if applied to the bankfull flow condition, would reduce to a form similar to the Chezy equation, except that the exponent on the slope is 0.33 rather than 0.5, because $Y = Y_m$. Leopold et al. (1960) suggested that rivers in regime tend towards a constant bankfull

Froude number. Figure 3.4 shows the Froude number plotted against discharge for 22 rivers in the United States over a wide range of flows (several orders of magnitude) obtained from the USGS NWIS data base. The plots show two distinctive patterns. One is a logarithmic increase in Froude number which converges to a constant value at high discharge, and the other is a random scatter at low discharge which also converges to a constant at high discharge. In either case, there appears to be a tendency for the Froude number to reach a constant value as discharge increases toward the bankfull or regime flow.

The rivers shown on Figure 3.4 represent channels that are unrestricted (as determined by inspection of topographic maps of the river stations) and therefore the Froude number would not reflect backwater or accelerating flow conditions. It is not known (because it was not recorded in the data base) whether any of the discharges shown on the plots are greater than bankfull. However, it can be surmised that if the Froude number reaches a constant value at high discharge, the asymptote would occur at or near the bankfull discharge, and may persist in overbank flow conditions assuming that the majority of flow even in floods remains in the channel (which may be the case for smaller overbank flood events).

The asymptotic Froude number for each river shown on Figure 3.4 was estimated by inspection and tabulated in Table 3.2 along with the channel slope measured from topographic maps. These data are plotted on Figure 3.5 and show that the asymptotic Froude number is a function of the channel slope and can be fit to the following equation, which is plotted as the trend line given by:

$$\mathbf{F} = 3.5 \mathbf{S}^{0.33} \tag{3-31}$$

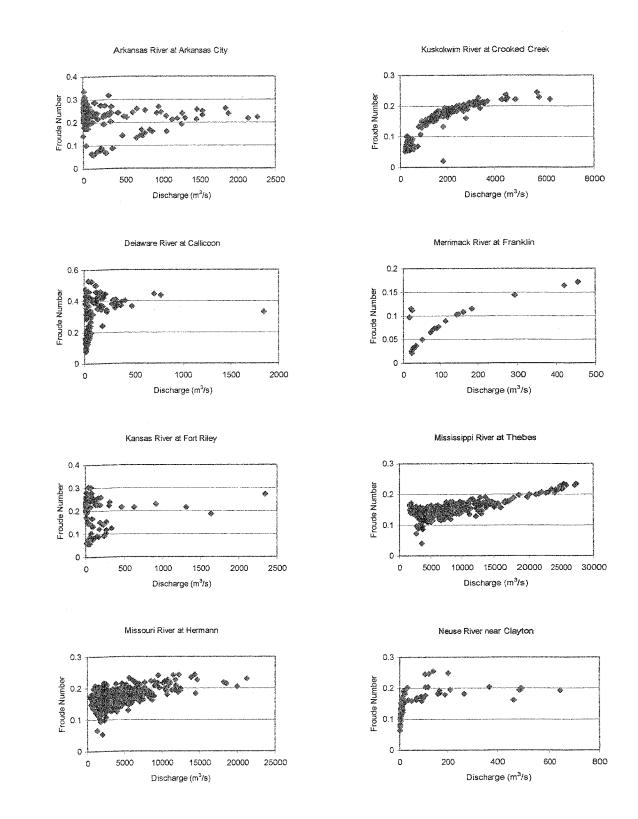
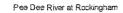
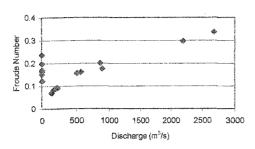
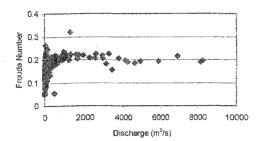


Figure 3.4 – Froude number as a function of discharge at 22 gaging stations.

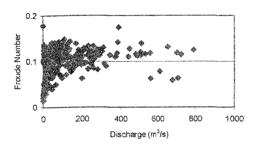


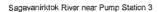


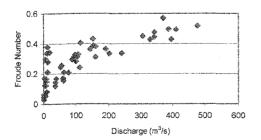


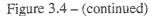


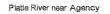
Red River of the North at Fargo

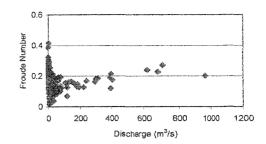




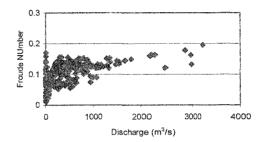




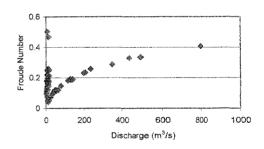




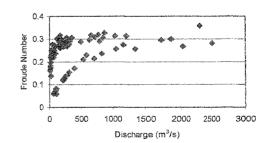




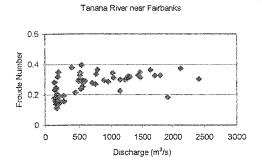
Saco River near Conway

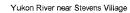


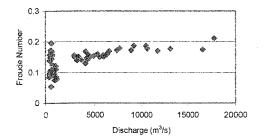












Chena River at Two Rivers

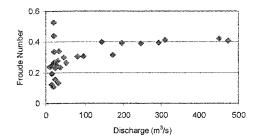
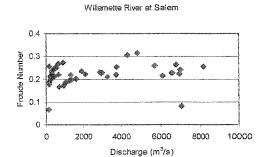
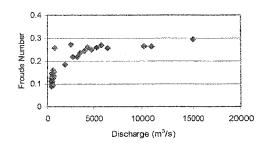


Figure 3.4 – (continued).



Yukon River near Eagle



Kobuk River at Kiana

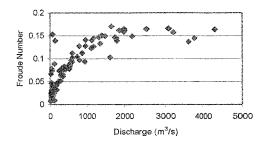


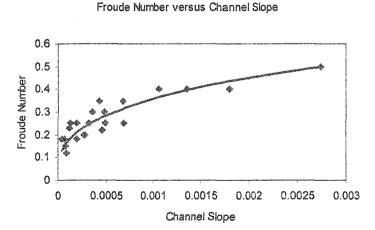
TABLE 3.2 - High Flow Froude Numbers and Channel Slopes

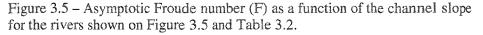
River Gaging Station	Channel Slope ¹	Froude Number ²
Arkansas River at Arkansas City, Kansas	0.000685	0.25
Delaware River at Callicoon, New York	0.00107	0.4
Kansas River at Fort Riley, Kansas	0.00049	0.25
Kuskokwim River at Crooked Creek, Alaska	0.000198	0.25
Mississippi River at Thebes, Illinois	0.000137	0.25
Missouri River at Hermann, Missouri	0.00013	0.23
Platte River near Agency Missouri	0.00046	0.22
Red River of the North at Fargo, North Dakota	0.00009	0.12
Willamette River at Salem, Oregon	0.00032	
Yukon River at Stevens Village, Alaska	0.00068	0.18
Yukon River at Eagle, Alaska	0.00036	
Saco River at Conway, New Hampshire	0.0018	
Chena River at Two Rivers, Alaska	0.00136	
Kobuk River at Kiana, Alaska	0.0008	
Sagavanirktok River near Pump Station 3, Alaska	0.00274	0.5
Merrimack River at Franklin, New Hampshire	0.0002	
Neuse River at Clayton, North Carolina	0.00028	
Pee Dee River at Rockingham, North Carolina	0.00068	
Potomac River at Point of Rocks, Maryland	0.00027	
Susquehanna River at Waverley, New York	0.00048	
Tanana River near Fairbanks, Alaska	0.00043	
Red River of the North at Grand Forks, North Dakota	0.000043	0.18

Notes:

1 - Channel slope measured from topographic maps

2 - Froude number based on inspection of Figure 3.4





Substituting the definition for the Froude number, $F = V/(gY)^{0.5}$, into equation (3-31), rearranging and multiplying by the cross-sectional area yields the following equation for bankfull discharge:

$$O = 3.5g^{0.5}W_m Y_m^{1.5}S^{0.33}$$
(3-32)

which verifies the form of equation (3-29) if Y_m is substituted for Y. The different discharge coefficient in equation (3-29) compared to equation (3-31) (i.e. 3.5 versus 2.74 respectively) is probably due in part to the small data set used to develop equation (3-31), but may also indicate that the discharge coefficient varies with flow conditions.

Thus, equation (3-29) is considered to be a rational form of Model 1. However, a model that requires an estimate of both the depth and maximum depth, as would be needed for equation (3-29), increases the potential for compounding errors because depth cannot be directly measured remotely and would need to be estimated. Therefore, a more practical general formulation for Model 1 would be that given by equation (3-26). This equation appears to have comparable (possibly somewhat better) predictive characteristics compared to equation (3-29) after calibration of the discharge coefficient (Figure 3.2; also see Table 3.3 following this section).

Model 1, as described by equation (3-26) can be used to develop a general form of Model 2, (which does not require an estimate of the depth) by re-arranging and solving for depth, and then equating it to continuity, Q = WYV. The resulting form of Model 2 is:

$$Q = kWV^{2.5}S^{-0.5}$$
(3-33)

This form of Model 2 is more similar to the equivalent model developed from regression analysis of the synthetic data base, compared to the equivalent model developed from regression analysis of the measurement data. However, because it can be easily developed from Model 1, it is viewed as an appropriate general form for Model 2.

Model 3 can be developed assuming that the concept of predictable river hydraulic geometry (Leopold et al., 1964) can be considered a physical principle. Within this conceptual framework, theoretical and observed values for the down-the-channel and at-a-station relationship between discharge and depth both tend towards approximately the same relationship, $Y = kQ^{0.4}$ (Leopold et al. 1964). Thus, Model 3 can be developed by substituting $kQ^{0.4}$ for Y in the continuity equation yielding:

$$O = k W^{1.67} V^{1.67}$$
(3-34)

Model 4 is a special case of Model 1 (as represented by equations (3-26) and (3-29)), developed by assuming width is a function of depth. Based on the previous discussion, an appropriate assumed geometric shape for a channel cross-section is a parabola. With this assumption, the form of Model 4 would be

$$Q = kW_m Y_m^{-0.5} Y^{2.17} S^{0.33}$$
(3-35)

or in a more complete rational form

$$Q = Cg^{0.5}W_m Y_m^{-0.5} Y^{2.0} S^{0.33} / (Y_m S / Y)^{0.17}$$
(3-36)

derived by substituting the equation of a parabola in terms of depth in to Model 1 for the width, i.e. $W^2 = aY$ where $a = W_m/Y_m^{0.5}$.

Calibration of General Equations and Comparison with Comparable Regression Models

To facilitate a comparison between the derived general prediction models and corresponding statistically based models (derived from the measurement and synthetic data), the channel-control data base was randomly divided into a calibration (N = 680) and validation (N = 387) data set. Table 3.1 compares the range of data in both the calibration and validation data sets. The conductance (or discharge) coefficients for the general prediction models were

optimized by finding the constant value that minimized the log-residual of the predicted minus observed discharge for the calibration data. The log-residual was chosen for the minimization process because the discharge estimates are bounded by zero at the low end, with no constraint at the upper end.

The regression and general models were used to predict the discharge for the validation data set and the prediction errors of the various models were compared. The comparative error statistics included the log-residual, relative residual, actual residual (predicted minus observed) and the root mean square error (RMSE) of the predictions. Both the relative residual and the anti-log of the log-residual are measures of the percent error of the estimates. The comparative validation statistics of the models are shown on Table 3.3. The log-residuals for all of the general models are unbounded and in general are normally distributed as illustrated on Figure 3.6. The actual residuals and the relative residuals are not normally distributed because they are bounded by zero on the low end, with no upper boundary, thus they tend to have a skewed probability distribution.

The error statsitics indicate that the models derived from the synthetic data performed the worst. However, similar statistical results can be obtained using the synthetic data models if the coefficient is optimized from the calibration data in a similar manner as the general model (Table 3.3). This suggests that the general form of the synthetic models are applicable provided they are calibrated, similar to the general models. An interesting apsect of the synthetic models is that Model 1 (width-depth-slope) tends to overpredict discharge, whereas Models 2 and 3 (width-velocity-slope and width-slope) tend to underpredict discharge. These results suggest that the theoretical data used to develop the synthetic models under-represents the magnitude of resistance in the channel – with a subsequent smaller depth and higher velocity. This feature of the

synthetic data would not be expected to affect the previous conclusions regarding the prediction of Froude number from the velocity head index, because these variables are dimensionless.

The various forms of Model 1, with the exception of the one based on the non-optimized synthetic data, all performed similarly, suggesting that any of the general forms of Model 1 can be used with the same confidence as a model developed from multiple- regression analysis. The expected accuracy of this model, using ground-measured depth and width, and slope measured from a topographic map, would be better than 5% on average, and approximately +/- 50% two thirds of the time. Model 2 and 3 performed reasonably well, with mean accuracies of less than+/- 6% for all forms of the models except those derived from the non-optimized synthetic data. However, the estimates exhibit more variability than those using Model 1, with 67% of the estimates falling within a factor of 2. In general, Model 2 performed better than Model 3 and the regression models developed from the measurement data performed the best. The form of Model 4 developed from the measurement data performed the best. The form of Model 3 and the regression models developed from the measurement data performed as well if not slightly better than Models 2 and 3. Additionally, the rational form of Model 4 performed as well if not slightly better than the same model developed from regression.

ABLE 3.3

MODEL COMPARISON

Measurement Data (m^3/s) (m^3/s) Model 1Q = 4.84W ^{1.10} y ^{1.93} g ^{0.33} Mean0.1471.80.00130Q = 0.08W ^{1.16} y ^{1.95} g ^{-0.34} Mean0.19-99.30.01836.7Q = 0.028W ^{1.16} y ^{1.95} g ^{-0.34} Mean0.19-99.30.01836.7Model 3Stdev0.75687.70.19932.8Q = 0.23W ^{1.48} y ^{1.45} Mean0.16-103.20.00132.8Model 4Stdev0.576130.24937.4Q = 4.74W m ^{1.04} y ^{-0.56} y ^{2.11} g ^{0.28} Mean0.325.1-0.00237.4Stdev1.52707.40.28431.4Model 1VStdev1.27938.60.200Q = 8.42W ^{0.88} y ^{1.74} g ^{0.31} Mean0.49257.20.11051.4Model 2V0.26776.10.25742.6Model 3G0.21-218.9-0.16142.6Model 4V ^{1.53} y ^{1.86} Mean-0.24-199.8-0.20142.3Synthetic with Optimized CoefficientMean0.1539.70.00125.8Model 2Q = 0.09W ^{1.07} y ^{2.25} g ^{0.38} Mean0.1539.70.00125.8Q = 6.54W ^{0.68} y ^{1.74} g ^{0.31} Mean0.1539.70.00125.8Q = 0.09W ^{1.07} y ^{2.25} g ^{0.38} Mean0.1930.20.01529.2Stdev0.54552.20.257720.257Model 3Q = 0.18	Model Type Regression Models	Validation	Relative Residual			Root Mean Square Error
Model 1 Q = 4.84W ^{1,10} $\varphi^{1.85}S^{0.33}$ Mean Stdev0.14 0.9271.8 			<u>(Q' - Q)/Q</u>	<u>(Q' - Q)</u>	log(Q'/Q)	RMSE
$ \begin{array}{c c c c c c c c c c c c c c c c c c c $	Measurement Data			(m³/s)		(m ³ /s)
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	$0 - 4.9414/1.10 \times 1.63 - 0.33$					
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	Q = 4.04WV Y S					30
$ \begin{array}{c c c c c c c c c c c c c c c c c c c $	Model 2	Stdev	0.92	562.1	0.192	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	$\Omega = 0.08 W^{1.16} V^{1.65} e^{-0.34}$	84	0.40			
$\begin{array}{c c c c c c c c c c c c c c c c c c c $						36.7
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	Model 3	Sidev	0.75	687.7	0.199	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	$Q = 0.23W^{1.48}V^{1.45}$	Maan	0.40	400.0		
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$						32.8
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	Model 4	Oldev	0.57	013	0.249	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	$Q = 4.74W_{m}^{1.04}Y_{m}^{-0.58}Y_{m}^{2.11}S_{m}^{0.29}$	Mean	0.32	5 1	0.000	07 /
Synthetic DataModel 1 Q = $8.42W^{0.98}Y^{1.74}S^{0.31}$ Mean Stdev 0.49 1.27 257.2 938.6 0.100 Model 2 Q = $0.06W^{1.07}V^{2.25}S^{-0.38}$ Mean Stdev -0.21 0.36 -218.9 776.1 -0.161 0.257 Model 3 Q = $0.12W^{1.53}V^{1.85}$ Mean Stdev -0.24 0.45 -199.8 774 -0.201 0.290 Synthetic with Optimized Coefficient Model 1 Q = $6.54W^{0.98}Y^{1.74}S^{0.31}$ Mean Mean 0.15 39.7 0.99 0.001 25.8 Model 2 Q = $0.09W^{1.07}V^{2.25}S^{-0.38}$ Mean Mean 0.15 0.15 39.7 30.2 0.001 25.8 29.2 Model 2 Q = $0.09W^{1.07}V^{2.25}S^{-0.38}$ Mean Mean 0.19 0.19 30.2 0.015 0.257 29.2 Model 3 Q = $0.09W^{1.07}V^{2.25}S^{-0.38}$ Mean Mean 0.19 0.15 552.2 0.257 29.2 Model 3 Q = $0.18W^{1.53}V^{1.85}$ Mean Mean 0.15 59 -0.025 30.5 30.5						37.4
Model 1 $Q = 8.42W^{0.98}Y^{1.74}S^{0.31}$ Mean Stdev0.49 1.27257.2 938.60.110 0.200Model 2 $Q = 0.06W^{1.07}V^{2.26}S^{-0.38}$ Mean Stdev-0.21 0.366-218.9 776.1-0.161 0.25742.6Model 3 $Q = 0.12W^{1.53}V^{1.85}$ Mean Stdev-0.24 0.45-199.8 774-0.201 0.29042.3Synthetic with Optimized Coefficient Model 1 $Q = 6.54W^{0.98}Y^{1.74}S^{0.31}$ Mean Mean Mean 0.1539.7 39.70.001 0.00125.8Q = 0.09W^{1.07}V^{2.25}S^{-0.38}Mean Mean Mean 0.150.19 30.20.015 0.25729.2Model 3 $Q = 0.18W^{1.53}V^{1.85}$ Mean Mean 0.150.19 552.20.25729.2Model 3 $Q = 0.18W^{1.53}V^{1.85}$ Mean Mean 0.150.15 59-0.025 50.330.5	Synthetic Data	Sluev	1.52	/0/.4	0.284	
Stdev1.27938.60.200Model 2 $Q = 0.06W^{1.07}V^{2.25}S^{-0.38}$ Mean Stdev-0.21 0.36-218.9 776.1-0.161 0.25742.6Model 3 $Q = 0.12W^{1.53}V^{1.85}$ Mean Stdev-0.24 0.45-199.8 774-0.201 0.29042.3Synthetic with Optimized Coefficient Model 1 $Q = 6.54W^{0.98}Y^{1.74}S^{0.31}$ Mean Stdev0.15 0.9939.7 486.20.001 0.19925.8Model 2 $Q = 0.09W^{1.07}V^{2.25}S^{-0.38}$ Mean Stdev0.19 0.5430.2 552.20.015 0.25729.2Model 3 $Q = 0.18W^{1.53}V^{1.85}$ Mean Mean 0.150.15 59-0.025 30.530.5	Model 1					
Stdev1.27938.60.200Model 2 $Q = 0.06W^{1.07}V^{2.25}S^{-0.38}$ Mean Stdev-0.21 0.36-218.9 776.1-0.161 0.25742.6Model 3 $Q = 0.12W^{1.53}V^{1.85}$ Mean Stdev-0.24 0.45-199.8 774-0.201 0.29042.3Synthetic with Optimized Coefficient Model 1 $Q = 6.54W^{0.98}Y^{1.74}S^{0.31}$ Mean Stdev0.15 0.9939.7 486.20.001 0.19925.8Model 2 $Q = 0.09W^{1.07}V^{2.25}S^{-0.38}$ Mean Stdev0.19 0.5430.2 552.20.015 0.25729.2Model 3 $Q = 0.18W^{1.53}V^{1.85}$ Mean Mean 0.150.15 59-0.025 30.530.5	$Q = 8.42 W^{0.98} Y^{1.74} S^{0.31}$	Mean	0.40	257 3	0.440	F.4. 4
Model 2 Q = $0.06W^{1.07}V^{2.25}S^{-0.38}$ Mean Mean Stdev-0.21 0.36-218.9 776.1-0.161 0.25742.6Model 3 Q = $0.12W^{1.53}V^{1.85}$ Mean Stdev-0.24 0.45-199.8 774-0.201 0.29042.3Synthetic with Optimized Coefficient Model 1 Q = $6.54W^{0.98}Y^{1.74}S^{0.31}$ Mean Stdev0.15 0.9939.7 486.20.001 0.19925.8Model 2 Q = $0.09W^{1.07}V^{2.25}S^{-0.38}$ Mean Stdev0.19 0.5430.2 552.20.015 0.25729.2Model 3 Q = $0.18W^{1.53}V^{1.85}$ Mean Mean0.15 0.5459 59-0.025 0.02530.5						51.4
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	Model 2	01007	1.2.1	550.0	0.200	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	$Q = 0.06W^{1.07}V^{2.25}S^{-0.38}$	Mean	-0.21	212 0	0 161	42.0
Model 3 Q = $0.12W^{1.53}V^{1.85}$ Mean Stdev-0.24 0.45-199.8 774-0.201 0.29042.3Synthetic with Optimized Coefficient Model 1 Q = $6.54W^{0.98}V^{1.74}S^{0.31}$ Mean Stdev0.15 0.9939.7 486.20.001 0.19925.8 25.8Model 2 Q = $0.09W^{1.07}V^{2.25}S^{-0.38}$ Mean Mean Stdev0.19 0.5430.2 552.20.015 0.25729.2 20.257Model 3 Q = $0.18W^{1.53}V^{1.85}$ Mean Mean0.1559 0.15-0.02530.5						42.0
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	Model 3	01407	0.00	770.1	0.237	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	$Q = 0.12W^{1.53}V^{1.85}$	Mean	-0.24	-199.8	-0.201	12 3
Synthetic with Optimized CoefficientModel 1 $Q = 6.54W^{0.98}Y^{1.74}S^{0.31}$ Mean0.1539.70.00125.8Stdev0.99486.20.199Model 2 $Q = 0.09W^{1.07}V^{2.25}S^{-0.38}$ Mean0.1930.20.01529.2Stdev0.54552.20.257Model 3 $Q = 0.18W^{1.53}V^{1.85}$ Mean0.1559-0.02530.5						42.0
Model 1 Q = $6.54W^{0.98}Y^{1.74}S^{0.31}$ Mean Mean Stdev0.15 0.9939.7 486.20.001 0.19925.8 20.199Model 2 Q = $0.09W^{1.07}V^{2.25}S^{-0.38}$ Mean Mean 0.190.19 30.20.015 552.229.2 0.257Model 3 Q = $0.18W^{1.53}V^{1.85}$ Mean Mean0.1559 59-0.02530.5	Synthetic with Optimized Coefficient		01.10		0.200	
Model 2Stdev 0.99 486.2 0.199 Q = $0.09W^{1.07}V^{2.25}S^{-0.38}$ Mean 0.19 30.2 0.015 29.2 Stdev 0.54 552.2 0.257 Model 3Q = $0.18W^{1.53}V^{1.85}$ Mean 0.15 59 -0.025 30.5	Model 1					
Model 2Stdev 0.99 486.2 0.199 Q = $0.09W^{1.07}V^{2.25}S^{-0.38}$ Mean 0.19 30.2 0.015 29.2 Stdev 0.54 552.2 0.257 Model 3Q = $0.18W^{1.53}V^{1.85}$ Mean 0.15 59 -0.025 30.5	$Q = 6.54W^{0.98}Y^{1.74}S^{0.31}$	Mean	0.15	39.7	0.001	25.8
Model 2Q = $0.09W^{1.07}V^{2.25}S^{-0.38}$ Mean0.1930.20.01529.2Stdev0.54552.20.257Model 3Q = $0.18W^{1.53}V^{1.85}$ Mean0.1559-0.02530.5		Stdev				
Stdev0.54552.20.257Model 3Q = 0.18W ^{1.53} V ^{1.85} Mean0.1559-0.02530.5	Model 2					
Stdev0.54552.20.257Model 3Q = 0.18W ^{1.53} V ^{1.85} Mean0.1559-0.02530.5	$Q = 0.09W^{1.07}V^{2.25}S^{-0.38}$	Mean	0 19	30.2	0.015	29.2
Model 3 Q = 0.18W ^{1.53} V ^{1.85} Mean 0.15 59 -0.025 30.5						box V + Kis
$Q = 0.18W^{1.53}V^{1.85}$ Mean 0.15 59 -0.025 30.5	Model 3	0.004	VIVY	VV4.,6	0.201	
	$Q = 0.18W^{1.53}V^{1.85}$	Mean	0.15	59	-0.025	30.5
		Stdev	0.68	574.7	0.29	00.0

TABLE 3.3 (Continued)

Model Type	Validation	 Statistics 			
Regression Models		Relative Residual (Q' - Q)/Q	Actual Residual (Q' - Q)	Log Residual <u>log(Q'/Q)</u>	Root Mean Square Error RMSE
Calibrated General Models			(m ³ /s)	109(0704)	(m ³ /s)
Model 1					
$Q = 7.14WY^{1.67}S^{0.33}$	Mean	0.15	-24.6	-0.003	23.2
	Stdev	1.04	437.7	0.195	
Model 1 (Rational)					
$Q = 2.74 g^{0.5} WY^{1.67} S^{0.33} Y_{m}^{-0.17}$	Mean	0.15	-132.0	-0.007	31.7
	Stdev	1.21	585.9	0.201	
Model 2					
$Q = 0.05WV^{2.5}S^{-0.5}$	Mean	0.18	134.9	0.001	47.6
	Stdev	0.57	889.6	0.296	
Model 3					
$Q = 0.1W^{1.67}V^{1.67}$	Mean	0.18	183.9	-0.020	38.1
	Stdev	0.8	697.7	0.297	
Model 4					
$Q = 6.87W_{\rm m}Y_{\rm m}^{-0.5}Y^{2.17}S^{0.33}$	Mean	0.34	49.0	0.007	40.0
	Stdev	1.49	755.6	0.283	
Model 4 (Rational) Q = 2.64g ^{0.5} W _m Y _m ^{-0.67} Y ^{2.17} S ^{0.33}					
$Q = 2.64g^{0.5}W_{m}Y_{m}^{-0.67}Y^{2.17}S^{0.33}$	Mean	0.34	-68.2	0.004	36.2
	Stdev	1.61	681.4	0.287	

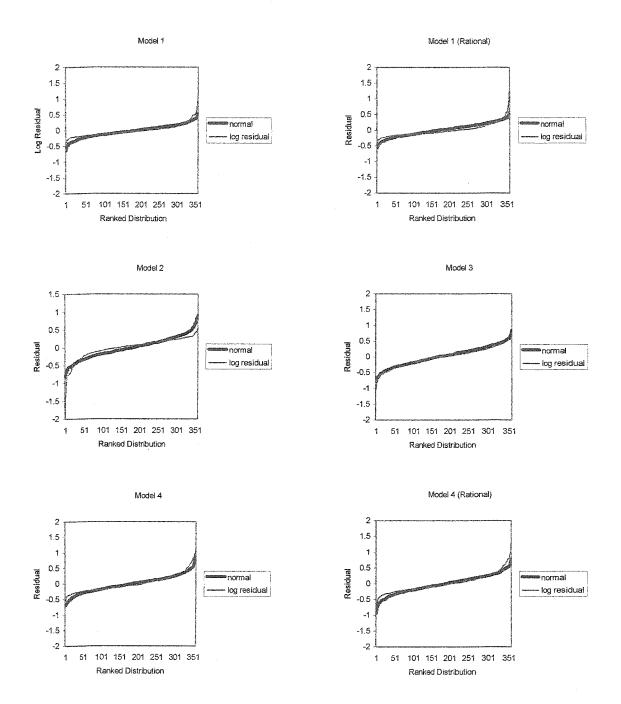


Figure 3.6 - Log-residual distribution from the validation data for Models 1 through 4 plotted with the equivalent normal distribution assuming the same standard deviation and mean of zero.

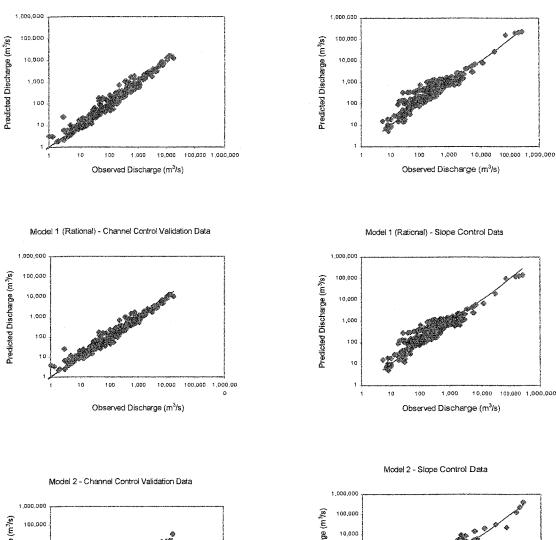
Slope-Controlled Reaches

The relationships that have been developed provide a set of equations that can be used to estimate in-bank river discharge depending on the type of information that is available. However, all of the relationships are based on an underlying assumption that the channel is adjusted to a characteristic slope. There are many river reaches where the channel is not adjusted to the slope, such as behind run-of-the-river dams, and where rivers are constricted by both natural and manmade features. In order to evaluate the use of the equations developed here for these types of rivers, a data set of discharge measurements were obtained from the USGS NWIS data base for rivers judged to exhibit control on the slope.

Selected flow measurements that did not meet the criteria for the channel-control data described in the hydraulic data section, were compiled into a slope-control data base that includes 293 measurements from 17 rivers, including the Amazon River at Obidos narrows (Oltman, 1968). The slope-control data includes rivers where there is an identifiable feature that creates a backwater or in other ways controls the hydraulic slope of the channel. These features include bridges or canyons that constrict the channel, and measurement stations that are located within run-of-the-river reservoirs behind dams or are suspected of being affected by backwater from dam and lock systems.

A data set of river stations where slope could not be effectively measured from topographic maps, and where large wetland and swamp systems are associated with the river channel were also compiled. These latter stations are presumed to exhibit significant lateral water exchange with the associated wetlands and swamps, and therefore the traditional concept of channel slope being the only significant mechanism driving the downstream motion may not be appropriate.

Figure 3.7 shows the predicted versus observed discharge for these data using the general models as compared to the same models applied to the channel control validaton data set. Figure 3.7 also shows that the estimates for the non-conforming reaches are generally subject to greater error. The mean prediction errors are significantly greater when applied to these data for all of the models. However, the standard deviation of the errors are comparble to those obtained for the channel control data. This suggests that in rivers where slope is controlled by hydraulic features, correction factors could be applied to the various models. Using the anti-log of the log residual as the best measure of prediction accuracy, the mean error for Model 1 is approximately 35% and the mean error for Model 2 is -58%. The mean error using Model 3, which does not use slope as a predictor variable is less than 5%, indicating that this model is the preferred model for situations where channel slope is not the primary hydraulic control. Interestingly, Model 4 showed the lowest mean error (less than 1%) suggesting that the additional information provided by the maximum width compensates to some degree for the effects of hydraulic control.



Model 1 - Slope Control Data

100,000 1,000,000

Model 1 - Channel Control Validation Data

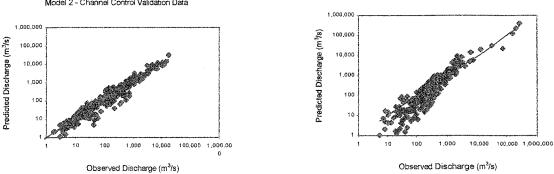


Figure 3.7 - Observed versus predicted discharge for the validation data and the slopecontrol data. The validation data are plotted on the left and the non-conforming data on the right.

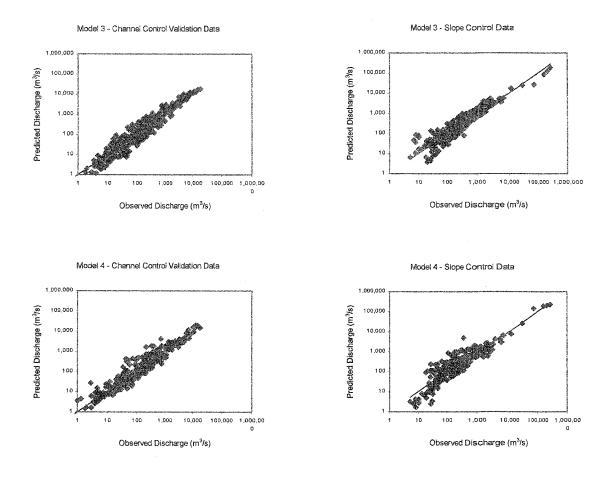


Figure 3.7 - (continued).

Discussion

Based on these comparisons, the calibrated general models and the models developed from the synthetic data base can be considered as useful and applicable as the regression models based on observed data. This suggests that river discharge is predictable from fundamentalhydraulic principles and can be estimated with reasonable accuracy for a wide range of flow conditions using constant values for coefficients calibrated on observations. An important finding is that uniform flow equations that use a slope exponent of 0.33 rather than 0.5 tend to have better predictive qualities in natural rivers, including less variation of the discharge coefficient and greater predictive accuracy. An advantage to using the general equations developed here rather than multiple-regression-based models is that they can be adapted to any flow conditon because they are based on well founded hydraulic principles and considerations rather than specific data sets. Thus, the assigned discharge coefficients can be adjusted based on knowledge of specific river reaches, for example where some ground-based data is available, without changing the predictive qualities of the model.

The general relationships provide a means to estimate in-bank river discharge from limited hydraulic information potentially obtained completely from remote sources. Model 2 or 3 combined with equations (3-11), (3-13) and (3-14), at a minimum, provides a method to estimate in-bank river discharge given knowledge of the bank-full width, wetted dynamic width and the channel slope. These variables can all be directly measured from remote platforms and available topographic information. The accuracy of the relationships vary, however the ability to use theoretically based synthetic data to generate models that predict as well as models developed from measured data suggests the general applicability of the formulations. Rationally derived relationships enable the predictive models to be calibrated and updated as specific knowledge is gained regionally or for individual rivers. In addition to the potential use of these relationships to estimate discharge in rivers from remotely obtained data, they can also be used in combination as tools to synthesize and map hydraulic geometry of rivers, and to interpolate hydraulic conditions in rivers based on limited field data or output from land-surface hydrology models.

CHAPTER IV

ESTIMATING DISCHARGE IN RIVERS USING REMOTELY SENSED HYDRAULIC INFORMATION

As discussed in Chapter 2, a mean water-surface width for a river can be readily measured (by measuring the wetted surface area and then dividing by the reach length) from a variety of existing remote imagery sources over large portions of the earth,. However, existing remote data sources do not provide coverages of river water-level elevations in areas where discahrge measurements are also readily available, and data sets of remote surface velocity mesurements are unavailable. Thus, at the present time, measurements of the water-surface width of rivers, combined with channel features such as the maximum channel width and the channel slope could be used to develop estimates of river discharge in remote areas or between river stations.

This chapter tests a methodology, based on the hydraulic relationships described in Chapter 3, to estimate in-bank river discharge using remotely sensed width information and channel-slope information obtained from topographic maps. Additionally, the use of watersurface velocity information observed from a single SAR image (Moller, 2002 personal communication) is used to evaluate the application and improvement in discharge estimates that can be achieved with this additional source of information. The results of these tests contribute to an assessment of the data requirements and potential accuracy of space-based discharge estimating methods.

Images and Remote Data

A data base of hydraulic information measured from various remote sources was compiled for this study. River reaches selected for analysis were located at or near established river gaging stations so that measured discharge values were available for comparison with estimates made from the remotely-sensed data. Mean daily discharge observations were obtained from the USGS NWIS on-line data base or from the Water Survey of Canada (Smith et al., 1996). Although the discharge estimates made from the remote data strictly only apply to the moment when the remote observation was made, the mean daily discharge, in all cases, did not vary widely through the day when the remote data were obtained. Thus, the average daily discharge is considered to be nearly equivalent to the instantaneous discharge at the time of the remote measurement.

Fourteen air photos, taken as part of the National Aerial Photography Program (NAPP), were obtained from the USGS EROS Data Center for analysis. The photos depicted the channel reach of 7 different rivers in New England near the corresponding USGS gaging station on each river during different flow conditions. These photos are geo-referenced and routinely taken as part of the USGS topographic mapping program. The photos were printed at a scale of 1:10,000. The mean water-surface width and mean maximum channel width were measured by averaging many equally spaced sections perpendicular to the channel banks.

Eleven digital orthophoto quadrangles (DOQs) available from the National Digital Orthophoto Program (NDOP), showing the selected river reach in 9 large rivers, were also obtained from the EROS data center for analysis. The resolution of the DOQs is 1 m. The watersurface width and maximum channel widths were measured from the DOQs by delineating the

total water surface and channel surface areas within the reach by defining the area of interest within a series of polygons. The polygons were fit as closely as possible to the observed boundaries, and then the total area of the polygons summed and divided by the total reach length to obtain the mean-width estimate.

The maximum channel width measured from the aerial photos and the DOQs was assumed to be the active channel (Figure 4.1), identified by the presence of sand and gravel bars, marked changes in vegetation on the channel banks (typically sparse) that suggest a riparian zone with frequent inundation, and areas where recent scour or deposition could be observed. Islands with prominent point bars and sparse riparian vegetation were included in the maximum width. Islands with stable vegetation and areas that appeared to be old meander scars or scars from scour were not included. The extent of the maximum channel width often varied considerably along the channel reach (Figure 4.1). In some cases, the maximum channel width was not an obvious feature and a certain amount of operator judgment was required to define its extent. Thus, determination of the maximum channel width is a source of operator error. Comparing the channel surface area delineated for the Missouri River and the Sacramento River in Figure 4.1, this source of operator error is most likely greater in highly active and irregular channels.

The localized variability is minimized by using aerial mean averages of width (and other variables) that more closely approximate the mean conditions in a channel, thus defining the appropriate reach length is a key element of the data collection. Leopold et al. (1964) and Leopold (1994) suggest that mean values for determining channel geometry should be averaged over at least one meander length (typically 11 channel widths) because this length reflects the energy dissipation regime of the reach. Rosgen (1994) suggests that data be averaged over a minimum of two meander lengths in order to provide the most meaningful values. For this study,

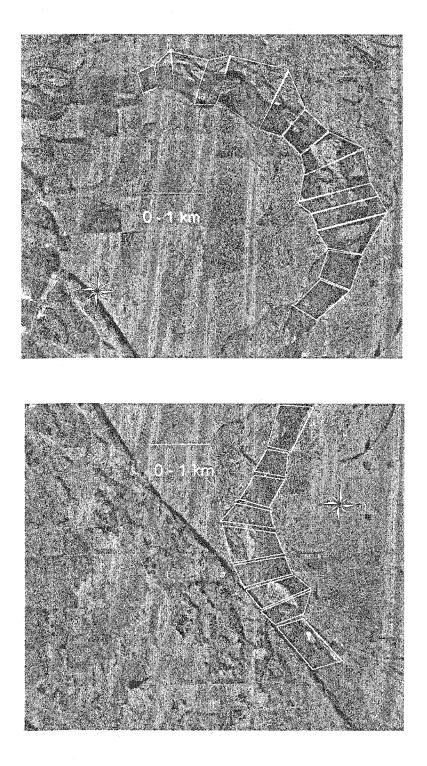


Figure 4.1 – Missouri River near Elk Point South Dakota, showing digitized plygons delineating the maximum channel surface area (Source: 3.75 minute DOQs for Elk Point (top) and Ponca (bottom) South Dakota, National Digital Orthophoto Program (NDOP)).

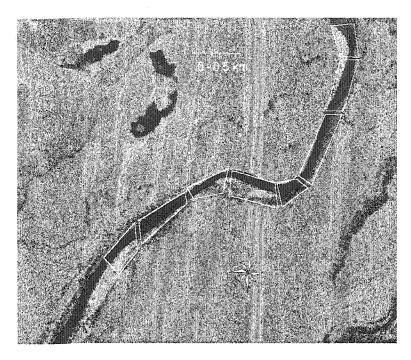


Figure 4.1 (continued) – Sacramento River near Red Bluff California, showing digitized polygons delineating the maximum channel surface area (Source: 3.75 min. DOQ for Bend, California, National Digital Orthophoto Program (NDOP)).

the widths were averaged over a reach length that included at least one meander wavelength and was limited to a length that did not inleude any tributary inflow or change in morphology.

The channel slope for all of the river reaches was measured from the corresponding USGS 1:24,000 scale topographic map by measuring the channel length between consecutive contour lines (approximately 3 meter contour interval). All of the images were obtained for river reaches at or near USGS stream gaging stations, and thus the mean daily discharge for the day of each image was available for comparison with the discharge estimates made from the images.

A time series set of SAR images obtained from ERS-1, were analyzed by Smith et al.

(1996) to measure the water-surface area at different discharges in three large braided rivers (the

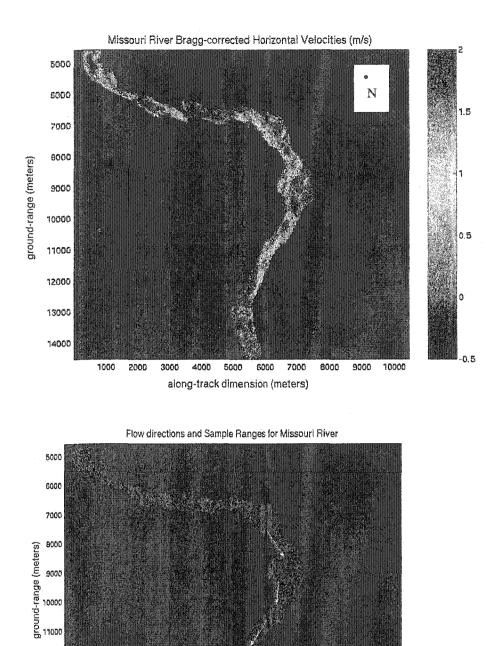
Tanana and Taku Rivers in Alaska, and the Iskut River in British Columbia). The resolution of the images was 25m with a processed pixel resolution of 12.5 m, and were collected at C-band. A total of 41 water- surface areas were obtained for the three rivers, 19 for the Iskut, 11 for the Taku, and 11 for the Tanana. The water-surface area estimates were made by summing all pixels classified as water based on a procedure developed by Smith et al. (1996). The total water-surface area within the braided channel system observed was divided by the valley length to obtain a mean or "effective" water-surface width for the reach. A measurement uncertainty was not reported (for more detail on these data and on the processing techniques used to extract the effective widths from the SAR images, refer to Smith et al. (1996)).

The reach lengths observed ranged from 9 to 16 km (approximately 20 to 30 times the effective width). The channel slope was assumed to be represented by the valley slope for the braided rivers, and was measured from topographic maps. A maximum channel width was not specifically measured by Smith et al. (1996). The maximum water-surface width from the time series was assumed to represent the maximum channel width for the purposes of this analysis. In each river, the maximum observed width occurred during high flow conditions, and likely reflects a high flow event that is near the mean annual flood. This assumes that the maximum channel width would generally correspond to a discharge near the mean annual flood (Leopold, 1964).

An airborne along-track interferometric (ATI) SAR imager (AirSAR), flown by NASA-JPL, obtained an image of the Missouri River near Elk Point South Dakota on March 25, 2002 (Figure 4.2). The resolution of the image was 5 m and was collected at C-Band. The watersurface width and the surface velocity were obtained from the image. The surface velocity was obtained using a Doppler technique developed by JPL (Goldstein et. al. 1994). Figure 4.2 shows the AirSAR radial velocity estimate projected onto the water surface. Note that in this figure positive velocities are flowing away from the radar to the south.

The velocities have also been corrected for the Bragg-resonant effect (Bragg 1913) whereby short wind-driven waves on the river surface have the effect of biasing the velocity estimate by their phase speed (Kinsman 1965). In this case the Bragg velocity is approximately 0.23 m/s although the correction increases with range due to the increasing incidence angle. At the time of the image, a mild wind blowing in the direction of the river flow (approximately 10 knots) was inferred from the nearby weather station in Sioux City, Iowa. Given the flat topography it is reasonable to assume that the wind direction in the imaged area will be consistent with the weather station's observation. Without the Bragg correction, the south-bound wind would have the effect of biasing the velocities high.

Because the ATI-SAR measures velocity in the radial direction only, the portion of the river which is oriented nearly parallel to the flight direction detects very low velocities (Figure 4.2). As such, for this case study, the analysis includes the region where the river is directed toward the radar. Techniques to alleviate this limitation are suggested in the discussion. The slope of the river channel was obtained from USGS topographic mapping, and the approximate maximum channel width was measured from a recent DOQ of the same reach (photo taken on April 4, 1993). The reach of river where the image was taken is characterized by large sand and gravel bars, and is much wider than both upstream and downstream sections of the river. This reach of river is considered atypical of the Missouri River for the region.



4000 5000 5000 700 along-track dimension (meters)

Figure 4.2 - C-band SAR image of the Missouri River near Elk Point, Souoth Dakota showing radial surface velocity projected onto the horizontal plan (upper) and inferred flow direction (lower) indicated by arrows (Source:NASA-JPL Air SAR).

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Because of the large sand and gravel bars, estimating the effective water-surface width was problematic. It was decided not to include shallow or sluggish water that did not contribute significantly to the flow. For this reason, the effective water-surface width was assumed to include only those regions where the surface velocity was greater than a threshold value. Taking this approach avoids potential complications associated with non-parallel flow lines, which would more likely be present over the shallow bars, and reduces the potential for assigning too much weight to generally non-contributing flow regions. The threshold velocity value used to estimate the mean water-surface width and velocity field was 0.15 m/s, thus velocities lower than this number were excluded when estimating mean velocities and river widths. This velocity threshold was chosen because below this value, the velocity estimate becomes too uncertain. This approach, while simplistic, was effective in excluding the sand-bar regions which would otherwise bias the velocity and width estimates.

A mean cross-channel width and velocity were determined for four portions of the observed river reach that were oriented towards the radar, and which were able to provide reliable estimates of both width and velocity. Vector velocity estimates were inferred from the radial velocities by assuming that the direction of flow was parallel to the river direction. Figure 4.2 shows the inferred direction of flow and regions of the river that were used to obtain four discharge estimates.

The river lengths varied in absolute range depending on the estimated direction of flow. Note that the flow direction estimates in Figure 4.2 are biased toward the high-flow regions and exclude the obvious sandbars (compare with upper frame of Figure 4.2). The absolute ranges were [1107, 765, 976, 730] m respectively (from north to south) while the estimated watersurface width (adjusted for the direction of flow and excluding sand-bar regions) was 330 m on

average (previously mentioned velocities <0.15 m/s were excluded from the estimation process). Although the range-to-width ratio is quite low, this was necessitated by the meandering nature of the river.

The accuracy of the water-surface and maximum-channel width estimates measured from the images are, in part, a function of the resolution of the images and the accuracy of the measuring tool. Thus, the resolution of the DOQs (1m) and the SAR images (10 m ERS-1 SAR, and 5m NASA-JPL AirSAR) indicate the accuracy of an estimated width measurement if it were a single measured value. However, the widths were estimated by measuring the total watersurface area of the reach divided by the reach length. This procedure would likely improve the accuracy of the estimate due to averaging. However, the methods used to measure the surface area may introduce additional unknown error. In the case of the NAPP aerial photos, the image resolution is a function of the ability to sharply see the boundary of the defined object (since these are not digital). It is estimated that the resolution of these photos at 1:10000 scale is approximately 4 m. The width estimates made from the photo is assumed to be somewhat better than the resolution implies, however, due to averaging along the reach (i.e. the balance of positive and negative estimte errors would tend to improve the overall estimate for the reach). Overall, the accuracy of the width estimates made from the various images is not precisely known.

Discharge Estimating Methodology

The observed water-surface width (W), bank-full (or maximum) channel width (W_m), and the channel or valley slope (S) can be used to estimate the river discharge at the time of the observation for the SAR images obtained by Smith, and for the 26 NAPP photos. This is accomplished by estimating the mean velocity (V) using a general relationship for estimating the

Froude number (F) and a relationship to estimate the discharge using width, velocity and slope as developed in Chapter 3. The Froude number estimate is obtained from the following general form of equation (3-13):

$$\mathbf{F} = \mathbf{c}(\mathbf{V}^2/2\mathbf{g}\mathbf{W})^{\mathbf{m}} \tag{4-1}$$

where
$$c = \alpha W_m^{0.12} S^{0.449}$$
 (4-2)

$$m = 0.881 W_m^{-0.046} S^{0.148}$$
(3-14)

The equations for determining c and m were developed from the synthetic flow-measurement data base, as described in Chapter 3. The coefficient α is assumed, for convenience, to be a calibration coefficient that reflects specific channel conditions. Calibration procedures for α are described later.

As can be seen from equation (4-1) and the variables used to predict c and m, all of the variables needed to compute F are obtained from the image except V. Combined with a dimensionally formulated discharge-estimating equation that uses width, velocity and Froude number given as:

$$Q = g^{-1}WV^{3}F^{-2}$$
(4-3)

where g is the accleration due to gravity, and equation (3-33) given below

$$O = 0.05 WV^{2.5} S^{-0.5}$$
,

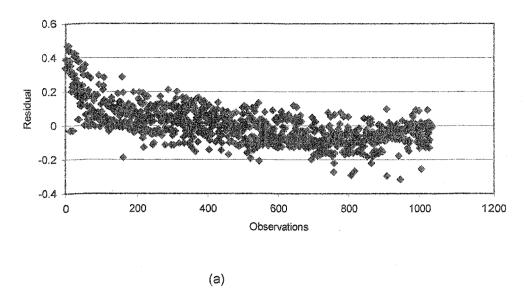
an estimate of the velocity can be made by substituting F from equation (4-1) (with the values of c and m determined from W_m and S) into equation (4-3), then equating this to (3-33) and rearranging. Once V is estimated, the discharge is then computed directly from equation (4-3). Thus, a discharge estimate can be made from a minimum of three variables all obtained from remote sources including: 1) observed water-surface width (W) measured from an image, 2) the maximum (or bankfull) channel width (W_m) measured from an image, and 3) the channel slope (S) measured from a topographic map.

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In order to evaluate the magnitude of the calibration coefficient α , the observed discharge measurements in the channel-control data base (Chapter 3) were used to calibrate equation (4-2). The best fit value of α found by minimizing the mean of the log residual of the estimates is 20. Figure 4.3a shows that the distribution of F, estimated using a constant value for α (= 20) is non-linear at low values (F less than about 0.2). The distribution was linearized by adjusting the value of α for observed inflection points in the distribution. Accordingly, for Froude numbers in the range 0 to 0.1, α was adjusted to 11.3, for Froude numbers in the range 0.1 to 0.2, α was adjusted to 17.7, and for Froude numbers in the range 0.2 to 0.4 and larger, α was adjusted to 22.3 (Figure 4.3b). The linearized values of α provide a means to self-calibrate equation (4-2) as follows: the Froude number determined from an initial value for α is used to determine a new value of α according to the Froude number ranges described above. The value of α is then adjusted accordingly. When the predicted Froude number and the value of α used to determine the Froude number are in the appropriate range class, the self calibration is complete (usually after one adjustment).

Discharge Estimation Results

Initial estimates of the discharge for the single-channel data were made using a value of 20.0 for α . Table 4.1 lists the observed data and the estimated discharges for the single-channel rivers derived from the air photos. The mean and standard deviation of the relative and log residuals of the estimates are also provided. Using the average value for the calibration coefficient resulted in very poor discharge estimate accuracy. Improvements in the estimates for



Froude Number Residual - Constant Value for Coefficient $\boldsymbol{\alpha}$

Froude Number Residual - Adjusted Value for Coefficient $\boldsymbol{\alpha}$

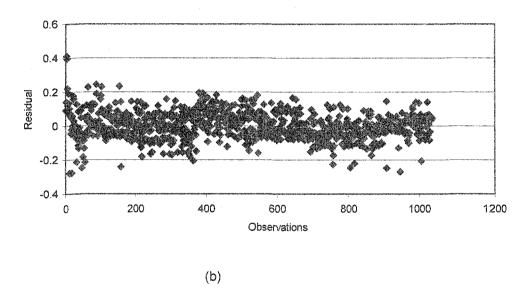


Figure 4.3 – Residuals of the predicted and observed Froude numbers when α constant value of a is used (Plot a), and when α is adjusted for Froude number ranges between 0 - 0.1, 0.1 - 0.2 and greter than 0.2, showing that the distribution becomes more linear.

the single channel rivers were made by adjusting the calibration coefficient as a function of the Froude number, as described above.

The initially estimated Froude numbers for the single-channel estimates greater than 0.2 were recomputed using $\alpha = 22.3$, and the initial Froude number estimates between 0.1 and 0.2 were recomputed using $\alpha = 17.7$. If the adjustment in α forced the Froude number out of range for the value of α used, then the previous α and the adjusted α were averaged. If the revised Froude number remained in range for the adjusted α , no additional adjustments were made. This approach does not require any new information or assumptions to be introduced into the calibration, and thus is considered to be a self calibration process. The revised discharge estimate accuracy is much improved (Table 4.1) using this calibration procedure.

An alternative calibration procedure was also explored by observing that the optimal value for α , determined by adjusting it until the predicted discharge equaled the observed discharge, was correlated with the maximum width of the river channels that were analyzed. Best-fit linear predictive relationships (Figure 4.4) between W_m and the optimized α were determined for rivers where W_m < 200m and W_m > 200m given by:

$\alpha = -0.075 W_{m} + 23.7$ and	$r^2 = 0.38$	for $W_m < 200m$	(4-4)
$\alpha = -0.005 W_{\rm m} + 20.3$	$r^2 = 0.79$	for $W_m > 200m$	(4-5)

nonometustasi sense antina antina River	Maximum		Water Surface	Observed Discharge	α = 20.0	колоциян , колонист иян	*****		Adjusted α from Froude number				Adjusted α from maximum width function			
	Channel	Slope			5°	19 - 6lan - 6a - 5	Phone and	1	Pasta dad	Patropast	Barrank				Damant	100
	Width		Width	(m³/s)		Estimated	Percent	Log		Estimated	Percent	Log		Estimated	Percent	Log
	(m)		(m)		Froude	Discharge	Error	Residual	•	Discharge	Error	Residual	Froude	Discharge	Error	Residual
Pemigewassett River at Plymouth, NH	60				Number	(m ³ /s)	and the second sec		Number	<u>(m3/s)</u>	-		Number	(m3/s)		
	82	0.0017	69.2	43.0	0.30		131.2	0.364		39.2	-8.8	-0.040	1		37.7	0.139
Pemigewassett River at Plymouth, NH	82	0.0017	78.6	78.0	0.30		94.5	0.289			-23.3	-0.115	0.29	90.4	15.9	0.064
Pemigewassett River at Plymouth, NH	82	0.0017	73.2	59.0	0.30		103.1	0.308		47.3	-19.9	-0.098	0.28	71.4	21.0	0.083
Pemigewassett River at Woodstock, NH	67.1	0.0026	54.6	26.0	0.33		180.9	0.449		32.3	24.4	0.095	0.31	40.7	56.7	0.195
Pernigewassett River at Woodstock, NH	67.1	0.0026	51.4	20.0	0.33		201.3	0.479			33.4	0.125		33.6	68.0	0.225
White River at West Hanford, Vermont	83.5	0.0012	78.6	93.0	0.27	135.3	45.5	0.163	1	48.3	-48.0	-0.284	1		-16.6	-0.079
Ammonoosuc River at Bethlehem, NH	27.9	0.0075	26.8	9.9	0.45		244.5	0.537	£	19.0	92.0	0.283	,		78.5	0.252
Ammonoosuc River at Bethlehem, NH	27.9	0.0075	15.7	6.4	0.41	7.2	11.7	0.048		4.0	-37.7	-0.206		3.7	-42.1	-0.237
Baker River near Rumney, NH	23.5	0.0013	19.9	5.4	0.26	16.1	198.1	0.474			22.0	0.087		5.7	5.0	0.021
Baker River near Rumney, NH	23.5	0.0013	16.9	3.5	0.25	9.4	169.3	0.430		3.9	10.3	0.042	1	3.3	-5.1	-0.023
Smith River at Bristol, NH	18.6	0.0037	17.7	11.0	0.36	16.5	50.4	0.177	1		-23.0	-0.113		7.3	-33.4	-0.177
Smith River at Bristol, NH	18.6	0.0037	13.6	6.1	0.34	7.5	22.4	0.088			-37.3	-0.203	0.31	3.3	-45.8	-0.266
Pomperaug River at Southbury, CT	18.4	0.0021	16.3	3.8	0.30	13.0	241.2	0.533		6.0	58.0	0.199	1	5.1	33.4	0.125
Pomperaug River at Southbury, CT	18.4	0.0021	13.1	3.3	0.28	6.5	98.0	0.297	0.26	3.0	-8.3	-0.038		2.6	-22.8	-0.111
Mississippi River at Thebes, IL	801	0.000137	710.0	14326.0	0.10	34.5	-99.8	-2.619		1444.5	-89.9	-0.996	ſ	18107.2	26.4	0.102
Mississippi River at Thebes, IL	801	0.000137	657.0	4700.0	0.09	21.5	-99.5	-2.339		901.8	-80.8	-0.717	0.18	11305.2	140.5	0.381
Potomac River at Point of Rocks, MD	381	0.00027	280.0	144.0	0.14	75.8	-47.4	-0.279		776.9	439.5	0.732	0.17	373.0	159.0	0.413
Missouri River near Elk Point, SD	651	0.00023	466.0	680.0	0.12	50.3	-92.6	-1.131	0.16	762.4	12.1	0.050	0.18	1764.8	159.5	0.414
Missouri River near Elk Point, SD	651	0.00023	336.0	450.0	0.11	9.7	-97.8	-1.667	0.14	147.0	-67.3	-0.486	0.16	340.3	-24.4	-0.121
South Platte River near Kersey, CO	125	0.00093	78.0	38.0	0.23	48.2	26.9	0.103	0.20	15.2	-60.1	-0.399	0.22	43.9	15.6	0.063
Missouri River near Culbertson, MT	343	0.000156	258.0	484.0	0.11	34.7	-92.8	-1.145	0.15	673.8	39.2	0.144	0.14	206.1	-57.4	-0.371
Kansas River at Fort Riley, KS	115	0.00049	77.0	17.5	0.18	40.7	132.5	0.366	0.18	40.7	132.5	0.366	0.18	31.0	76.9	0.248
Sacramento R. below Bend near Red Bluff, CA	163	0.000575	92.0	459.0	0.19	32.4	-92.9	-1.151	0.19	32.4	-92.9	-1.151	0.19	53.8	-88.3	-0.931
Willamette River at Satern, OR	219	0.00032	164.0	221.0	0.16	79.0	-64.2	-0.447	0.20	589.2	166.6	0.426	0.17	154.0	-30.3	-0.157
Delaware River at Port Jervis, DE	221	0.00098	162.0	172.0	0.24	167.9	-2.4	-0.011	0.21	50.2	-70.8	-0.535	0.25	264.8	54.0	0.187
Wenatchee River at Monitor, WA	126	0.0032	69.0	37.0	0.34	54.8	48.0	0.170	0.31	24.2	-34.5	-0.184	0.33	51.8	40.0	0.146
Mean	in an	an di banan anana kuala di sa dipanan	an second as a special second seco	iyiyaani dar oo taalahan diffanilara fir	T		50.4	-0.212			12.6	-0.116	n and a data data ana ang ang ang ang ang ang ang ang an	,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,	23.9	0.023]
Standard Deviation							114.4	0.899			108.5	0.421			64.1	0.285

α Predicted from Maximum Width

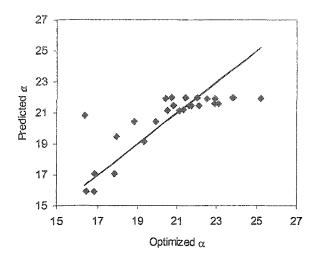


Figure 4.4 – Optimized α plotted against α predicted from maximum width relationships.

These calibration relationships improve the discharge predictions particularly for the Mississippi River, however they are based on limited data and are somewhat spurious because they are derived from the observed data (which for application purposes would be unknown). Both of the calibration procedures used to improve the discharge estimates indicate that specific channel characteristics (W_m) and the energy regime of the river reach (Froude number) are important to consider when applying the methods developed in this paper. Figure 4.5 a, b and c show the predicted discharge plotted against the observed discharge for the single-channel rivers using each of the calibration options described above.

Table 4.2 lists the discharge estimates developed for the braided channels derived from the SAR images by Smith et al. (1996). All of the Froude-number estimates for the braided channels were above 0.2, indicating that a value of 22.3 for α should be used to recompute the discharge. However, if this is done, the estimation accuracy becomes poor, with all of the

estimates biased low. This suggests that the calibration for braided river channels is different than for single- channel rivers. This would not be surprising, as mean values of depth and velocity, averaged across the braided channel system (i.e. the water-surface area) reflect somewhat different dynamics compared to the single channels. Many researchers have found that there is a distinct regime threshold between braided and single channel rivers (Henderson, 1966; Ferguson, 1986). Figure 4.5d shows the predicted discharges plotted against the observed discharges for the braided rivers.

The accuracy of the discharge estimates developed from channel width, water-surface width, and channel slope varied depending on the calibration procedure used. The error was evaluated from the relative residual ([Q' - Q]/Q) and the log residual ($\log Q' - \log Q$) where Q' is the predicted discharge and Q the observed discharge. For the single channel rivers, assuming a constant value for α , the estimates were rather poor (Table 4.1), exhibiting a mean error on the order of +/- 50%. The standard deviation of the error was large using either error index and a distinct break in the predictive quality for larger rivers was evident (Figure 4.5). The Froude number calibration markedly improved the prediction of 23% based on the anti-log of the log residual. The standard deviation of the errors was also markedly reduced and the distinct break in predictive quality for the larger rivers nearly eliminated. Even further improvement in the accuracy of the estimates was made using the width-based calibration method.

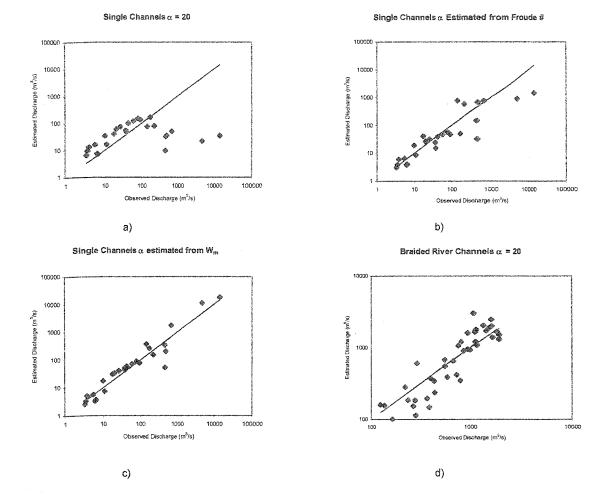


Figure 4.5 – Plots a, b, and c show the predicted discharge for the single channel rivers plotted against the observed discharge using different estimates of α . Plot d shows the predicted discharge plotted against the observed discharge for the braided rivers using a constant value for α .

The prediction error for the braided rivers was generally less than the error for the singlechannel rivers even though there was no calibration (Table 4.2, Figure 4.5d). This suggests that the braided rivers constitute a more homogeneous data set. These results suggest that grouping rivers by channel type, size and energy regime may provide a means to improve overall estimation accuracy, and that improved and more robust self-calibration methods could be developed based on experience. The SAR image obtained by NASA-JPL for the Missouri River provided both the surface velocity and water-surface width, enabling the use of equation (3-33) directly. The mean velocity for the cross-section was estimated by applying a correction factor of 0.86 to obtain the mean velocity in the vertical (Rantz et al., 1982). Recent experiments by Costa et al. (2000) in several rivers in which surface velocity was measured using bank-side and helicopter-borne radar showed that this correction factor appears to provide reasonable estimates of mean velocity in the cross-section.

Table 4.3 provides the measured values of water-surface width and mean velocity in four relatively short sections of river within the observed reach. The nearest USGS gaging station on the Missouri River is located at Sioux City, Iowa, approximately 20 miles downstream of the observed reach. For the date of the SAR image, the discharge at this station was approximately 450 m³/s. There are no major tributaries entering the River between the observed reach and the gaging station at Sioux city, so the discharge at Sioux City is assumed to be approximately the same as for the observed reach.

The discharge estimate using equation (3-33) is approximately 70% higher than the observed discharge. This is within the expected accuracy of the statistical model (Model 2), which indicates that approximately 67% of the estimates would be within a factor of 2 (Chapter 3). Given that the reach is non-conforming, i.e. it is atypical for the river, the relatively large error is not surprising. As a comparison, the width and mean velocity for two discharge measurements made at the Sioux City gage on March 6 and March 13, 2002 with approximately the same discharge (442 and 476 m³/s respectively) were 173 m and 165 m for width, and 1.00 m/s and 0.96 m/s for mean velocity. The channel slope at the gage is approximately the same as

for the observed reach. Using equation (3-33) with these data provides estimates of discharge of 597 and 514 m^3 /s respectively (errors of +8 and +35%).

River	Maximum Channel	Slope	Water Surface Width	Observed Discharge	α = 20.0	Estimated	Percent	Log
	Width		(m)	(m³/s)	Estimated			
	(m)				Froude	Discharge	Error	Residual
					Number	(m³/s)		
Iskut River, British Columbia	700	0.0022	437.0	292	0.32	604.1	106.9	0.316
	700	0.0022	579.0	951	0.34	1603.0	68.6	0.227
	700	0.0022	656.0	1570	0.35	2471.7	57.4	0.197
	700	0.0022	584.0	1110	0.34	1651.5	48.8	0.173
	700	0.0022	490.0	862	0.33	898.6	4.2	0.018
	700	0.0022	393.0	735	0.31	418.1	-43.1	-0.245
	700	0.0022	291.0	388	0.29	147.5	-62.0	-0.420
	700	0.0022	261.0	164	0.28	101.1	-38.3	-0.210
	700	0.0022	316.0	370	0.29	196.3	-47.0	-0.275
	700	0.0022	621.0	1320	0.35	2043.7	54.8	0.190
	700	0.0022	596.0	1140	0.34	1772.2	55.5	0.192
	700	0.0022	498.0	948	0.33	950.5	0.3	0.00
	700	0.0022	694.0	1080		3004.7	178.2	0.444
	700	0.0022	533.0	1121	0.33	1203.0	7.3	0.03
	700	0.0022	534.0	818		1210.8	48.0	0.17
	700	0.0022	446.0	681	0.32	648.4	-4.8	-0.02
	700	0.0022	311.0	235		185.7	-21.0	-0,10
	700	0.0022	294.0	266	0.29	152.8	-42.6	-0.24
	700	0.0022	381.0	403		375.5	-6.8	-0.03
Taku River ,Alaska	580	0.0015	301.0	277	0.26	183.5	-33.8	-0.17
	580	0.0015	358.0	436	0.27	342.1	-21.5	-0.10
	580	0.0015	541.0	1840	0.30	1508.0	-18.0	-0.08
	580	0.0015	520.0	1840		1308.0	-28.9	-0.14
	580	0.0015	360.0	801	0.27	349.1	-56.4	-0.36
	580	0.0015	229.0	309		68.7	-77.8	-0.65
	580	0.0015	339.0	221	0.27	281.3	27.3	0.10
	580	0.0015	288.0	136		156.6	15.1	0.06
	580	0.0015	290.0	130	0.25	160.5	29.5	0.11
	580	0.0015	574.0	1480		1865.4	26.0	0.10
	580	0.0015	491.0	765		1064.4	20.0	0.14
Tanana River, Alaska	865	0.0010	820.2	1764	0.30	1670.2	-5.3	-0.02
allalla River, Alaska	865		782.1					-0.02
	865	0.0010	733.3	1617	0.26 0.25	1390.7 1085.2	-14.0 -6.3	-0.08
		0.0010	733.3 562.2	1158				
	865	0.0010		595	0.23	390.2	-34.4	-0.18
	865 865	0.0010	494.4 407.9	445	0.23	237.9	-46.5	-0.27 -0.39
		0.0010		283	0.21	113.5	-59.9	
	865	0.0010	614.6	566	0.24	549.9	-2.8	-0.01
	865	0.0010	704.8	1000	0.25	931.6	-6.8	-0.03
	865	0.0010	825.9	1413		1715.4	21.4	80.0
	865	0.0010	857.7	1586		1983.9	25.1	0.09
	865	0.0010	649.1	561	0.24	678.5	21.0	0.08
Viean							3.8	-0.03
Standard Deviation							49.5	0.21

Table 4.2 - Hydraulic Data and Discharge Estimate Statistics for Braided Channel Rivers

Table 4.3	SAR Image Missouri River near Elk Point South Dakota Discharge Estimates Using Equation (3)								
Cross-section	Width (m)	Surface Velocity (m/s)	Estimated Mean Velocity (m/s)	Channel Slope	Estimated Discharge (m ³ /s)	Discharge at Sioux City (m³/s)	Percent Error		
1	315	1.13	0.97	0.00023	966.8	450	114.9		
2	313	1.07	0.92	0.00023	838.2	450	86.3		
3	370	0.80	0.69	0.00023	478.9	450	6.4		
4	321	1.05	0.90	0.00023	820.0	450	82.2		
Average	330	1.01	0.87		776.0		72.4	-	
	Discharg $\alpha = 22.3$	e Estimate	Using Maximum V	Vidth					
Cross-section	Width	Mean Velocity	Estimated Maximum Width	Channel Slope	Estimated Froude	Estimated Discharge	Discharge at Sioux City	Percent Error	
	<u>(m)</u>	_(m/s)	(m)		Number	<u>(m³/s)</u>	(m ³ /s)		
Mean	336	0.88	651	0.00023	0.2	435	450	-3.3	

The discharge estimate can be improved if the maximum channel width is also used in the analysis. The maximum channel width for the reach coincident with the SAR image was measured from recent DOQs (Figure 4.1, taken April 4, 1993), and estimated to average 651 m. Assuming that this value has not changed between 1993 and 2002, the method described previously to estimate discharge from the single-thread and braided river reaches was used to develop the discharge estimate with the inclusion of the measured surface velocity rather than the estimated velocity. The resulting discharge estimate was 434 m³/s, which is within 5% of the observed discharge at Sioux City. Thus, discharge estimates developed from two channel variables (maximum channel width and channel slope) and two dynamic variables (water-surface width and velocity) appear to provide an optimum set of hydraulic information for prediction (based on one observation).

Discussion

This analysis indicates that relatively accurate estimates of in-bank river discharge can be made from remote observations of water-surface width in rivers provided two channel-

characteristic variables are also known or measured remotely, the maximum channel width and the channel slope. However, it may be difficult to automate or readily obtain the maximum (bankfull) channel data, especially considering that a certain amount of judgment is required to define the maximum areal extent of the active channel. Assuming that the channel dimensions and the channel slope are relatively constant (at least over a period of years), inventories of this information can be developed from air photo and map analysis and from field surveys. These data can then serve as baseline information that is coupled with dynamic tracking of water-surface width to obtain time series estimates of river discharge over large areas or selected sets of rivers.

Another approach to defining the maximum channel width of rivers would be based on accumulated water-surface width measurements developed over time. Similar to the Smith et al. (1996) braided river-width data, a sufficiently long time series of widths would enable the maximum channel width to be identified and catalogued. This approach would be preferable to methods that rely on the identification of the active channel from morphologic features, because the water is relatively easy to identify. Additionally, identification of water surface areas and widths can be automated depending on the type of imagery (for example color infrared, SAR, and panchromatic) because water can be readily distinguished from surrounding land.

The accuracy of the discharge estimates reported on Table 4.1 and 4.2 indicate that robust calibration procedures will be necessary to successfully develop discharge estimates from imagery and other remotely sensed information. Experience may provide the data necessary to develop these methods, as there is strong indication from this analysis that characteristic channel features, including maximum width and channel type, can be used as calibration tools. Additionally, there is an indication from this analysis that self-calibration methods based on the Froude number can be developed. As data sets of remotely sensed water-surface widths,

velocities, maximum channel widths, and channel slopes are collected and associated with channel-type information, robust methods for assigning calibration coefficients can be developed.

The successful use of SAR imagery to simultaneously observe water-surface width and velocity holds great promise as a tool for substantially improving the accuracy of river-discharge estimates, especially when coupled with maximum-channel width and channel-slope information. Surface- velocity measurements require information about surface wind speed and direction in order to correct for these effects. For rivers in deep gorges one can generally assume that the wind will blow in the direction of the river banks and ameliorate this restriction. An additional limitation, whereby river flow orthogonal to the radar line-of-sight results in extreme radial velocities, may be addressed by flight lines that cross the river from alternate directions and deriving the vector velocities by assuming, as was done here, that the flow is parallel to the banks or by combining directionally diverse paths. A further preferable alternative is a system that can measure velocity in a single pass by means of directionally diverse multi-beam interferometric measurement capability (Moller et.al. 2002, Frasier and Camps 2001).

The equations developed in Chapter 3 indicate that discharge-estimating models that include width, depth, and slope have generally greater accuracy, especially for larger rivers, compared to models that use width and slope only; or width, slope, and velocity. Inclusion of remotely observed stage (water-surface elevation) from altimetry (Birkett, 1998) may provide an additional dynamic variable that can be used to estimate the depth and thus improve the accuracy of estimates even further. Depth estimates could be developed from stage if knowledge of the river-bottom elevation is available, or from time series of stage observations over a range of water levels. Observation of water-surface area (and width) and river-channel characteristics can be made with currently operating satellites, frequently and over much of the globe on a routine basis from a variety of sensors (Chapter 2). However, surface velocity and stage data may be available only on an occasional basis depending on the orbits of satellites, sensor capabilities and availability. In these circumstances, more accurate discharge estimates could be made when these data (surface velocity and stage) are available and used to calibrate routinely made estimates based on measured widths and map slopes. This approach would maximize the use of the more readily available data (water-surface area and channel slope) and enable less frequently available data (surface velocity and stage) to be successfully incorporated into a river-discharge observing strategy.

This analysis has shown that water-surface width, maximum channel width and channel slope can be used to estimate in-bank river discharge with an accuracy of 20% or better on average, however the standard deviation of the error could be 50 to 100% depending on the type of river and calibration technique. Additional data, including surface velocity (and stage) are likely to markedly improve the discharge estimates. Development of time-series data sets of water-surface area (and thus width), stage, and surface velocity of rivers will be key to fully developing robust estimating methods, calibration tools, and channel morphology inventories that will provide the basis for remotely tracking and estimating river discharge on large scales from space.

CHAPTER V

APPLICATION OF RIVER CHANNEL SLOPES DERIVED FROM A SIX MINUTE DEM FOR HYDRAULIC MODELING OF RIVERS

The water-surface slope is one of the key indicators of the hydraulic conditions within the river, and is an important predictor of the velocity, channel resistance, and stable channel geometry (Henderson, 1966). Most hydraulic models of river flow, sediment transport, bank stability, flooding, flood routing, and habitat conditions rely on an independently derived energy or water-surface slope as an input variable. Many hydraulic modeling applications (such as floodplain modeling) assume uniform flow between measurement points in a channel network, and thus inherently assume that the channel slope derived from channel-bottom elevations is equivalent to the water-surface and energy slope (Chow, 1959). With this assumption, the channel slope can be considered to be representative of the average energy slope in a river reach. Thus channel slopes obtained from topographic maps can be used in lieu of field-measured slopes in hydraulic models (as shown in Chapter 2), and provide a way of remotely obtaining estimates of channel slope.

However, measuring channel slopes from topographic maps is usually done manually, and is therefore labor intensive. Additionally, the scale of the map used to derive the slope is critical to its accuracy. Altimeters mounted on aircraft or satellites have the potential for measuring channel and water-surface slopes over large areas; however there are problems of accuracy inherent in these measurements due to the low slopes that rivers exhibit relative to surrounding topography and the accuracy of the altimeters themselves (Birkett, 1998). Digital elevation models (DEMs) can be used to obtain river channel slopes with automated routines, thus eliminating the labor intensive task of meausring slope from topographic maps. However, the use of DEM derived slopes for instream hydraulic studies is limited due to the difficulty of obtaining hydraulically meaningful values. This problem arises because routines used to develop the DEM cannot effectively determine the exact channel location and watersurface elevations within any specific grid cell. Often, a channel slope derived from a DEM will have large variability within a channel network, exhibiting sharp rises and troughs between adjacent grid cells. Fekete (2002, personal communication) has developed a method to estimate the channel slope by a technique that smooths large slope fluctuations between grid cells and maintains a continuous downstream slope direction.

This Chapter evaluates the application of a river-channel slope field generated from a six minute DEM by Fekete (2002, personal communication) for modeling of rivers. The riverchannel slopes obtained from the DEM are used in conjunction with river hydraulic variables potentially obtained or estimated from remote data sources to estimate discharge in rivers using a set of general hydraulic relationships based on the Manning equation. The potential accuracy of discharge estimating equations, which rely on slope as an input variable, is evaluated and some potential applications of using the DEM-derived slope and the hydraulic relationships are explored and discussed.

Data and Methods

River discharge measurements obtained from the USGS National Water Information System (NWIS) flow measurement data base were downloaded (http://www.water.usgs.gov/nwis/measurements/) for more than 5,000 gaging stations in the

United States. This data base consists of nearly one million records each providing measured discharge, flow width, flow cross-sectional area, mean velocity and other information about the gaging station and measurement conditions for stations across the United States. A DEM-derived slope was also available for each station (Fekete, 2002 personal communication).

The mean annual flow (Q_a) was used as a characteristic discharge for each station because it is considered to be correlated with the general morphological characteristics of the channel (Leopold et al., 1964; Osterkamp and Hedman, 1982) such that the water-surface slope associated with Q_a is assumed to be approximately the same as the general topographic channel slope. Additionally, the mean annual flow is assumed to be subject to fewer potential backwater effects from upstream or downstream controls, such as bridges, and is more reliably identified

For this analysis, the mean annual flow was determined from a composite flow field developed by Fekete (2002), which combines USGS long-term flow records with estimates made from a continental water-balance model (Vörösmarty et al., 1999). These data were used because the inclusion of the modeled flow data in the mean-annual discharge field provides a means to compensate for the varying record lengths inherent in the USGS flow data. The USGS flow measurement (Q_c) nearest to the estimated mean annual flow was extracted from NWIS flow measurement data base for analysis.

The USGS gaging station data were linked to a 6-minute gridded river network (STN-06, Fekete, 2002) for spatial analysis. The USGS-reported basin area was compared with the STN-06 calculated basin area, and all stations where the difference between the two basin areas was greater than 15% were discarded. This eliminated those stations with basin area less than the size of the 6 minute grid cell (approximately 100 km²). Stations with missing mean-annual flow data 106

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or missing channel-geometry data were also eliminated from the data base. The resultant data base included mean-annual-flow measurement data for 2,256 stations.

The characteristic flow measurement (Q_c) data were used with Model 1 (equation 3-26) to estimate an associated hydraulic slope. Model 1 is given by:

$$Q_{c} = KWY^{1.67}S_{c}^{0.33}$$
(5-1)

and is rearranged to calculate the slope:

$$S_{c} = [Q_{c}/(KWY^{1.67})]^{3}$$
(5-2)

where $S_c =$ the characteristic hydraulic slope

 Q_c = the characteristic discharge (m³/s)

 W_c = the characteristic water-surface width (m)

 Y_c = the characteristic mean flow depth (m)

K = a discharge coefficient

The hydraulic slope calculated from equation (5-2) represents a general or "typical" slope associated with the particular flow and channel geometry obtained from the characteristic-flow data. A constant value of K was determined through a calibration process that minimized the log-residual between the predicted and observed discharge from an independent data set consisting of over 1,000 flow measurements in 81 rivers (Chapter 3). The optimized estimate of K is 7.2, +/- 3.9 within one standard deviation. Because of the range of variability in the estimate of K, the calculated hydraulic slope has an associated uncertainty that reflects this variability.

Because the characteristic slope is derived directly from a calibrated hydraulic model, comparing this slope and the DEM-derived slope provides a means to evaluate the efficacy of the DEM slope for use in general hydraulic models. Additionally, evaluating the differences, or error, between the two slope estimates can provide insight into range of applicability of the DEM slope in hydraulic models.

The DEM derived slope (S_{dem}) was used in conjunction with the USGS measured hydraulic variables for each station as input to Model 1, Model 2 (equation 3-33) and Model 3 (equation 3-34) to estimate the discharge. Model 1 is given by equation (5-1), and Models 2 and 3 are respectively given as (Table 3.3):

$$Q = 0.05 W_c V_c^{2.5} / S_{dem}^{0.5}$$
(5-3)

$$Q = 0.1 W_c^{1.0/V_c} V_c^{1.0/V_c}$$
(5-4)

where $V_c =$ the characteristic mean velocity (m/s).

The discharge coefficients for Model 2 and 3 were optimized on the same data set as Model 1. Models 1 and 2 use S_{dem} and the measured characteristic values for width (W_c) and either depth (Y_c) or velocity (V_c) as input. Model 3 uses the measured characteristic width (W_c) and velocity (V_c), and thus is independent of slope. Discharge was estimated using the three models and compared against the measured characteristic discharge from the USGS NWIS data for each station. Because Model 3 is independent of slope, the effect of potential error in the slope can be evaluated by comparing the variability of this Model against the other two models.

Analysis and Results

The spatial distribution of the DEM-derived slope and the calculated hydraulic slope for the 2,256 river stations are shown on Figure 5.1. Figure 5.1 also shows the spatial distribution of the log residuals (error) between the hydraulic and DEM slopes. The log residual was chosen as the best measure of error because the error is bounded by zero on the low end and is not bounded on the high end. The log residual is calculated as:

$$Error = \log(S_{dem}) - \log(S_c)$$
(5-5)

The log residual can also be expressed as the log of the ratio of the two slopes:

$$Error = \log(S_{dem}/S_c)$$
(5-6)

Equation (5-6) shows that the antilog of the residual is the ratio of the DEM slope and the hydraulic slope, and thus can be thought of as a correction factor between the two slopes. This characteristic of the residual provides a direct measure of the percent difference between the slope estimates.

Inspection of Figure 5.1 shows that the distribution of the DEM slope appears to be consistent with the general topographic trends of the continental United States. The distribution of the hydraulic slope shows greater variability, possibly indicating the effects of smaller scale topographic relief on channel slope. Many of the slope residuals are quite large, in the range of several orders of magnitude. The mean residual between the two slopes is $(10^{-0.1})$ and the standard deviation of the residuals indicates that the difference between the two slopes is nearly one order of magnitude (Table 5.1).

Figure 5.2 and Table 5.1 show the discharge prediction results for the three models, with Model 1 and Model 2 using the DEM slope as input. It can be seen that the DEM slope provides reasonably accurate results using the models, and that results for Model 1 and 2, which use the DEM slope, are comparable to Model 3 which does not. Model 3 is comparable to Model 1 and 2 when a map-derived slope is used (Chapter 2 and 3). This suggests that the DEM slope provides results, on average, with the same accuracy as a map derived slope even considering some of the rather large deviations between the DEM slope and the hydraulic slope.

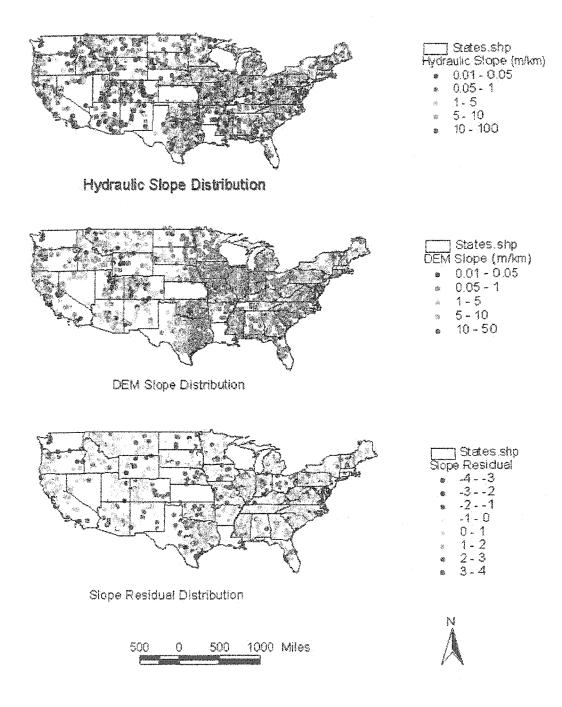


Figure 5.1 – Distribution of the hydraulic slope, DEM slope and the slope residual determined at 2,223 gaging stations in the continental United States.

Table 5.1 - Mean and Standard Deviation of the Log Residuals

Percent of Data	Condition		Residuals Slope log(S _{dem} /Sc)	Model 1 log(Q'/Q)	Model 2 log(Q'/Q)	Model 3 log(Q'/Q)
100%	All Slope Data	Mean Stdev	-0. 0.	10 0.00 86 0.29		
90%	Upper and Lower 5% Maximum Deviation of Difference Between DEM and Characteristic Slope	Mean	-0.	11 0.0	5 -0.08	-0.03
3078	Siope	Stdev		65 0.2		
80%	Upper and Lower 10% Maximum Deviation of Difference Between DEM and Characteristic Slope	Mean Stdev	-0. 0.	12 0.09 53 0.18		
70%	Upper and Lower 15% Maximum Deviation of Difference Between DEM and Characteristic Slope	Mean	-0.	13 0.04	4 -0.07	7 -0.01
		Stdev	0.	44 0.1	5 0.22	2 0.25
note:	S _{dem} = the DEM derived slope Sc = the characteristic hydraulic slope Q' = the predicted discharge					

Q = the measured characteristic discharge

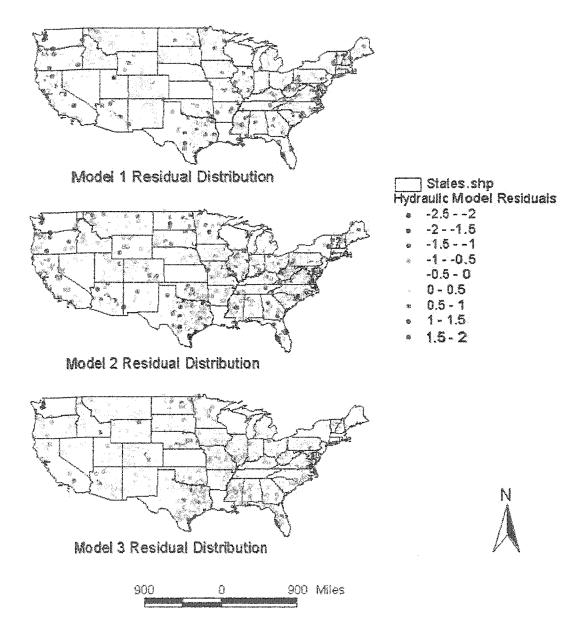


Figure 5.2 – Distribution of the discharge estimate residuals for discharge estimating Models 1, 2 and 3 determined at 2,223 gaging stations in the continental United States.

This conclusion, considering the large deviations between the DEM and hydraulic slopes, indicates that the discharge estimates using Models 1 and 2 are not highly sensitive to the slope, and that the DEM derived slope can be used in hydraulic river modeling with an acceptable level of confidence. However, given the rather large variation in DEM slopes compared to the calculated hydraulic slopes, understanding the variables that control the error could provide insight into the applicability and constraints of using the DEM slope for hydraulic modeling.

The slope residual and the residuals for each of the three models are normally distributed as seen on Figure 5.3. This indicates that inferential statistics regarding probable accuracy can be made when using the DEM derived slope and the hydraulic models. The ranked distributions also indicate that the DEM slope is unbiased relative to the hydraulic slope. This is indicated by the coincident residual distribution relative to a normal distribution with a mean of zero and the same standard deviation. However, the residuals from Models 1, 2 and 3 appear to be biased, as shown by the fact that the residuals plot either above (in the case of Model 1) or below (in the case of Model 2 and Model 3) the normal distribution. This suggests that these models could be linearly corrected by adjusting the magnitude of the coefficient of each model. However, because the models are used to predict only the mean annual flow, adjusting the model coefficients could result in greater errors for higher and lower flows if the models were used to predict a wider range of discharge at each station.

As an example, if the largest 5, 10, and 15% of the positive and negative deviations between the DEM slope and the hydraulic slope are eliminated from the data base, the discharge estimates improve significantly, as seen on Figure 5.4 and Table 5.1. This can also be seen on Figure 5.3, which shows that the largest residuals occur at the extreme ends of the ranked residual distributions for each model. Thus if the occurrence of the largest slope deviations can be

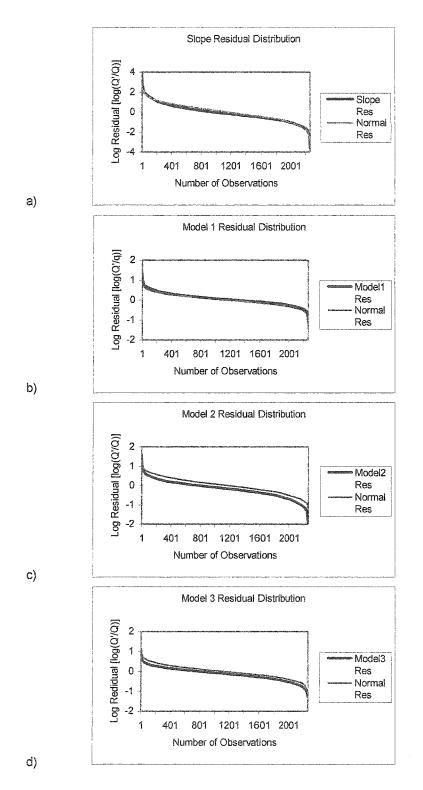


Figure 5.3 – Ranked log-residual distribution for the a) DEM and hydraulic slope; b) Model 1; c) Model 2; d) Model 3, plotted with a normally distributed residual with a mean of zero and the same standard deviation for comparison.

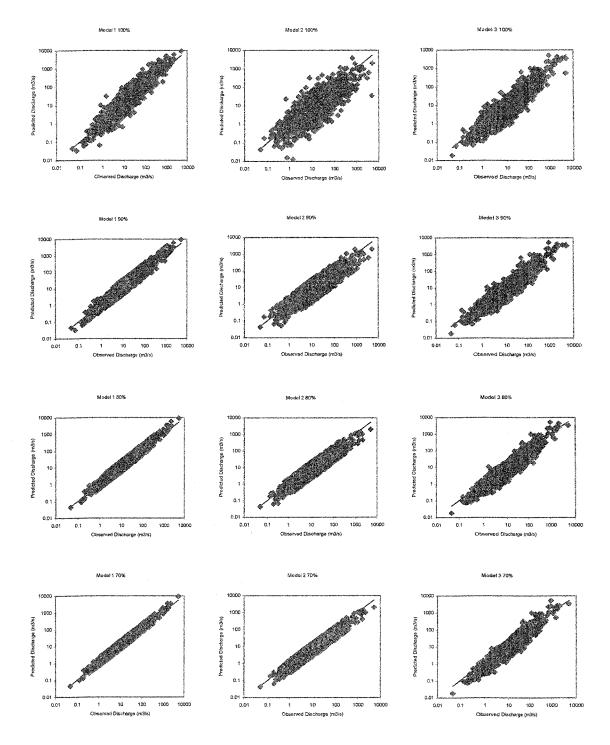


Figure 5.4 – Improvement in discharge prediction after eliminating the indicataed percentage of extreme positive and negative slope residuals.

predicted, application of the DEM slope for these streams can be understood to be subject to large error, and other slope measures obtained from maps or field surveys could be used for these reaches.

It can be reasoned that the slope deviations may be related to the variability in microtopography that is not captured adequately by the DEM, and error associated with using the general discharge coeffcient. Based on this, an initial assumption was that the smaller streams may be subject to more small scale topographic variation. However, measures of stream size including discharge, width, and basin area did not explain very much of the variation in the deviation either individually or combined, as evidenced by low correlation coefficients and high standard errors obtained from regression analysis. A combined variable, the width times the DEM slope, provided marginal improvement in predictability. This combined term indexes both stream size and potential energy gradient. Based on this, it was reasoned that these terms would provide a good index for predicting the deviation.

The Froude number, obtained from the USGS measurements, was chosen as an indicator of the balance between inertial and gravitational (retarding) forces in the channel. As it turned out, the Froude number was much more strongly correlated to the deviations than any of the stream size indices. This is evidence that the general dishcarge coefficient, which is related to the FRoude number, can explain much of the resdual error. Since the width-slope term provided some predictability, multiple regression of the Froude number and the width-slope term were used as predictor variables, and found to provide a good predictive model of the slope residual. Inclusion of the basin area as a predictor further improved the estimating relationship, as shown on Table 5.2 and Figure 5.5.

Table 5.2 - Prediction of Log Slope Residual

Regression Equation $\log(S_{dem}/S_c) = -2.794 + 0.794(\log[WS_{dem}]) - 3.000(\log[F]) - 0.180(\log[A])$

Regression Statistics	Coefficients	Star	ndard Err	t Stat	P-value L	ower 95%Up	oper 95%
Multiple R 0.96091	Intercept	-2.794	0.028	-99.651	0.000	-2.849	-2.739
R Square 0.923348	log[WS]	0.794	0.009	93.000	0.000	0.777	0.811
Standard E 0.238142	log[F]	-2.999	0.020	-149.019	0.000	-3.039	-2.960
Observatic 2257	log[A]	-0.180	0.007	-24.285	0.000	-0.195	-0.166

notes: S_{dem} = DEM slope

S_c = characteristic hydraulic slope

W = water surface width (m)

F = Froude Number

A = Contributing basin area (km²)



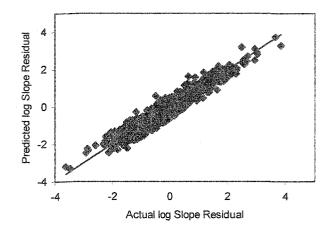


Figure 5.5 – Predicted slope residual plotted against actual slope residual.

With knowledge of the DEM derived slope, channel width and basin area, all which can be obtained remotely, coupled with a modeled or a-priori estimate of the Froude number, rivers which may provide relatively poor candidates for hydraulic modeling using the DEM slope can be identified. The question arises, however, as to the range of Froude numbers which are likely to result in relatively poor discharge estimates. This was evaluated by using the slope-deviation prediction equation with the width, basin area and DEM slope from the data base. Keeping all of the values of width, slope, and basin area constant, and varying only the Froude number, the mean and standard deviation of the log-residual discharge-prediction errors were determined assuming various ranges of the Froude number. It was found that within a Froude number range of 0.09 to 0.45, the prediction accuracy of the models was comparable to the case where the upper and lower 10% of the ranked slope residual errors were removed from the data (Table 5.1). The mean and standard deviation, respectively, of the errors within the above Froude number range was 0.03 and 0.22 for Model 1, -0.143 and 0.34 for Model 2, and – 0.1 and 0.26 for Model 3. Thus, it is concluded that the slope-residual predictive relation can provide selective knowledge about which streams can be modeled most accurately, and that relatively good predictability can be obtained for rivers that exhibit a range of Froude numbers between 0.09 and 0.45. This Froude number range is typical for many natural rivers, as evidenced by data compiled in Chapter 3 (Appendix) and as discussed by Leopold et al., (1964). Thus, if Froude numbers determined from modeling are outside of this range, the discharge estimate should be considered to be relatively inaccurate.

For comparison, a box plot of the log-residual range for the slope (slp) and the discharge prediction (Mod1, Mod2 and Mod3) have been summarized by physiographic province in the continental United States on Figure 5.6. The physiographic province boundaries were obtained from the USGS (<u>http://www.water.usgs.gov/pub/dsdl/physio.e00.gz</u>) and represent regions of similar topography, rock types, and geologic/geomorphologic history. The residuals for a fourth discharge estimating model (Modws) that requires only slope and width as predictor variables is also shown on Figure 5.6 to illustrate the effect of using only one dynamic variable (width) to predict the discharge. This prediction model is developed by equating Models 2 and 3 (equations 5-3 and 5-4) and then solving for the velocity. With velocity estimated, equation 5-3 is then used to estimate the discharge.

Figure 5.6 shows that Modws has the largest potential prediction bias and range of error, as anticipated. In general, the largest mean prediction bias for the other models occurs in the Atlantic plains region. The lowest prediction bias for these models occurs in western and interior regions. For Mod1, Mod2 and Mod3, the range of prediction error is greatest for Mod2, and is comparable for Mod1 and Mod3 across all regions. Mod2 also shows the largest potential (negative) bias, and Mod1 shows the largest potential for positive bias.

The slope residual shows a general low bias in the interior regions, and a high bias in the Atlantic Plains region, similar to Figure 5.1. The error in Mod1, Mod2 and Modws, which use slope as a predictor, does not always follow the slope error. This is due to the opposite predictive effective that the Froude number and the combined WS parameter have on estimating the magnitude of the slope residual. The trend and variability of the discharge-estimate residuals illustrated in Figure 5.6 indicate that the models will provide the most accurate estimates in the Interior Plains and Rocky Mountain System regions because of less variability in the slope residual. This may indicate that channel slope generally conforms to the topographic slope in these regions with less small-scale variation. Additonally, the error distribution suggests that regional adjustments to the discharge coefficients could be made to account for the observed variance. For example, in the Atlantic Plain region, a lower coefficient value using Model 1 and a higher coefficient value using Model 2 would correct much of the error using these models in this physiogrtaphic region.

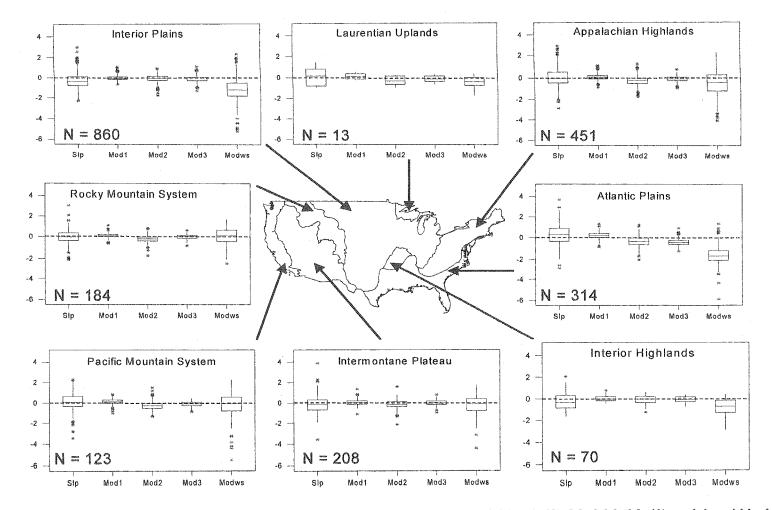


Figure 5.6 – Boxplots of the log-residual of the slope (slp), Model 1 (Mod1), Model 2 (Mod2), Model 3 (Mod3), and the width-slope model (Modws) summarizing the regional distribution.

Applying the DEM Slope to Large Scale River Modeling

The general hydraulic relationships developed in Chapter 3, used in combination, enable the discharge to be estimated from a minimum of necessary data including the maximum or bankfull channel width, the dynamic width (i.e. the water surface width at the time of the observation) and the channel slope. Conversely, these relationships can also be used to develop channel geometries if the discharge, width and slope are known. This latter capability is of special application to large-scale land-surface water-balance and runoff modeling because it provides the framework for development of realistic river-routing schemes. Assigning realistic channel geometry to the river network can also provide an estimate of the channel capacity, and hence the occurrence of over bank flooding can be modeled.

An example of the application of these relationships in this capacity can be demonstrated using a high-resolution runoff field developed by combining a water-balance runoff model (WBM) with observed discharge from ground-based discharge monitoring networks (Fekete and Vorosmarty, 2002). The mean annual discharge values are derived for a gridded river network at a 30 minute spatial resolution to obtain a mean annual discharge field for North America. An approximation of the bankfull-channel width is then estimated for every 30-minute grid cell along the river-channel network using a general regime relationship (Leopold et al, 1964) that relates the bankfull channel width with the mean annual discharge. Osterkamp and Hedman (1982) have statistically developed the coefficients of this relationship from a large data base of rivers in the Missouri River Basin

$$W_b = 8.1 Q_a^{0.58}$$
 (5-7)

where $Q_a =$ the mean annual discharge (m³/s)

 W_b = the bankfull or regime channel width (m).

A similar relationship correlating the bankfull width to the mean annual flood, taken to be equivalent to the bankfull discharge (Leopold et al, 1964) has also been determined from a data base of bank-full channel geometry compiled from various sources (Schumm, 1960; Barnes, 1967; Osterkamp and Hedman, 1982; Church and Rood, 1983; Dingman and Palaia, 1999) (see Chapter 3, and Appendix3). The channel geometry data were measured in the field and the bankfull discharge estimated according to various methods for 521 rivers in North America. The resulting relationship is

 $Q_{\rm b} = 0.24 W_{\rm b}^{1.64} \tag{5-8}$

where $Q_b =$ the mean annual flood (m³/s).

The estimated bankfull width and discharge, obtained from the mean annual discharge via equations (5-7) and (5-8), are coupled with a general physically based discharge relationship (equation 3-29) to estimate the bank-full depth and velocity, thus defining the bankfull channel geometry and flow regime. For the condition where the dynamic wetted width (W) equals the bankfull width and the dynamic mean depth (Y) equals the bank-full depth, the bankfull depth can be calculated given the bankfull discharge from (equation 3-29):

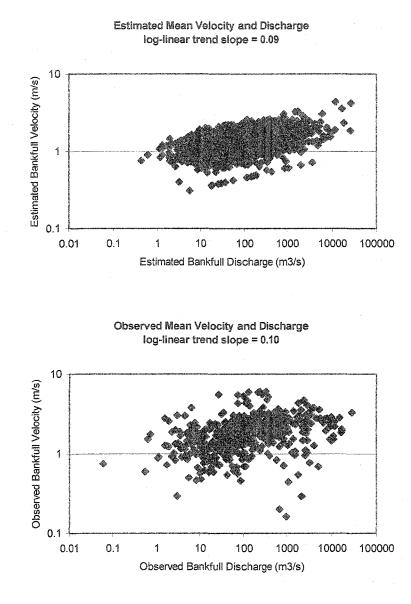
$$Q = 2.74g^{0.5}WY^{1.67}S^{0.33}/Y_b^{0.17}$$
(5-9)

The slope can be taken as S_{dem} , and the bankfull velocity can then be estimated from the equation of continuity, $V_b = Q_b/(W_bY_b)$. Figures 5.7 and 5.8 show the estimated mean bankfull depth and estimated mean bankfull velocity plotted as a function of the estimated bankfull discharge for grid cells coinciding with 2,256 USGS gaging stations. These are compared to similar plots obtained from the bankfull channel geometry data base.

log-linear trend slope = 0.30 100 Estimated Bankfull Depth (m) 10 ~ 0.1 0.01 1 10 100 1000 10000 100000 0.1 Estimted Bankfull Discharge (m3/s) **Observed Depth and Discharge** log-linear trend slope = 0.37 100 Observed Bankfull Depth (m) 10 1 0.1 0.01 1 10 100 1000 10000 100000 0.1 Observed Bankfull Discharge (m3/s)

Estimated Depth and Discharge

Figure 5.7 – Mean bankfull depth plotted against bankfull discharge for the estimated and observed data showing the general distribution patterns.



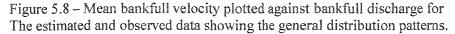


Figure 5.7 shows that the estimated depths have a similar pattern of variability compared to the observed depths, however the trend slope for the estimated values is less steep. The estimated velocities (Figure 5.8) show a similar trend compared to the observed data, but with less overall variability. In, general, the estimated depth and velocity compare well with the observed relationships.

A similar approach would combine the estimated bankfull width and discharge to calculate the bankfull velocity using a width-velocity relationship (that does not require slope) as described by equation 5-4, or other forms of Model 3 developed in Chapter 3 (Table 3.3) shown as equations (5-10) and (5-11) below:

$$Q = 0.23 W^{1.48} V^{1.45}$$
(5-10)

$$Q = 0.18 W^{1.53} V^{1.85}$$
(5-11)

.

Once velocity is estimated, the bankfull depth is calculated from continuity. The estimated bankfull depth, therefore is dependent on which equation is used to compute the mean bankfull velocity.

Histograms of the approximated values for the bankfull discharge and width developed from the mean annual discharge field for North America using equation (5-7) and (5-8) are shown on Figure 5.9. The mean width for 10043 30 minute grid cells (approximate land surface of North America) is 113 m, and for velocity and depth it is 1.59 m/s and 2.64 m using equation (5-4), 1.62 m/s and 2.32 m using equation (5-10), and 1.46 and 2.68 using equation (5-11). Figure 5.9 also shows histrograms of the approximated values for bankfull velocity and depth derived from equations (5-4), (5-10) and (5-11). Figure 5.9 indicates that bankfull velocity is much less variable than either bankfull width or depth, suggesting that it is relatively constant for a wide range of river channels. This is in agreement with general predictions of regime theory, which indicates a small velocity exponent when correlated with the bankfull discharge (Savenije, 2003; Lacey, 1935; and Leopold, 1964). However, observed values (Bray, 1979; Williams, 1978; and Church and Rood, 1983) of bankfull velocity tend to have a much wider range than predicted here or suggested by regime theory, thus the velocity derived from equation (5-1), which is based on statistical analysis of actual discharge measurements, appears to be the most realistic.

If equation (5-10) is used to estimate velocity, then the estimated depth is lower (Figure 5.7) than if equations (5-4) or (5011) are used. Thus, equation (5-10) may bet he most appropriate if the goal is to model velocity, and equations (5-4) or (5-11) would be most appropriate if the goal is to model depth. Equation (5-11) yields a greater range of velocity than equation (5-4), with a similar range of depth, thus equation (5-11) appears to provide the best overall values for both depth and velocity. The inclusion of slope in the development of the channel geometry would introduce more site specific information and would therefore result in a larger range of velocity and depth which would be more realistic, as demonstrated by Figures 5.7 and 5.8. Figure 5.10 shows the spatial distribution of the estimated bankfull width, depth and velocity assigned to the river network for North America using equations (5-7), (5-8) and (5-11).

Once the bank-full channel geometry is defined, a depth-discharge rating can also be defined. This is accomplished using equation (5-9) assuming a suitable channel cross-section shape. If a parabola is assumed, the following general discharge equation for in-bank channel depths is obtained from equation (5-9):

$$Q = 2.74g^{0.5} (W_b/Y_b^{0.67}) Y^{2.17} S^{0.33}$$
(5-12)

Because W_b , Y_b and S are constant values, equation (5-12) defines a unique depth-discharge rating for depths ranging from 0 to Y_b .

A depth can be calculated from equation (5-12) for any in-bank discharge estimate generated from the WBM. Given the assumed channel cross-section shape, a width and velocity can also be calculated, thus defining the necessary routing parameters for the channel network. Thus, an explicit river runoff routing scheme for the WBM model can be developed. Because the bank-full depth is prescribed, when the model-estimated discharge results in a depth exceeding this value, over bank flooding is assumed to be occurring. When over-bank flow occurs, routing can be adjusted to account for over- bank storage. Additionally, this capability can be used to evaluate

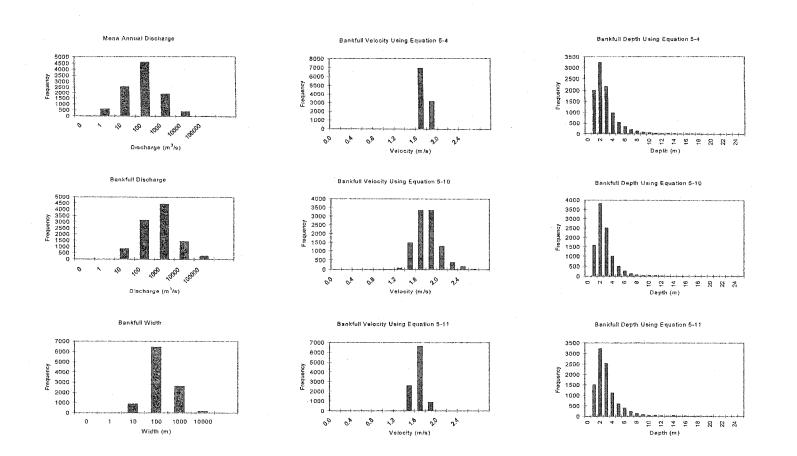


Figure 5.9 – Histograms showing the 30 minute distribution of approximated bankfull hydraulic variables for North America including discharge, width, velocity and depth. Velocity and depth distributions estimated from equations (5-4), (5-10), (5-11) are shown.

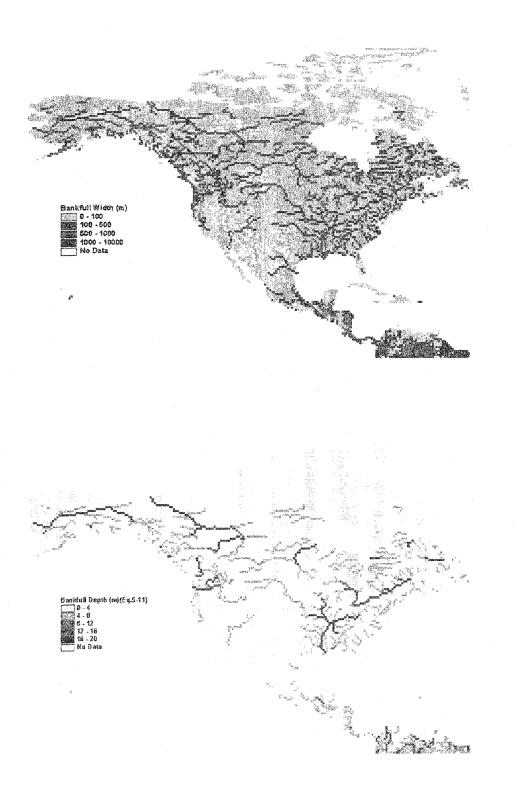


Figure 5.10 – Distribution of estimated bankfull width (top) and bankfull depth (bottom) for North America based on general width-discharge function and equation (5-11).

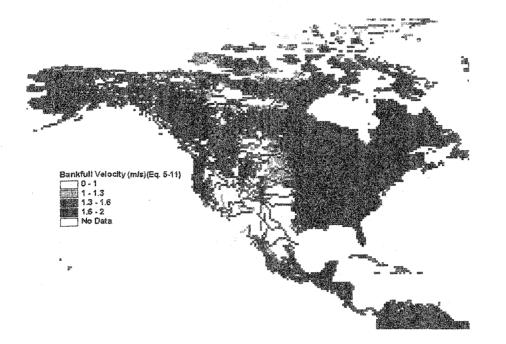


Figure 5.10 (continued)– Distribution of estimated bankfull velocity for North America based on general width-discharege function and equation (5-11).

the frequency of over bank flooding generated by the model, and coupled with the DEM, could define the areal extent of flooding.

The strategy outlined above cannot provide reach-specific rating curves because equations (5-7) and (5-8) represent relationships fitted to data from diverse geographic and hydrologic regions, with relatively large potential estimation errors. In order to evaluate the specific prediction errors associated with the procedure developed above, equation (5-9) was used to calculate the expected depth associated with the mean annual discharge, which was then compared to the observed depth associated with the mean annual discharge obtained for the 2,256 USGS gaging stations.

Figure 5.11 compares the estimated and observed mean flow depth, and shows relatively poor agreement, tending to under-estimate at high depth, and over-estimate at low depth. This suggests that the initial estimates of the bank-full width and the bank-full discharge, obtained from equations (5-7) and (5-8), do not adequately reflect the channel-specific conditions at the gaging stations. In order to provide greater site specificity, observed elements of the actual channel geometry within each reach would be needed. Bank-full widths obtained from imagery would provide sufficient additional site-specific information such that equation (5-7) would not be necessary. Additionally, if dynamic widths were also available, a unique channel cross-section shape need not be assumed, because equation (5-9), which does not assume a specific cross-section shape, could be used to develop the rating. Thus, it is anticipated that a more accurate river routing scheme can be developed by coupling the general hydraulic relationships with observed channel width (bankfull), the dynamic water-surface width and the channel slope.

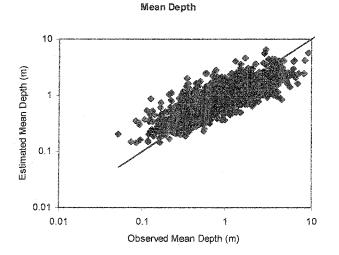


Figure 5.11 – Estimated mean depth plotted against the observed mean depth for the mean annual flow data.

Discussion

River-channel slopes developed from a 6-minute DEM using the Fekete method can be effectively used in general uniform-flow hydraulic models of river flow. This enables the largescale application of river-flow models that require an independently derived channel slope as an input variable. General flow-routing models applied on a global or continental scale can therefore be effectively developed and linked to large-scale water-balance and runoff models. Additionally, large-scale evaluation of in-stream hydraulic conditions in rivers can be made where high accuracy in any given reach is not critical. The DEM slope could also be coupled with remotely sensed estimates of river channel and dynamic width to estimate discharge in rivers over large areas.

Errors associated with using the DEM-derived slope are normally distributed over a wide range of rivers in the United States. The magnitude of the error between the DEM slope and the functional hydraulic slope associated with each discharge measurement is a function of the DEM slope, the river width, the contributing drainage area, and the Froude number of the flow. Because the error associated with using the DEM slope in lieu of a hydraulic slope is normally distributed and predictable, straight-forward statistical evaluation of modeling results can be undertaken, and those rivers that exhibit conditions conducive to greater modeling errors can be identified.

With improvements in the accuracy of DEM models, it can be assumed that estimates of the channel slope would improve, enabling even better hydraulic-modeling results in the future. The DEM slope could also be used in conjunction with available field data and in-stream hydraulic information to develop an improved composite field of the hydraulic slope in rivers.

CHAPTER VI

SUMMARY AND CONCLUSIONS

The development of methods to estimate the discharge of rivers using remotely sensed data will provide the means to increase the streamflow measurement network globally. This component of the land-surface water-budget is currently measured at ground-based gaging stations for many of the larger rivers in populated regions, however large rivers in remote areas and small to intermediate sized rivers over much of the globe are not currently monitored. Additionally, the global river-gaging network and access to these data have been decreasing in recent years. Because of these trends, the current ground-based streamflow gaging network does not provide adequate spatial coverage for many scientific applications, including verification of the land-surface runoff contribution to the oceans and the spatial distribution of intra-continental runoff.

Calibration of continental scale runoff and climate models depend on adequate spatial density and length of streamflow records. Remote sensing of river discharge has the potential to provide this needed data by filling in gaps within the existing streamflow gaging network, and by adding new information from inaccessible regions that have not been gaged in the past. Generally applicable open-channel hydraulic equations, including the Manning and Chezy equations, have been in use for decades, and can be adapted to remote sensing applications because the dynamic constitutive elements of the equations can all be measured or potentially measured remotely, provided a general estimate of the resistance can be made.

This research has shown that much of the variance associated with resistance can be predicted and measured through its effect on other elements of flow, including depth, velocity and channel slope. In particular, the usual formulation of the Manning and Chezy equation states that velocity varies as the square root of slope, whereas this research has shown that greater variance in velocity can be explained if the cubed root of slope were used. A physical explanation of this phenomena has been developed, and modified forms of the Manning and Chezy equations described that minimize the uncertainty associated with the resistance term.

Additionally, this research has shown that estimating the in-bank discharge of rivers from remotely sensed hydraulic information can be accomplished with reasonable accuracy (on average within 20% of the ground measured value) given observations of reach averaged water-surface width and maximum-channel (bankfull) width coupled with channel slope. Additional information such as surface velocity, measured using Doppler lidar or SAR techniques, appear to enable much higher accuracies. Mean accuracy for large numbers of estimates can be expected to be within +/- 20% of the actual discharge, with a relatively wide range of variability. For example, the accuracy for 67% of a large number of estimates (one standard deviation) made over a wide range of rivers would be expected to be within +/- 50 to 100% depending on the model used and the data available.

Satellite and other remotely obtained images of the land surface have the capability of providing accurate measurments of the water-surface width of rivers around the globe on a nearly real-time basis. Additionally, satellite imagery and other remote sources of land-surface information can provide measurements of channel geomorphic characteristics including the bankfull (or active) channel width and the channel slope. There are potential difficulties in measuring water-surface slope; however measurement of slope from topographic information,

and thus a channel constant for a particular river reach, can be used in lieu of a measured watersurface slope in discharge- estimating equations while maintaining reasonable accuracy.

There is an indication that general features of a river, such as its channel morphology and size (indicated by its maximum width), can be used to self-calibrate the estimation procedure, thereby improving the accuracy of remotely based estimates. Additionally, self-calibration methods based on the predicted Froude number also show promise with regard to improving estimate accuracy. As more river hydraulic data become available from satellites and aerial surveys, improved methods and calibration procedures can be developed. These improvements will be based on experience with large data sets of remotely sensed hydraulic and river-channel information.

Considering that traditional ground-based, non-contact discharge measurements (e.g. the slope-area method) may provide an expected accuracy in the range of +/- 20%, the mean estimate accuracy potentially provided from remotely sensed information is certainly comparable. Although discharge estimates made from aerial or satellite sensed information will likely never provide the level of accuracy that can be achieved from direct in-stream measurment of depth, velocity and width (using the velocity-area method), there are numerous applications for remote discharge estimates. Where data gaps in flow records exist, and in rivers that have poor accessibility and costs for obtaining ground-based discharge are high, satellite and aerial paltforms can be used to supplement the ground-based network. In addition, because of the potential for global coverage by satellites, relatively frequent and accurate estimates of discharge over large areas can provide much needed understanding of the spatial distribution of discharge across the continents on a near-real-time basis.

The equations and methods developed here to estimate river discharge are easily adapted to other hydrologic modeling applications. This is because they are based on general principles of open-channel flow, and therefore can be generalized to any river environment. Used in combination, the hydraulic relationships can be used to estimate river discharge from a minimum set of observed channel and hdyraulic data; conversely, they can also be used to estimate river-channel geometry and discharge ratings from estimates of discharge. Thus, the equations and relationships developed to estimate discharge from measurable hydraulic variables can be used in watershed-modeling applications to generate realistic river-routing parameters based on channel geometry, including a criteria for identifying the occurrence of overbank flow (i.e. flooding).

Although the various estimating equations and relationships developed here pertain to inbank discharge and river flow conditions, the basic hydraulic relationships may also be adaptable to estimating discharge in overbank conditions, with adequate understanding of the variability of the discharge coefficient in these situations. Similar to the calibration of the equations for inbank discharge, the calibration of the equations for overbank discharge will require a large and diverse data set of overbank flow measurements along with information about the nature of the flooded areas. Another key issue with regard to estimating the discharge of overbank areas would be the identification of those flooded areas where the flow direction cannot be assumed to be in the downstream direction, and identification of stagnant or flooded areas that do not contribute to downstream flow. Even though the surface area of a flooded reach can be measured remotely, and an average width determined, all of this area may not be contributing to downstream flow. This problem can be addressed through observations of surface velocity, as non-contributing reaches can be readily identified as those areas with a minimum downstream surface velocity vector.

The current generation of spaced-based and aerial imagers and sensors are adequate to measure river hydraulic variables and thus to provide estimates of river discharge. However, much of the data that is potentially available for this prupose has not been developed to provide large spatial and temporal data sets for analysis. These data sets are critical to formulating improved calibrations and more complete understanding of the error characteristics of the discharge estimates. Development of a comprehensive data set that includes remote observations of water-surface area, stage, surface velocity, channel slope and observed discharge for a large number of river reaches is particularly important.

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APPENDIX RIVER DISCHARGE AND CHANNEL HYDRAULIC DATA

Appendix 1 - Channel Control Flow Measurement Data

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Table A1 - Channel Control Flow Measurement Data

River Name	Source USGS-NWIS USGS-NWIS	(m ³ /s)	(m)	(m)	(m)	(m)	(200)		
Vississippi at Thebes			W	Ŷ	v	Wm	(m) Ym	(m/m) Sa	F
	USGS-NWIS	27576.4			2.45				
		24547							
	USGS-NWIS	19762.2	894.8		1.99			0.000137	
	USGS-NWIS	14496						0.000137	
	USGS-NWIS	12542.5						0.000137	0.14
	USGS-NWIS	10504							
	USGS-NWIS	9003.4						0.000137	
	USGS-NWIS	8210.6						0.000137	
	USGS-NWIS	7163.1							
		1							
	USGS-NWIS	6398.6				1003			
	USGS-NWIS	5832.4						0.000137	0.144
	USGS-NWIS	5322.8			1.03				
	USGS-NWIS	4728.2			1.14				
	USGS-NWIS	4416.8			1.07			0.000137	
	USGS-NWIS	4246.9	573	6.31	1.18	1003	11.2	0.000137	0.1500
	USGS-NWIS	3822.2	570.9	6.02	1.12	1003	11.2	0.000137	0.1458
	USGS-NWIS	3652.3	572.1	5.81	1.1	1003	11.2	0.000137	0.145
	USGS-NWIS	3171	566	5.79	0.97			0.000137	
	USGS-NWIS	2641.6			0.92				
	USGS-NWIS	2216.9							
Potomac at point of rocks	USGS-NWIS	8154			1.82				0.189
	USGS-NWIS	1687.4							0.220
	USGS-NWIS	860.7			1.1				0.220
	USGS-NWIS	543.6			0.9				0.202
	USGS-NWIS	393.5			0.84				0.210
		328.4							
	USGS-NWIS	1			0.8				0.2020
	USGS-NWIS	264.4			0.66				0.176
	USGS-NWIS	213.5			0.71	477			0.217
	USGS-NWIS	166.8			0.55				0.168
	USGS-NWIS	140.7			0.54				0.167
	USGS-NWIS	105.9			0.45		9.4		0.151
	USGS-NWIS	91.7	288.6	0.82	0.39	477	9.4	0.00027	0.137
	USGS-NWIS	70.8	253.3	0.6	0.46	477	9.4	0.00027	0.189
	USGS-NWIS	69.6	235.3	0.56	0.52	477	9.4	0.00027	0.221
	USGS-NWIS	67.1	248.4	0.66	0.4	477	9.4	0.00027	0.157
	USGS-NWIS	59.7	240.5	0.78	0.32	477	9.4	0.00027	0.115
	USGS-NWIS	54.9			0.44				0.193
	USGS-NWIS	48.7			0.46				0.212
	USGS-NWIS	34.8	235.9		0.36				
	USGS-NWIS	15.6			0.24				
Vissouri at Hermann	USGS-NWIS	14439.4	602.3		1.83				0.182
Alssouri at Hermann	USGS-NWIS	13816.5	599.8						
		1			2.31	602			0.2338
	USGS-NWIS	11268.4	737.6		1.77				
	USGS-NWIS	8635.3	585.8		1.56				0.162
	USGS-NWIS	8323.9	437.4		1.8				0.176
	USGS-NWIS	7304.6	435.8		1.72				0.175
	USGS-NWIS	6993.2	432.8		1.74				0.182
	USGS-NWIS	6002.3	434.6		1.55				0.165
	USGS-NWIS	4954.7	424.9		1.4				0.154
	USGS-NWIS	4360.1	425.5	7.86	1.3	602	10.2	0.00013	0.148
	USGS-NWIS	3822.2	404.1	7.03	1.33	602	10.2	0.00013	0.160
	USGS-NWIS	3454.1	428.5		1.2				0.147
	USGS-NWIS	3001.1	424		1.09				0.136
	USGS-NWIS	2681.2							0.146
	USGS-NWIS	2406.6	384.6		1.05				
	USGS-NWIS	2143.3			0,98				0.137
	USGS-NWIS	1865.8	303		1.11				0.150
	USGS-NWIS	1582.7	332.2						0.150
					0.98				
	USGS-NWIS	1274.1	324.6		1.01	602			0.163
	USGS-NWIS	784.3	237.1	3.66	0.96				0.160
/ukon at Stevens Village	USGS-NWIS	17836.9	597.4		2.35				
	USGS-NWIS	16647.8	698	12.39	1.93		12.4	0.000068	0.175
	USGS-NWIS	13165.3	644.6	11.1	1.84	698	12.4	0.000068	0.176
	USGS-NWIS	10617.2	548.6		1.81				
	USGS-NWIS	8890.1	527.3		1.7			0.000068	
	USGS-NWIS	7644.4	676.6		1.52				
	USGS-NWIS	6540.2	655.3		1.41	698	12.4		
	USGS-NWIS	5945.6	655.3 544		1.41			0.000068	

		Discharge		Mean Depth	Mean Velocity	Width	Maximum Depth		Froude Number
River Name	Source	(m ³ /s) Q	(m) W	(m) Y	(m) V	(m) Wm	(m) Ym	(m/m) Sa	F
	USGS-NWIS	4898.1	505.9	7.44	1.3	698	12.4	0.000068	0.1522
	USGS-NWIS	4530	509	7.08	1.26	698	12.4	0.000068	0.1512
	USGS-NWIS	4161.9			1.2	698			
	USGS-NWIS	3624			1.1	698	12.4		
	USGS-NWIS	3284.3		5.27	1.09	698			
	USGS-NWIS	2916.2			1.05	698			
	USGS-NWIS	1056.1				698			
	USGS-NWIS	753.1				698			
	USGS-NWIS	523.8				698			
	USGS-NWIS	472.8			0.57	698			
	USGS-NWIS	416.2			0.41	698			
Milliomette et Spiem	USGS-NWIS	8188			2.15	517			0.2547
Willamette at Salem		6936.6				517			
	USGS-NWIS	1			1.98				
	USGS-NWIS	4869.8			2.17	517			0.3147
	USGS-NWIS	4331.8				517			
	USGS-NWIS	3793.9			2.02	517			
	USGS-NWIS	3312.6		7.58	1.81	517			
	USGS-NWIS	3029.4		6.62	1.84	517			0.2284
	USGS-NWIS	2117.8	212.1	5.91	1.69	517			0.2220
	USGS-NWIS	1922.4	206	5.46	1.71	517	7.3	0.00032	0.233
	USGS-NWIS	1613.8	213.3	5.22	1.45	517	7.3	0.00032	0.202
	USGS-NWIS	1373.2			1.37	517			0.197
	USGS-NWIS	1087.2			1.23	517			0.188
	USGS-NWIS	713.5			0.98	517			0.165
	USGS-NWIS	552.1			1.23	517			
	USGS-NWIS	512.5			1.2	517			
	USGS-NWIS	402		2.08	1.07	517			
		1							
	USGS-NWIS	353.9	182.3	1.93	1.01	517			
	USGS-NWIS	267.6		1.77	0.87	517			
	USGS-NWIS	199.9		1.57	0.75	517			0.191
	USGS-NWIS	163.1	120.4	3.5	0.39	517			
ted River of the North at Grand	USGS-NWIS	2972.8	304.8	7.92		351			
	USGS-NWIS	2457.5	350.5	7.02	1	351	7.9	0.000043	0.120
	USGS-NWIS	1797.8	182.9	7.62	1.29	351	7.9	0.000043	0.149
	USGS-NWIS	1545.9	184.4	7.15	1.17	351	7.9	0.000043	0.139
	USGS-NWIS	1285.4	182.9	6.04	1.16	351	7.9	0.000043	0.150
	USGS-NWIS	982.4		6.26	1	351	7.9	0.000043	0.127
	USGS-NWIS	724.8	130.4	5.58	1	351			
	USGS-NWIS	560.6	110.3	5.27	0.93	351			
	USGS-NWIS	526.6	91.4	5.88	0.98	351			
	USGS-NWIS	501.1	90.5	5.52	0.00	351			
	USGS-NWIS	472.8	103.6	5.32	0.86	351			
	USGS-NWIS								
		421.9	88.4	5.17	0.92	351			
	USGS-NWIS	359.6	85.3	4.69	0.9	351			
	USGS-NWIS	302.9	84.1	4.15	0.87	351			
	USGS-NWIS	273.5	78	4.14	0.85	351			
	USGS-NWIS	222.5	77.4	3.53	0.81	351			
	USGS-NWIS	193.9	76.5		0.73	351			
	USGS-NWIS	128.3	76.2		0.53	351		0.000043	
	USGS-NWIS	98.8	74.4		0.43	351		0.000043	
	USGS-NWIS	44.5	71.3	3.33	0.19	351	7.9	0.000043	
rkansas River at Arkansas Cit	USGS-NWIS	2265	278.6	5.14	1.58	285	5.1	0.000685	0.222
	USGS-NWIS	2140.4	285.6	4.94		285			
	USGS-NWIS	1874.3	285.3	4.27	1.54	285			
	USGS-NWIS	1537.4	273.1	3.74		285			
	USGS-NWIS	1291.1	271.3	3.63	1.3	285			
	USGS-NWIS	971.1	266.1	2.84		285			
	USGS-NWIS	719.1	105.2		1.11	285			
	USGS-NWIS	475.7	179.8	2.41	1.1	285			
	USGS-NWIS	404.9	162.5	2.22	1.12	285			
	USGS-NWIS	342.6	139.3	4.3	0.57	285		0.000685	
	USGS-NWIS	305.8	153	1.96	1.02	285			
	USGS-NWIS	263.9	123.7	4.7	0.45	285	5.1		
	USGS-NWIS	246	146	1.77	0.95	285	5.1	0.000685	0.228
	USGS-NWIS	219.4	106.7	4.08	0.51	285			
	USGS-NWIS	174.4	149.3	1.44	0.81	285			
	USGS-NWIS	139	193.5	1.02	0.9	285			
	USGS-NWIS	111.8	150.6	1.12	0.66	285			
	USGS-NWIS	74.7	135.6	0.75	0.73	285 285		0.000685	
	USGS-NWIS	54.9	150.3	0.63	0.58			0.000685	0.233

		Discharge	Width	Mean Depth	Mean Velocity	Maximum Width	Maximum Depth	Slope	Froude Number
River Name	Source	(m ³ /s) Q	(m) W	(m) Y	(m) V	(m) Wm	(m) Ym	(m/m) Sa	F
	USGS-NWIS	39.4			0.48	285			0.19160
Kuskokwim at Crooked Creek	USGS-NWIS	6189.1	408.4		1.94				
	USGS-NWIS	4473.4			1.82	442		0.000198	
	USGS-NWIS	3595.7	426.7	5.44	1.55	442	7.8		0.2122
	USGS-NWIS	3001.1	441.9	4.77	1.42	442	7.8	0.000198	0.207
	USGS-NWIS	2641.6	389.5	5.01	1.35	442	7.8	0.000198	0.1926
	USGS-NWIS	2347.1			1.3	442			
	USGS-NWIS	2081			1.24				0.189
	USGS-NWIS	1823.3			1.12				
	USGS-NWIS	1528.9			1.01	442			
	USGS-NWIS USGS-NWIS	1449.6			1.05 0.97	442 442			
	USGS-NWIS	1404.3			0.97	442			0.1330
	USGS-NWIS	1217.4			. 0.9	442			
	USGS-NWIS	1078.7			0.85	442			0.1378
	USGS-NWIS	730.5			0.44	442			
	USGS-NWIS	560.6			0.35	442			
	USGS-NWIS	515.3			0.49	442		0.000198	0.0923
	USGS-NWIS	498.3			0.47	442			
	USGS-NWIS	376.6			0.33	442			
Platte near Agency	USGS-NWIS	965.5			1.35	156			
	USGS-NWIS	707.8			1.54	156			
	USGS-NWIS	622.9			1.37	156			0.238
	USGS-NWIS	404.9			1.32	156			
	USGS-NWIS USGS-NWIS	396.4			1.18	156			
	USGS-NWIS	390.7			0.95 1.1	156			
	USGS-NWIS	314.3			1.23	150			
	USGS-NWIS	222.8			0.9	156			0.127
	USGS-NWIS	119.2			0.7	156			
	USGS-NWIS	82.7			0.62	156			
	USGS-NWIS	54.4	44.5	2.78	0.44	156	4.6	0.00046	0.0842
	USGS-NWIS	45.6	41.8	2.38	0.46	156	4.6	0.00046	0.0952
	USGS-NWIS	39.1	39.6		0.66	156			
	USGS-NWIS	30.9	43.3		0.59	156			
	USGS-NWIS	23.4	41.5		0.52	156			0.1583
	USGS-NWIS	19.5			0.14	156			
	USGS-NWIS USGS-NWIS	12.9			0.13 0.36	156 156			
	USGS-NWIS	3.6			0.38	156			
Sagavanirktok near Pump Stati		478.5	233.2		1.76	233			0.5197
ougaraim kok noar r emp etan	USGS-NWIS	410.5		1.6	1.96	233			
	USGS-NWIS	387.9	132.9		1.91	233			
	USGS-NWIS	342.6	129.5		1.72			0.00274	0.4441
	USGS-NWIS	328.4	131.1	1.52	1.65	233	1.6	0.00274	0.4275
	USGS-NWIS	305.8	129.5		1.67	233			
	USGS-NWIS	239.5	132		1.26	233			0.335
	USGS-NWIS	201.6	128.3		1.2	233			
	USGS-NWIS	191.1	126.2		1.26	233			0.3659
	USGS-NWIS	160.5	125.9		1.22	233			0.3803
	USGS-NWIS USGS-NWIS	113,8 88.9	102.1 83.8	0.92	1.21 0.97	233 233			0.4029
	USGS-NWIS	60.3	96.9		0.97				0.2954
	USGS-NWIS	52.9	90.9		0.69	233			0.2464
	USGS-NWIS	39.4	86.9		0.5	233			0.1674
	USGS-NWIS	18	131.1		0.54	233			
	USGS-NWIS	12.5	91.7		0.39	233			
	USGS-NWIS	9.2	81.7	0.39	0.29	233			0.1483
	USGS-NWIS	6.9	67.1	0.23	0.44	233			0.2930
	USGS-NWIS	4.6	67.1	0.32	0.22	233			0.1242
	USGS-NWIS	2349.9	337.7		1.71	338			0.2710
	USGS-NWIS	1641.8	320		1.2				
	USGS-NWIS	1308	332.2		1.21	338			0.2147
	USGS-NWIS	908.8	315.5		1.14				
	USGS-NWIS	622.9	213.3		1.1	338			
	USGS-NWIS USGS-NWIS	458.7 331.3	206.6 147.8	2.21 3.2	1 0.7	338 338			0.2148 0.1
	0000"199910	. 331.3	141.0	<u>ی</u> .د	0.7	330	÷	0.00049	0.1
			105 4	1 64	0.05	220	A 9	0 00040	0.2360
1	USGS-NWIS USGS-NWIS	305.8 262.2	195.4 131.1	1.64 2.59	0.95 0.77	338 338			

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		Discharge		Mean Depth	Mean Velocity	Width	Maximum Depth		Froude Number
River Name	Source	(m³/s) Q	(m) W	(m) Y	(m) V	(m) Wm	(m) Ym	(m/m) Sa	F
	USGS-NWIS	219,7		2.63	0.7	338			0.1378
	USGS-NWIS	190.3	178.3	1.2	0.89	338	4.3	0.00049	0.2595
	USGS-NWIS	132.5	174.6	1.01	0.76	338	4.3	0.00049	0.2415
	USGS-NWIS	103.1	171.3	0.86	0.7	338	4.3	0.00049	0.2411
	USGS-NWIS	75.6	78.3	2.09	0.46	338	4.3	0.00049	0.1016
	USGS-NWIS	47	104.5	0.75	0.6	- 338	4.3	0.00049	0.2213
	USGS-NWIS	35.7	92.7	0.63	0.61	338	4.3	0.00049	0.2454
	USGS-NWIS	19.3	64.3	0.55	0.55	338	4.3	0.00049	0.2369
	USGS-NWIS	12.9	59,4	0.44	0.49	338	4.3	0.00049	0.235
Kobuk at Kiana	USGS-NWIS	4331.8	528.8	6.34	1.29	533	6.3	0.0008	0.1636
	USGS-NWIS	3765.6	533.4	6.25	1.13	533	6.3	0.0008	0.1443
	USGS-NWIS	3199.3	509	5.53	1.16	533	6.3	0.0008	0.1575
	USGS-NWIS	3086.1	475.5	5.39	1.2	533	6.3	0.00008	0.165
	USGS-NWIS	2556.6	475.5	4.77	1.13	533	6.3	0.00008	0.1652
	USGS-NWIS	2197.1	472.4	4.64	1	533	6.3	0.0008	0.1482
	USGS-NWIS	1865.8	437.4	4.14	1.03	533	6.3	0.0008	0.1617
	USGS-NWIS	1628	463.3	4.93	0.71	533	6.3	0.0008	0.1021
	USGS-NWIS	1327.9	408.4	3.68	0.88	533	6.3	0.00008	0.1465
	USGS-NWIS	968.3	344.4	3.67	0.77	533			0.1283
	USGS-NWIS	739	304.8	3.81	0.64	533	6.3	0.00008	
	USGS-NWIS	608.7	283.5	3.57	0.6	533	6.3	0.00008	0.1014
	USGS-NWIS	526.6	274.3	3.66	0.52	533	6.3	0.00008	0.0868
	USGS-NWIS	489.8	274.3	3.66	0.49	533	6.3	0.00008	0.0818
	USGS-NWIS	410.5	265.2		0.44	533	6.3	0.0008	0.075:
	USGS-NWIS	342.6	259.1	3.18	0.42	533	6.3	0.00008	0.0752
	USGS-NWIS	252.3	246.9	2.69	0.38	533	6.3	0.00008	0.0740
	USGS-NWIS	144.7	231.6	2.76	0.23	533	6.3	0.00008	0.0442
	USGS-NWIS	59.2	222.5	1.75	0.15	533	6.3	0.00008	0.0362
lissouri nr Culbertson	USGS-NWIS	1155.2	205.7	3.71	1.51	206	3.7	0.000156	0.2504
	USGS-NWIS	1092.9	202.7	3.14	1.72	206	3.7	0.000156	0.3100
	USGS-NWIS	761.6	204.2	3.62	1.03	206	i 3.7	0.000156	0.17
	USGS-NWIS	566.3	205.7	3	0.92	206	3.7	0.000156	0.1690
	USGS-NWIS	461.5	207.3	2.58	0.86	206	3.7	0.000156	0.171
	USGS-NWIS	396.4	198.1	2.22	0,9	206	3.7	0.000156	0.1929
	USGS-NWIS	393.5	193.5	2.52	0.81	206	3.7	0.000156	0.1629
	USGS-NWIS	365.2	192.9	2.63	0.72	206	3.7	0.000156	0.1418
	USGS-NWIS	328.4	175,9	2.4	0.78	206	3.7	0.000156	0.1608
	USGS-NWIS	297.3	192	2.23	0.69	206	3.7	0.000156	0.147
	USGS-NWIS	275.5	160	2.43	0.71	206	3.7	0.000156	0.145
	USGS-NWIS	239	150.9	2.28	0.69	206	3.7	0.000156	0.145
	USGS-NWIS	213.2	164.6	2.03	0.64	206	3.7	0.000156	0.143
	USGS-NWIS	175.3	157	1.88	0.59	206	3.7	0.000156	0.1374
	USGS-NWIS	145.8	181.3	1.28	0.63	206	3.7	0.000156	0.177
	USGS-NWIS	144.1	181.3	1.56	0.51	206	3.7	0.000156	0.130
	USGS-NWIS	138.4			0.57	206			
	USGS-NWIS	116.4			0.64	206			
	USGS-NWIS	109.3			0.42	206			
	USGS-NWIS	98.2			0.53	206			
S. Platte near Kersey	USGS-NWIS	436			1.17	203			0.270
•	USGS-NWIS	402	202.7		1.16				0.282
	USGS-NWIS	207.8	139.6		0.98	203	1.9	0.00093	0.253
	USGS-NWIS	139.6			0.82				0.227
	USGS-NWIS	72.8			0.73				0.239
	USGS-NWIS	36.5			0.82				0.335
	USGS-NWIS	33.1	72.5		0.77	203			0.317
	USGS-NWIS	32.3	68.9	0.64	0.73	203	1.9	0.00093	0.291
	USGS-NWIS	28.3			0.71	203			0.290
	USGS-NWIS	27.6			0.71	203			0.303
	USGS-NWIS	24.4			0.7	203			0.319
	USGS-NWIS	22		0.5	0.66	203			0.298
	USGS-NWIS	19.6	67.7		0.64	203			0.304
	USGS-NWIS	15.7	57.6		0.65	203			0.320
	USGS-NWIS	14.4	48.5		0.64	203			0.298
	USGS-NWIS	13.9	57		0.61	203			0.308
	USGS-NWIS	13.5			0.64	203			0.308
	USGS-NWIS	12.2	49.7	0.41	0.6	203			0.299
	USGS-NWIS	10	45.1	0.36	0.61	203			0.324
	USGS-NWIS	458.3	73.1	2.9	2.16	74			0.405
nena near two www.s									
Chena near Two Rivers	USGS-NWIS	311.4	73.8	2.2	1.92	74		0.00136	0.413

		Discharge		Mean Depth	Mean Velocity	Width	Maximum Depth		Froude Number
River Name	Source	(m³/s) Q	(m) W	(m) Y	(m) V	(m) Wm	(m) Ym	(m/m) Sa	F
	USGS-NWIS	246			1.72				
	USGS-NWIS	195.1	70.1	1.72	1.62	74	2.9	0.00136	0.3945
	USGS-NWIS	171.3	65.5	5 1.91	1.37	74	2.9	0.00136	0.3166
	USGS-NWIS	143							0.3995
	USGS-NWIS	95,1				74			0.3058
	USGS-NWIS	80.1							0.3028
	USGS-NWIS	51.5				74			0.2640
	USGS-NWIS	45.3							0.2982
	USGS-NWIS	36.5			0.66				0.2314
	USGS-NWIS	32.8							0.3410
	USGS-NWIS USGS-NWIS	31.7							0.1321
	USGS-NWIS	30 25.3			0.69 0.62	74 74			0.2755
	USGS-NWIS	23.5							0.1552
	USGS-NWIS	19.1			0.96				0.5259
	USGS-NWIS	15.7			0.30	74			0.1938
elaware at Callicoon	USGS-NWIS	1851.6			1.86				0.3222
Selaware at Ballicooli	USGS-NWIS	767.3			1.82				
	USGS-NWIS	696.5							0.4450
	USGS-NWIS	464.3			1.52				0.3649
	USGS-NWIS	390.7							
	USGS-NWIS	359.6			1.48				0.395
	USGS-NWIS	314.3			1.4	290			0.3863
	USGS-NWIS	305.8	163.7	1.32	1.33	290	3.4	0.00107	0.369
	USGS-NWIS	268.4	162.8	1.24	1.33	290	3.4	0.00107	0.381
	USGS-NWIS	217.2	156	6 1.18	1.12	290	3.4	0.00107	0.3293
	USGS-NWIS	188.6	146.9	0.93	1.32	290	3.4	0.00107	0.43
	USGS-NWIS	166.5			1.13				0.3723
	USGS-NWIS	135.3			1.15				0.4299
	USGS-NWIS	109.6			1.05	290			0.4128
	USGS-NWIS	79.6			0.79				0.313
	USGS-NWIS	52.4			0.44				0.175
	USGS-NWIS	32.3							0.137
	USGS-NWIS	26.8			0.36	290			0.194
	USGS-NWIS	19.8			0.33				0,183
	USGS-NWIS	12.4			0.54				0.385
(ansas at DeSoto	USGS-NWIS	4784.8			2.4				
	USGS-NWIS USGS-NWIS	4473.4 3624	217 187.4		2.65 2.3				0.303
	USGS-NWIS	3114.4							
	USGS-NWIS	2944.5			2.13				
	USGS-NWIS	1990.4			1.6				
	USGS-NWIS	1823.3			1.68				
	USGS-NWIS	1625.1	178		1.00	206			
	USGS-NWIS	1415.6			1.53	206			
	USGS-NWIS	1192			1.38				
	USGS-NWIS	971.1	170.7		1.31	206			
	USGS-NWIS	671	172.2		1.05	206	9.7	7 0.00035	0.174
	USGS-NWIS	535.1	170.7		1.01	206			0.183
	USGS-NWIS	492.6	171.9	2.8	1.02	206	9.7	7 0.00035	0.194
	USGS-NWIS	419			1.11				
	USGS-NWIS	353.9		2.14	1.01	206			0.220
	USGS-NWIS	294.5			0.91	206			0.208
	USGS-NWIS	213.8							
	USGS-NWIS	106.2			0.58				0.168
	USGS-NWIS	50.4			0.45				0.162
leuse near Clayton	USGS-NWIS	648.4							
	USGS-NWIS	489.8			1.26				
	USGS-NWIS	461.5			1.14				
	USGS-NWIS	370.9			1.18				
	USGS-NWIS	269.5			1.06				
	USGS-NWIS	211.8	72.5		1.03				
	USGS-NWIS	190	71.6		0.94				0.179
	USGS-NWIS	160.5	53		0.99				0.181
	USGS-NWIS	140.4	51.8		1.2				
	USGS-NWIS	107.6							
	USGS-NWIS	71.6			0.76			5 0.00028	
	USGS-NWIS	49.8							
	USGS-NWIS	36.5			0.59			5 0.00028	
	USGS-NWIS	32.8	41.1	1.17	0.68	96	5	5 0.00028	0.2008

		Discharge		Mean Depth	Mean Velocity	Width	Maximum Depth		Froude Number
liver Name	Source	(m ³ /s) Q	(m) W	(m) Y	(m) V	(m) Wm	(m) Ym	(m/m) Sa	F
	USGS-NWIS	22.2		0.97	0.56	96	5		0.1816
	USGS-NWIS	17.3	43	0.77	0.52	96	5	0.00028	0.1892
	USGS-NWIS	15.9	42.4	0.84	0.45	96	5	0.00028	0.1568
	USGS-NWIS	13.1	40.8	0.73	0.44	96	5	0.00028	0.1645
	USGS-NWIS	10.4	38.1	0.67	0.41	96	5	0.00028	0.1600
	USGS-NWIS	7.6	41.1	0.62	0.3	96	5	0.00028	0.1217
Red R. of the North at Fargo	USGS-NWIS	792.8	204.2	4.64	0.84	204	4.8	0.00009	0.1245
	USGS-NWIS	724.8	193.5	4.78	0.78	204	4.8	0.00009	0.1139
	USGS-NWIS	702.2	189	4.65	0.8	204	4.8	0.00009	0.118
	USGS-NWIS	654	180.4	4.26	0.85	204	4.8	0.00009	0.1315
	USGS-NWIS	557.8	167	4.34	0.77	204	4.8	0.00009	0.118
	USGS-NWIS	532.3	178.3	4.08	0.73	204	4.8	0.00009	0.115
	USGS-NWIS	506.8	176.8	3.87	0.74	204	4.8	0.00009	0.120
	USGS-NWIS	447.3	164.6	3.6	0.76	204	4.8	0.00009	0.1271
	USGS-NWIS	342.6	152.7	3.41	0.66	204	4.8	0.00009	0.11
	USGS-NWIS	308.6	142.6	3.1	0.72	204	4.8	0.00009	0.130
	USGS-NWIS	278	109.7	3.67	0.69	204	4.8	0.00009	0.115
	USGS-NWIS	250	108.2	3.35	0.69	204	4.8	0.00009	0.120-
	USGS-NWIS	200.7	91.4	3.23	0.68	204	4.8	0.00009	0.120
	USGS-NWIS	185.7	85.3		0.67	204	4.8	0.00009	0.118
	USGS-NWIS	163.1	76.2	3.75	0.57	204	4.8	0.00009	0.094
	USGS-NWIS	132.5	73.1	3.53	0.51	204	4.8	0.00009	0,08
	USGS-NWIS	101.9	44.2	3.47	0.66	204	4.8	0.00009	0.113
	USGS-NWIS	74.7	39.6	3.21	0.59	204	4.8	0,00009	0.105
	USGS-NWIS	45.6	43.9	3.01	0.34	204	4.8	0.00009	0.062
	USGS-NWIS	35.4	37.8	2.24	0.42	204	4.8	0.00009	0.089
aco at Cornish	USGS-NWIS	744.6			1.69	81	5.6	0.0006	0.228
	USGS-NWIS	676.7	80.8	5.35	1.57	81	5.6	0.0006	0.216
	USGS-NWIS	543,6	80.8	4.69	1.44	81	5.6	0.0006	0.212
	USGS-NWIS	458.7	77.7		1.34	81	5.6		
	USGS-NWIS	447.3			1.32	81	5.6		
	USGS-NWIS	407.7			1.26	81	5.6		
	USGS-NWIS	351.1	76.2		1.15	81	5.6	0.0006	
	USGS-NWIS	314.3			1.1	81	5.6		
	USGS-NWIS	308.6			1.56	81	5.6	0.0006	
	USGS-NWIS	266.4	62.5		1.43	81	5.6	0.0006	
	USGS-NWIS	250.8	76.5	3.5	0.94	81	5.6	0.0006	0.160
	USGS-NWIS	217.4	77.1			81	5.6		
	USGS-NWIS	176.7	75.9		0.74	81	5.6		
	USGS-NWIS	163.6			0.7	81	5.6		
	USGS-NWIS	135.6			0.62	81	5.6		
	USGS-NWIS	103.3			0.82	81	5.6		
	USGS-NWIS	77.3			0.39	81	5.6		
	USGS-NWIS	49.3			0.44	81	5.6		0.120
	USGS-NWIS	42.5		2.28	0.25		5.6		
	USGS-NWIS	36	82.3		0.42	81	5.6		
acramento near Red Bluff	USGS-NWIS	3708.9	377.9		1.87	378			
	USGS-NWIS	3397.5	367.3		2.02	378			
	USGS-NWIS	3001.1	352		1.96	378			
	USGS-NWIS	2757.6	213.3		2.22	378		2 0.000575	
	USGS-NWIS	2406.6	210.3		2.09	378		2 0.000575	
	USGS-NWIS	2253.7	229.5		2.04			2 0.000575	
	USGS-NWIS	1970.6	204.2		1.94			0.000575	
	USGS-NWIS	1653.5	189		1.92	378		0.000575	
	USGS-NWIS	1376	166.1	4.42	1.87	378		0.000575	
	USGS-NWIS	1090	155.4		1.79	378		0.000575	
	USGS-NWIS	787.1	118.9		1.68	378		0.000575	
	USGS-NWIS	504	117		1.48	378		0.000575	
	USGS-NWIS	450.2	115.8		1.51	378			
	USGS-NWIS	376.6	115.8	2.39	1.36	378		2 0.000575	
	USGS-NWIS	331.3	114.3	2.03	1.4	378		2 0.000575	
	USGS-NWIS	267.6	112.5		1.52	378		2 0.000575	
	USGS-NWIS	207.0	112.5	1.50	1.52	378		2 0.000575	
	USGS-NWIS	165.9	109.7	1.13	1.34	378		2 0.000575	
	USGS-NWIS	138.2		0.94	1.34	378		2 0.000575	
			106.4						
Line and have - at 18/au and -	USGS-NWIS	107	105.2	0.83	1.22	378		2 0.000575	
Susquehanna at Waverley	USGS-NWIS	2514.2	301.1	4.6	1.89	301	4.6		
	USGS-NWIS USGS-NWIS	2301.8	260 301.1	3.97 4.01	2.23 1.68	301 301	4.6 4.6		0.357
		2027.2							

Diver News-	Source	Discharge		Mean Depth	Mean Velocity	Width	Depth	•	Froude Number
River Name	Source	(m ³ /s) Q	(m) W	(m) Y	(m) V	(m) Wm	(m) Ym	(m/m) Sa	F
	USGS-NWIS	1053.2			1.61	301	4.6		
	USGS-NWIS	789.9			1.47		4.6		
	USGS-NWIS	719.1			1.5		4.6		
	USGS-NWIS	656.9			1.49		4.6		0.3166
	USGS-NWIS USGS-NWIS	583.2 532.3			1.23 1.13		4.6 4.6		0.2299
	USGS-NWIS	396.4			0.89		4.6		
	USGS-NWIS	356.7			1.23		4.6		
	USGS-NWIS	242.4			1.04		4.6		0.2848
	USGS-NWIS	194.2	172.2	! 1.2	0.95	301	4.6	0.00048	0.2770
	USGS-NWIS	140.1	172.2	2 0.98	0.84		4.6		
	USGS-NWIS	77.3			0.64	301	4.6		
	USGS-NWIS	25.4			0.52				
anana near Fairbanks	USGS-NWIS	2420.7			1.75				0.3040
	USGS-NWIS USGS-NWIS	1922.4			1.59 1.73				0.1829
	USGS-NWIS	1305.2							0.3203
	USGS-NWIS	1039.1	277		1.53				
	USGS-NWIS	755.9			1.28				
4	USGS-NWIS	535.1	195.4	1.7	1.61	410	7.7	0.00043	0.3944
	USGS-NWIS	450.2	192.9	2.3	1.01				
	USGS-NWIS	390.7			1.56				0.3810
	USGS-NWIS	278.9			0.76				
	USGS-NWIS	271.2			0.98				0.191:
	USGS-NWIS USGS-NWIS	222.5 167.9			0.74 0.69				
	USGS-NWIS	158.8			0.84	410			
	USGS-NWIS	152.9			0.66				
	USGS-NWIS	147.5			0.66				
	USGS-NWIS	141	103.6		0.76				0.180
	USGS-NWIS	139.9	122.2	1.35	0.85	410	7.7	0.00043	0.233
	USGS-NWIS	137	118.9	1.2	0,96	410	7.7	0.00043	0.279
ukon at Eagle	USGS-NWIS	15062.3			2.97				0.293
	USGS-NWIS	10957	481.6		2.49				0.263
	USGS-NWIS	6483.6			2.07				
	USGS-NWIS	5294.5			1.94				
	USGS-NWIS USGS-NWIS	4756.5 4331.8			1.86 1.83				
	USGS-NWIS	3963.8	400.2		1.83				0.242
	USGS-NWIS	3510.8	417.6		1.65				
	USGS-NWIS	3227.6	396.2		1.56	488			0.217
	USGS-NWIS	2732.2			1.49	488			0.217
	USGS-NWIS	2531.1	414.5		1.64	488	10.4	0.00036	0.271
	USGS-NWIS	1896.9	368.8	4.31	1.19	488	10.4	0.00036	0.183
	USGS-NWIS	787.1	438.9		1.05	488			
	USGS-NWIS	705	449.6		0.71	488			
	USGS-NWIS	622.9	359.6		0.75				
	USGS-NWIS	506.8	344.4		0.67	488 488			
	USGS-NWIS	489.8 475.7	313.9		0.53				0.098
	USGS-NWIS USGS-NWIS	475.7	214.9 320		0.55 0.55				0.087
	USGS-NWIS	407.7			0.55				0.111
remper	Coon, 1998	2.4			0.73				0.433
ill	Coon, 1998	6.8			1.11				
	Coon, 1998	7			1.06				
	Coon, 1998	7.7			1.17				0.518
	Coon, 1998	8.9			1.2				
	Coon, 1998	10.1	13.3		1.25				
	Coon, 1998	16.9			1.52				
	Coon, 1998	29.4			1.93				0.635
	Coon, 1998	117	12.5		0.88				0.414
	Coon, 1998 Coon, 1998	11.7	13.9 13.9		1.19	16 16			
	Coon, 1998 Coon, 1998	14			1.21				0.455
	Coon, 1998	19.6	15.3		1.5				
	Coon, 1998	23.6	15.9		1.63				
loordenor	Coon, 1998	2.2			0.51	14			
ill	Coon, 1998	3.5			0.67				
	Coon, 1998	4	11.8		0.71	14			
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		Discharge		Mean Depth	Mean Velocity			Maximum Depth		Froude Number
River Name	Source	(m ³ /s) Q	(m) W	(m) Y	(m) V	(m) Wm		(m) Ym	(m/m) Sa	F
	Coon, 1998	9,4				97	14	0.8		
	Coon, 1998	10.6	5 1	4 0.8	1 0.	93	14	0.8	0.0015	0.33008
	Coon, 1998	11.6	5 14.	2 0.8	30.	98	14	0.8	0.0015	0.34361
Canisteo	Coon, 1998	4.1	9.	2 0.5	80.	77	12	1	0.00274	0.32297
River	Coon, 1998	5.8				1	12	1		0.40568
	Coon, 1998	7.4				12	12	1		0.43386
	Coon, 1998	12.8				51	12			
	Coon, 1998	13.8				53	12	1		0.53646
	Coon, 1998	14.3				49	12			0.5132
	Coon, 1998	14.5				63	12			0.5715
	Coon, 1998	14.6				54	12			0.5304
	Coon, 1998	14.6				51	12			
	Coon, 1998	14.8				48	12			0.50397
	Coon, 1998	16				1.6 61	12			0.54483
	Coon, 1998	16.3				61	12			
	Coon, 1998	12 6				1	12 12	1		
	Coon, 1998	13.8				1.5 1.7	12			
	Coon, 1998							1		
	Coon, 1998 Coon, 1998	17.9				55 61	12 12			0.51342
	Coon, 1998	19		0.8 2 0.9		64	12			0.5304
Mill	Coon, 1998	3.1				76	19	1.2		0.4103
Brook	Coon, 1998	4.8				81	19	1.2		
DIOON	Coon, 1998	5.7				1	19	1.2		
	Coon, 1998	6.1				03	19	1.2		
	Coon, 1998	22.9				86	19	1.2		
	Coon, 1998	48.7				44	19	1.2		
	Coon, 1998	70.8				11	19	1.2		
E.Branch	Coon, 1998	107.3				72	70	1.9		
Ausable	Coon, 1998	119.2				76	70	1.9		
River	Coon, 1998	161.9				92	70	1.9		
	Coon, 1998	178.1				96	70	1.9		
	Coon, 1998	248.9				18	70	1.9		0.53887
	Coon, 1998	305.8				35	70	1.9		0.55042
Beaver	Coon, 1998	16.3				64	68	2.6		
Kill	Coon, 1998	71.3				32	68	2.6		
	Coon, 1998	140.7				84	68	2.6		
	Coon, 1998	246.6				29	68	2.6		
	Coon, 1998	269.5				42	68	2.6		
	Coon, 1998	286				48	68	2.6		
	Coon, 1998	297.3				2.6	68	2.6		
	Coon, 1998	560.6				53	68	2.6		
	Coon, 1998	676.7	67.	7 2.6	1 3.	83	68	2.6	0.00451	0.75729
Tioughnioga	Coon, 1998	14.2	64.	3 0.5	з 0.	41	88	2.1	0.00118	0.17990
River	Coon, 1998	129.1	80.	8 1.2	91.	24	88	2.1	0.00118	0.3487
	Coon, 1998	153.5	i 8	2 1.3	91.	34	88	2.1	0.00118	0.36306
	Coon, 1998	159.7	82.	3 1.4	41.	34	88	2.1	0.00118	0.35670
	Coon, 1998	171.6	82.	9 1.4	91.	39	88	2.1	0.00118	0.3637
	Coon, 1998	182.9	83.	2 1.5	51.	42	88	2.1	0.00118	0.36434
	Coon, 1998	187.1	83.	5 1.5	41.	46	88	2.7	0.00118	0.375
	Coon, 1998	214.3	84.	4 1.6	7 1.	52	88	2.1	0.00118	0.37572
	Coon, 1998	281.4	86.	6 1.9	1 .	.7	88	2.1	0.00118	0.39293
	Coon, 1998	286	86.	9 1.9	3 [,]	1.7	88	2.1	0.00118	0.3908
	Coon, 1998	305.8	87.	2 1.9	91.	76	88	2.1	0.00118	0.3985
	Coon, 1998	308.6	87.	2	21.	77	88	2.1	0.00118	0.3998
	Coon, 1998	322.8	87.	5 2.0	5 -	1.8	88	2.1	0.00118	0.401
Kayderasasa	Coon, 1998	24.8	2	4 1.1	30.	91	30	1.4		
Creek	Coon, 1998	27	24.			95	30	1.4		0.2817
	Coon, 1998	28.6	24.	8 1.1	90,	97	30	1.4		0.2840
	Coon, 1998	29.7				98	30	1.4		0.2857
	Coon, 1998	30				96	30	1.4		
	Coon, 1998	30.3				98	30	1.4		
	Coon, 1998	31.4				1	30	1.4		
	Coon, 1998	48.1	30.	2 1.3	71.	16	30	1.4		
Indian	Coon, 1998	2.8			30.	47	19	0.8		0.2289
River	Coon, 1998	3.7				61	19	0.8		0.2937
	Coon, 1998	5.5	14.	5 0.4	90.	77	19	0.8	3 0.01217	0.3513
	Coon, 1998	6		4 0.4	9 D.	86	19	.0.	3 0.01217	0.3924
	Coon, 1998	8.4	16.	5 0.5	20.	98	19	0.8	3 0.01217	0.4341
	Coon, 1998			6 0.5		06	19	0.8		

		n	ما د (م (A	Mean	Mean		Maximum	Siope	Froude
	0	Discharge		Depth	Velocity	Width	Depth	(ma lac)	Number
River Name	Source	(m ³ /s) Q	(m) W	(m) Y	(m) V	(m) VVm	(m) Ym	(m/m) Sa	F
	Coon, 1998	10.2				19			
	Coon, 1998	12.8				19			
	Coon, 1998	18.1	18.3	0.7	1.4	19	0.8	0.01217	0.53452
	Coon, 1998	20.3	18.6	6 0.74	1.48	19	0.8	0.01217	0.54958
	Coon, 1998	22.5	18.8	0.78	1.54	19	0.8	0.01217	0.55700
Sacandaga	Coon, 1998	109.6	83.2	1.47	0.89	89			
River	Coon, 1998	112.4			0.9				
	Coon, 1998	116.9				89			
	Coon, 1998	119.5							
	Coon, 1998	376.6				89			
Esopus	Coon, 1998	63.4				67			
Creek	Coon, 1998	156.3				67			
	Coon, 1998	173.8				67			
	Coon, 1998	246.3			2.23	67			
	Coon, 1998	255.7				67 67			
	Coon, 1998 Coon, 1998	1058.9							
	Coon, 1998	1463.8				67			
E.B. Delaware	Coon, 1998	40.2				39			
River	Coon, 1998	52.1				39			
i vivot	Coon, 1998	56.3				39			
	Coon, 1998	59.5				39			
	Coon, 1998	81				39			
	Coon, 1998	186.9				39			
Oueleot	Coon, 1998	27.4				28			
River	Coon, 1998	30				28			
	Coon, 1998	31.1	24	0.86	1.51	28		0.00836	0.5201
	Coon, 1998	33.7	24.8	0.91	1.5	28	1.1	0.00836	0.5022
	Coon, 1998	40.2	25.9	0.95	1.64	28	1.1	0.00836	0.5374
	Coon, 1998	41.1	26.3	0.96	1.62	28	1.1	0.00836	0.5281
	Coon, 1998	44.2	27	' 1.01	1.63	28	1.1	0.00836	0.51
	Coon, 1998	47				28			
	Coon, 1998	47.6				28			
	Coon, 1998	50.4				28			
	Coon, 1998	53.2				28			
	Coon, 1998	24.8			1.33	28			
	Coon, 1998	29.7							
	Coon, 1998	36.5				28			
	Coon, 1998	45.6				28			
	Coon, 1998	45.9 49.5				28 28			
Superiorbonno	Coon, 1998 Coon, 1998	100.2				20 60			
Susquenhanna River	Coon, 1998	118.9				60			
i dvel	Coon, 1998	174.4				60			
	Coon, 1998	194.5							
	Coon, 1998	257.6				60			
	Coon, 1998	294.5				60			
	Coon, 1998	404.9				60			
	Coon, 1998	537.9				60			
	Coon, 1998	105.3	57.9	1.88	0.97	60	2.3	3 0.00081	0.2259
	Coon, 1998	119.2							0.2344
	Coon, 1998	122.3	58.5	5 1.97	1.06	60	2.3	3 0.00081	0.2412
	Coon, 1998	126				60			
	Coon, 1998	166.5							
Unadilla	Coon, 1998	40.5							
River	Coon, 1998	46.7							
	Coon, 1998	51							
	Coon, 1998	58.9				49			
	Coon, 1998	63.4				49			
	Coon, 1998	68.8							
	Coon, 1998	81.3							
	Coon, 1998	114.4							
	Coon, 1998	117.5							
	Coon, 1998	129.7				49			
	Coon, 1998	131.9				49			
	Coon, 1998	174.7							
	Coon, 1998	179.2				49			
	Coon, 1998	180.4				49			
	Coon, 1998	234.4	48.5			49			0.3684
	Coon, 1998	368.1	49.4	. 3.2	2.33	49			

		Discharge		Mean Depth	Mean Velocity	Width	Maximum Depth		Froude Number
River Name	Source	(m ³ /s) Q	(m) W	(m) Y	(m) V	(m) Wm	(m) Ym	(m/m) Sa	F
	Coon, 1998	51.8							
	Coon, 1998	73.6							
	Coon, 1998	88.9							0.32433
	Coon, 1998	94.3						0.00094	0.33142
	Coon, 1998	213.5	48.5			49	3.2	0.00094	0.36179
Tiouhgnioga	Coon, 1998	39.4	56.4	0.96	0.73	66	2.5	0.00051	0.23799
River	Coon, 1998	45	56.7	1.03	0.77	66	2.5	0.00051	0.2423
	Coon, 1998	47.6	57	1.06	0.79	66	2.5	0.00051	0.245
	Coon, 1998	54.4	57.3	1.14	0.83	66	2.5	0.00051	0.2483
	Coon, 1998	56.3	57.6	5 1.18	0.83	66	2.5	0.00051	0.2440
	Coon, 1998	64.8	57.9	1.27	0.88	66			0.2494
	Coon, 1998	66		1.28	0.89	66			0.2512
	Coon, 1998	77.9				66			0.2528
	Coon, 1998	77.9				66			0.2637
	Coon, 1998	79.6							0.2537
	Coon, 1998	79.8							0.2537
	Coon, 1998	101.6							0.2577
	Coon, 1998	118.1				66			0.2648
	Coon, 1998	122.3							0.2674
	Coon, 1998	159.7				66			0.2852
~	Coon, 1998	252							0.3097
Chenango	Coon, 1998	149.5				131			
River	Coon, 1998	182.1				131			
	Coon, 1998	187.4				131			
	Coon, 1998	209.8							
	Coon, 1998	234.1							
	Coon, 1998	239.2				131			
	Coon, 1998	253.7				131			
	Coon, 1998	302.9							
	Coon, 1998	317.1							
	Coon, 1998 Coon, 1998	325.6 385.1				131 131			
		399.2				131			
	Coon, 1998 Coon, 1998	416.2							
	Coon, 1998	410.2							
	Coon, 1998	424.7							
	Coon, 1998	447.3				131			
	Coon, 1998	569.1	123.0						
	Coon, 1998	750.3				131			
	Coon, 1998	164.2							
	Coon, 1998	212.9				131			
	Coon, 1998	231.6							
	Coon, 1998	235							
	Coon, 1998	270.1	121.3						
Genesee	Coon, 1998	94							
River	Coon, 1998	111	42.1						
	Coon, 1998	151.8							
	Coon, 1998	152.9							
	Coon, 1998	158.6							
	Coon, 1998	190.3				48			
	Coon, 1998	196.2							
	Coon, 1998	219.1	47.9						
Trout	Coon, 1998	5.8	27.6	0.48	0.44	32	1.6	0.00269	0.2028
River	Coon, 1998	17.2	29.2	0.72	0.82	32	1.6	0.00269	0.3086
	Coon, 1998	23.5	29.5	0.81	0.98	32	1.6	6 0.00269	0.3478
	Coon, 1998	93.4	31.7	1.53	1.92	32	1.6	0.00269	0,4958
	Coon, 1998	107.9	31.7	1.62	2.1	32	1.6	0.00269	0.5270
Naiau at Sunnyside	Hicks and Mason 1991	21.5	64.8	1.03	0.32	83			
	Hicks and Mason 1991	21.6	64.7	1.04	0.32				
	Hicks and Mason 1991	64.3	70.7	1.47					
	Hicks and Mason 1991	103	74.3						
	Hicks and Mason 1991	109	73.7	1.79	0.83	8 3			
	Hicks and Mason 1991	188	75.7	2.14	1.16	83			
	Hicks and Mason 1991	210	76.8	2.33	1.17	83	3.1	0.00011	0.2448
	Hicks and Mason 1991	405	80.4	2.8	1.8	83	3.1		
	Hicks and Mason 1991	527	83.1	3.08	2.06	83	3.1	0.00011	
Grey at Dobson	Hicks and Mason 1991	73	150.7	0.67	0.72	242	4	1 0.00094	0.2809
	Hicks and Mason 1991	116	158.4	0.77	0.95	242	4	0.00094	0.3458
	Hicks and Mason 1991	217	185.1	1.01	1.16	242	2	1 0.00094	0.3687

.	0	Discharge		Mean Depth	Mean Velocity	Width	Maximum Depth		Froude Number
River Name	Source	(m³/s) Q	(m) W	(m) Y	(m) V	(m) 	(m) Ym	(m/m) Sa	F
	Hicks and Mason 1991	358	193.2	1.32	1.4	242	4	0.00094	0.3892
	Hicks and Mason 1991	917	214.5	2.27	1.88	242	4	0.00094	
	Hicks and Mason 1991	1110	215	2.33	2.22	242	4	0.00094	0.4645
	Hicks and Mason 1991	3220	242.2	3,96	3.36	242	4	0.00094	0.539
Ongarue at Taringamotu	Hicks and Mason 1991	10.5	29.7	0.87	0.41	48	3	0.00067	0.1404
	Hicks and Mason 1991	14.8	30.3	1.01	0.48	48	3	0.00067	0.152
	 Hicks and Mason 1991 	18.7	30.7	1.08	0.56	48	3	0.00067	0.172
	Hicks and Mason 1991	19.2	30.8	1.06	0.59	48	3	0.00067	0.183
	Hicks and Mason 1991	35.1	31.8	1.31	0.84	48	3	0.00067	0.234
	Hicks and Mason 1991	35.8	32	1.32	0.85	48	3	0.00067	0.23
	Hicks and Mason 1991	41.7	32.2	1,38	0,94	48	3	0.00067	0.255
	Hicks and Mason 1991	241	47.5	3.03	1.67	48	3	0.00067	0.306
Hutt at Kaitoke	Hicks and Mason 1991	3.53	27	0.26	0.5	35	1.5	0.00473	0.313
	Hicks and Mason 1991	8.38	28.1	0.42	0.71	35	1.5	0.00473	0.349
	Hicks and Mason 1991	8.69	28.8	0.48	0.63	35	1.5	0.00473	0.290
	Hicks and Mason 1991	17.2			0.85	35	1.5		
	Hicks and Mason 1991	77.2			1.75				
	Hicks and Mason 1991	104			2.06				
Clarence at Jollies	Hicks and Mason 1991	7.62		0.38	0.65				
Construction of Construction	Hicks and Mason 1991	12.4			0.84	37			0.395
	Hicks and Mason 1991	17.5			1.01	37			0.333
	Hicks and Mason 1991	18.1	32.8		0.95				
	Hicks and Mason 1991	24			1.16				
	Hicks and Mason 1991	39.7			1.49				0.54
	Hicks and Mason 1991	64.8	35.4		1.85				
	Hicks and Mason 1991	106				37			
	Hicks and Mason 1991	1			2.38				
Annual - L - L - Drupper		120	36.8		2.36				0.64
Arnold at Lake Brunner	Hicks and Mason 1991	24.3			0.71	51			
	Hicks and Mason 1991	36.8	42.6		0.88				
	Hicks and Mason 1991	44.4			0.93				
	Hicks and Mason 1991	72.2			1.14	51			0.2
	Hicks and Mason 1991	84.4			1.27	51			
	Hicks and Mason 1991	125	50.5		1.31	51			
Rangitikei at Mangaweka	Hicks and Mason 1991	15.3	35.3		0.76				
	Hicks and Mason 1991	21.9			0.92				
	Hicks and Mason 1991	42.5	45.6		1.3				
	Hicks and Mason 1991	144	71.5	1.12	1.8	94			0.543
	Hicks and Mason 1991	173	79.7	1.28	1.7	94	2.3	0.00362	0.479
	Hicks and Mason 1991	342	86.3	1.83	2.16	94	2.3	0.00362	0.510
	Hicks and Mason 1991	413	89.2	2.04	2.27	94	2.3	0.00362	0.507
	Hicks and Mason 1991	542	94	2.34	2.46	94	2.3	0.00362	0.513
Buller at Woolfs	Hicks and Mason 1991	92.1	120.6	1.6	0.48	160	5.6	0.00076	0.121
	Hicks and Mason 1991	124	129.5	1.46	0.66	160	5.6	0.00076	0.174
	Hicks and Mason 1991	149	124.1	1.74	0.69	160	5.6	0.00076	0.167
	Hicks and Mason 1991	285	127.3	2.16	1.04	160	5.6	0.00076	0.226
	Hicks and Mason 1991	573	133.7	2.85	1.5	160			
	Hicks and Mason 1991	1079	140.5	3.75	2.05	160			
	Hicks and Mason 1991	2810	159.8	5.64	3.12	160			
Ngongotaha at SH5 Bridge	Hicks and Mason 1991	1.89	6.9	0.64	0.43	22			0.171
	Hicks and Mason 1991	2.07	7		0.44	22			0.171
	Hicks and Mason 1991	4.05	8	0.92	0.55	22			0.183
	Hicks and Mason 1991	5.5	9.2		0.56	22			
	Hicks and Mason 1991	5.95	9.5		0.58	22			0.17
	Hicks and Mason 1991	7.19	10.5	1.15	0.59	22			
	Hicks and Mason 1991	7.79	11.3	1.13	0.59	22			0.174
	Hicks and Mason 1991	8.98	12.8	1.24	0.59	22			0.174
	Hicks and Mason 1991 Hicks and Mason 1991	1				22			
		12	14.5	1.42	0.58				
Mongonui et Desterre	Hicks and Mason 1991	27.7	22.1	2.02	0.62				
Vanganui at Paetawa	Hicks and Mason 1991	32.6	83.1	1.54	0.25	155			
	Hicks and Mason 1991	45.9	84.7	1.76	0.31	155			0.074
	Hicks and Mason 1991	130	87.5	2.32	0.64	155			0.134
	Hicks and Mason 1991	381	89.6	3.57	1.19	155			0.201
	Hicks and Mason 1991	962	102.4	5.52	1.7	155			0.231
	Hicks and Mason 1991	1 1 90	107.2	6.14	1.81	155			
	Hicks and Mason 1991	1810	124	7.38	1.98	155			
	Hicks and Mason 1991	2130	131.7	7.97	2.03	155	9.2	0.00026	0.229
	Hicks and Mason 1991	2960	154.9	9.17	2.08	155			0.219
lutt at Taita Gorge	Hicks and Mason 1991	23.8	52.4	0.63	0.72	60			0.289
	Hicks and Mason 1991	59.4	53.9	0.97	1.14	60			
			20.0	2.21					

		Discharge		Mean Depth	Mean Velocity	Maximum Width	Depth		Froude Number
River Name	Source	(m³/s) Q	(m) W	(m) Y	(m) V	(m) Wm	(m) Ym	(m/m) Sa	F
	Hicks and Mason 1991	93							0.3843
	Hicks and Mason 1991	137	58.1		1.43				0.3556
	Hicks and Mason 1991	298							0.5341
Hoteo at Gubbs	Hicks and Mason 1991	24.1	16.6						
	Hicks and Mason 1991	27.9							0.1956
	Hicks and Mason 1991	39.4	20.1	2.01	0.97	43			
	Hicks and Mason 1991	39.8							
	Hicks and Mason 1991	52.3	22.4	2.16	1.08	43			0.2347
	Hicks and Mason 1991	54.3	23	2.21	1.07	43			0.2299
	Hicks and Mason 1991	72.1	24.9	2.43	1.19	43			
	Hicks and Mason 1991	99.2	30.6	2.62	1.24	43			
	Hicks and Mason 1991	149	41.1	2.92	1.24				
	Hicks and Mason 1991	156	42.6	3.03	1.21	43	: 3	0.00078	0.222
Vokau at Totoro Bridge	Hicks and Mason 1991	8.86	27	1.12	0.29	52	4.2	0.00137	0.0875
	Hicks and Mason 1991	98.2	31.9	2.15	1.43	52	4.2	0.00137	0.3115
	Hicks and Mason 1991	195	40.5	3.11	1.55	52	4.2	0.00137	0.2807
	Hicks and Mason 1991	240	46.6	3.52	1.46	52	4.2	2 0.00137	0.2485
	Hicks and Mason 1991	255	45.3	3.44	1.63	52	4.2	2 0.00137	0.2807
	Hicks and Mason 1991	271	49.1	3.77	1.46	52	4.2	0.00137	0.2401
	Hicks and Mason 1991	327	53.7	4.28	1.42	52	4.2	0.00137	0.2192
	Hicks and Mason 1991	349	51.8	4.21	1.6	52	4.2	0.00137	0.2490
Waipapa at Ngaroma Rd.	Hicks and Mason 1991	3.5	19.1	0.33	0.55	25	; 1	0.00748	0.305
	Hicks and Mason 1991	12.5	22	0.55	1.03	25	; 1	0.00748	0.4436
	Hicks and Mason 1991	22.9	22.8	0.67	1.5	25	1	0.00748	0.5853
	Hicks and Mason 1991	31.4	23.6	0.74	1.79	25	1	0.00748	0.6640
	Hicks and Mason 1991	38.5	23.9					0.00748	0.733
	Hicks and Mason 1991	57.4							
Whareama at Waiteko	Hicks and Mason 1991	23.1	17.9		0.64				
	Hicks and Mason 1991	26.6							
	Hicks and Mason 1991	30							
	Hicks and Mason 1991	36	21						
	Hicks and Mason 1991	200	37.9						
	Hicks and Mason 1991	220	41					1 0.00075	
	Hicks and Mason 1991	289	42.9						
Awanui at School Cut	Hicks and Mason 1991	8.8	12.3						0.202
	Hicks and Mason 1991	10.8	13.3			37			0.211
	Hicks and Mason 1991	13.6	14.5						
	Hicks and Mason 1991	22.7	18						0.181
	Hicks and Mason 1991	25.3	19.7		0.71	37			
	Hicks and Mason 1991	47.5	25.8						
	Hicks and Mason 1991	49.5			0.76				0.153
	Hicks and Mason 1991		20.1						0.155
	Hicks and Mason 1991	115	32.2						0.190
	Hicks and Mason 1991	1							
		143	34.5						
to to you at 181-top or deter	Hicks and Mason 1991	172	37.2						
Kaipara at Waimauku	Hicks and Mason 1991	13.6	16.8						0.132
	Hicks and Mason 1991	26.4	25.6						
	Hicks and Mason 1991	34.8	28.8						
	Hicks and Mason 1991	35.4	28.8						0.133
	Hicks and Mason 1991	36.2	29						0.133
	Hicks and Mason 1991	72							0.176
Ruakokapatuna at Iraia	Hicks and Mason 1991	0.08	4.8						
	Hicks and Mason 1991	0.2							
	Hicks and Mason 1991	0.22							
	Hicks and Mason 1991	0.29	7.1						
	Hicks and Mason 1991	0.41	7.1						
	Hicks and Mason 1991	0.89	7.8						
	Hicks and Mason 1991	3.89	9,5						
	Hicks and Mason 1991	6.63	10.3						
	Hicks and Mason 1991	10	11	0.63	1.45	12	. 0.	3 0.00601	
	Hicks and Mason 1991	10.9	11.1	0.64	1.53	12	.0.1	8 0.00601	0.610
	Hicks and Mason 1991	15.2	12.3	0.78	1.59	12	2 0,2	8 0.00601	0.575
Patea at McColls Bridge	Hicks and Mason 1991	2.8	21	1.3					0.028
-	Hicks and Mason 1991	46	27.1	2.1	0.81	38			0.178
	Hicks and Mason 1991	61	28.3						0.195
	Hicks and Mason 1991	62	28.5						0.197
	Hicks and Mason 1991	130	32.7						0.248
	Hicks and Mason 1991	218	37.6		1.65				0.281
		6.13							0.084
Pelorus at Bryants	Hicks and Mason 1991	1 1.1.1	35.4						

	_	Discharge		Mean Depth	Mean Velocity	Width	Maximum Depth		Froude Number
River Name	Source	(m ³ /s) Q	(m) W	(m) Y	(m) V	(m) Wm	(m) Ym	(m/m) Sa	F
	Hicks and Mason 1991	13.1	40.5	0.82	0.39	55	2.3		0.13757
	Hicks and Mason 1991	27.3	44	0.97	0.64	55	2.3	0.00359	0.20757
	Hicks and Mason 1991	79.7			1.2	55	2.3		0.35901
	Hicks and Mason 1991	164			1.78	55	2.3		
	Hicks and Mason 1991	290			2.34	55	2.3		
Collins at Drop Structure	Hicks and Mason 1991	0.07			0.1	15	1		
	Hicks and Mason 1991	0.16			0.13	15	1		0.0978
	Hicks and Mason 1991	0.23			0.17	15	1		
	Hicks and Mason 1991	0.55			0.31	15	1		
	Hicks and Mason 1991 Hicks and Mason 1991	2.35			0.71	15	1		
	Hicks and Mason 1991	5.31			1.2	15	1		
	Hicks and Mason 1991	13 30.9			1.68 2.05	15 15	1		
Mangara at Kara Wair	Hicks and Mason 1991	- 0.65			0.18	25	، 1.9		
Mangere at Kara Weir	Hicks and Mason 1991	0.86			0.23	25	1.9		
	Hicks and Mason 1991	1.34			0.23	25	1.9		
	Hicks and Mason 1991	6.95			0.7	25	1.9		0.23973
	Hicks and Mason 1991	7.67			0.72	25	1.9		
	Hicks and Mason 1991	7.69			0.73	25	1.9		
	Hicks and Mason 1991	9,24			0.66	25	1.9		0.20193
	Hicks and Mason 1991	12.5			0.71	25	1.9		0.19968
	Hicks and Mason 1991	20.4			0.95	25	1.9		0.25201
	Hicks and Mason 1991	29.5			1.16	25	1.9		
	Hicks and Mason 1991	46.6	19.8		1.4	25	1.9	0.00293	
	Hicks and Mason 1991	87	25.4	1.9	1.8	25	1.9	0.00293	0.41714
Waiwakaiho at SH3	Hicks and Mason 1991	2.44	19	0.45	0.29	35	1.7	0.01077	0.13809
	Hicks and Mason 1991	2.8	19.5	0.46	0.31	35	1.7	0.01077	0.14600
	Hicks and Mason 1991	3.43	20.3	0.49	0.35	35	1.7	0.01077	0.15971
	Hicks and Mason 1991	9.12	24.6	0.7	0.53	35	1.7	0.01077	0.20235
	Hicks and Mason 1991	21.8	26.3	0.92	0.9	35	1.7	0.01077	0.29973
	Hicks and Mason 1991	26.4	26.8		1.03	35	1.7		
	Hicks and Mason 1991	31.2			1.13	35	1.7		
	Hicks and Mason 1991	77.4	31.5		1.89	35	1.7		
	Hicks and Mason 1991	216	35		3.61	35	1.7		
Orere at Bridge	Hicks and Mason 1991	9.41	12.7		1.16	17	1.1		
	Hicks and Mason 1991	11.6	13.2		1.25	17	1.1		
	Hicks and Mason 1991	23.1	13.7		2.01	17	1.1		
	Hicks and Mason 1991	25.1	14		2.01	17	1.1		0.68059
	Hicks and Mason 1991 Hicks and Mason 1991	26.5 28.5	14.4 14.7		1.98 2.02	17 17	1.1		
	Hicks and Mason 1991	35.5	14.7		1.98	17	1.1		
	Hicks and Mason 1991	50.6	17.4		2.59	17	1.1		
Avon at Gloucester Street Bridg		1.83	11	0.4	0.42	15	1.1		
	Hicks and Mason 1991	2.32	11	0.42	0.5	15	1		
	Hicks and Mason 1991	3.74	11.3		0.61	15	1		
	Hicks and Mason 1991	4.48	11.6		0.65	15	1		
	Hicks and Mason 1991	4.87	11.5		0.68	15	1		
	Hicks and Mason 1991	6	11.7		0.74	15	1		
	Hicks and Mason 1991	8.91	12		0.89	15	1		
	Hicks and Mason 1991	12.2	13.3		1	15	1		0.33303
	Hicks and Mason 1991	15.6	14.9	1	1.05	15	1	0.00105	0.3354
	Hicks and Mason 1991	17.3	15.4	1.01	1.11	15	1		
	Hicks and Mason 1991	5.64	22.6		0.53	28	0.9		
	Hicks and Mason 1991	11.5	25		0.77	28	0.9		0.31754
	Hicks and Mason 1991	14.1	25.5		0.82	28	0.9		0.3200
	Hicks and Mason 1991	19.2	26.9		0.89	28	0.9		
	Hicks and Mason 1991	20.3	27.2		0.92	28	0.9		
	Hicks and Mason 1991	20.3	27	0.81	0.93	28	0.9		0.33008
	Hicks and Mason 1991	21.5	27	0.83	0.96	28	0.9		
	Hicks and Mason 1991	21.7	27.8		0.92	28	0.9		0.31876
	Hicks and Mason 1991	23	28.3	0.87	0.93	28	0.9		0.31850
	Hicks and Mason 1991	23.1	28	0.86	0,96	28	0.9		0.3306
	Hicks and Mason 1991	24.1	28.1	0.88	0.98	28	0.9		0.3337
Waikato at Ngaruawahia Cable		237	157.6	2.24	0.67	198	4.3		0.1430
	Hicks and Mason 1991	290	164	2.5	0.71	198	4.3		0.14344
	Hicks and Mason 1991	448	183.9	2.98	0.82	198	4.3		0.15173
	Hicks and Mason 1991	641 729	194.1	3.74	0.88	198	4.3		0.1453
	Hicks and Mason 1991	738	196.8	4.03	0.93	198	4.3		0.1479
	Hicks and Mason 1991	874	197.9	4.32	1.02	198	4.3	0.00016	0.15676

		Discharge	Width	Mean Depth	Mean Velocity	Maximum Width	Maximum Depth	Slope	Froude Number
River Name	Source	(m ³ /s)	(m)	(m)	(m)	(m)	(m)	(m/m)	Number
		Q	<u>w</u>	Y	<u>v</u>	Wm	Ym	Sa	F
Heathcote at Sloan Terrace	Hicks and Mason 1991	1.22	7.5		0.44	9	1.3		
	Hicks and Mason 1991	1.74	7.6		0.51	9	1.3		
	Hicks and Mason 1991	1.96	7.8		0.53	9			
	Hicks and Mason 1991	2.12	7.9		0.54	· 9 9			
	Hicks and Mason 1991 Hicks and Mason 1991	2.94	7.9		0.59 0.64	9			
	Hicks and Mason 1991	4.83	8.9		0.65	9			
	Hicks and Mason 1991	5.74	9.7		0.6				
	Hicks and Mason 1991	6.27	10.1		0.58	9			
	Hicks and Mason 1991	7.92	10.4		0.69	9			
	Hicks and Mason 1991	8.01	10.4		0.68	9			
	Hicks and Mason 1991	8.21	9.4	1.27	0.69	9	1.3	0.0005	0.1955
Taleri below Patearoa Power S	Hicks and Mason 1991	0.78	19.1	0.3	0.14	23	1	0.0009	0.081
	Hicks and Mason 1991	1.24	19.5	0.34	0.19	23	1	0.0009	0.1040
	Hicks and Mason 1991	6.13	20.2	0.54	0.56	23	1	0.0009	0.2434
	Hicks and Mason 1991	6.66	20	0.57	0.58	23			
	Hicks and Mason 1991	9.1	20.3		0.69	23			
	Hicks and Mason 1991	11.3	20.7		0.77	23			0.291
	Hicks and Mason 1991	11.8	20.1		0.81	23		0.0009	
	Hicks and Mason 1991	12	20.8		0.78	23			
	Hicks and Mason 1991	12.3	20		0.83	23			
	Hicks and Mason 1991	18.7	21.7		1.03	23			
	Hicks and Mason 1991	20.4			1.07	23			
	Hicks and Mason 1991	21.2	22.1 23		1.1	23 23			
Tahunatara at Ohakuri Road	Hicks and Mason 1991 Hicks and Mason 1991	27.1	13.4		1.23 0.25	20			
ranunatara at Onakun Koau	Hicks and Mason 1991	7.45	13.4		0.25	20			
	Hicks and Mason 1991	9.97	15		0.58	20			
	Hicks and Mason 1991	15.6	15.8		0.76	20			
	Hicks and Mason 1991	18.1	16.4		0.81	20			0.2218
	Hicks and Mason 1991	21.8	16.4		0.98	20			
	Hicks and Mason 1991	36	20.2		1.13	20			0.287
Rangitaiki at Te Teko	Hicks and Mason 1991	47.5	40.1		0.69	55			0.168
	Hicks and Mason 1991	53	40.6		0.73	55	2.7	0.00052	0.174
	Hicks and Mason 1991	74	42.9	2.07	0.83	55	2.7	0.00052	0.1842
	Hicks and Mason 1991	98	46.8	2.35	0.89	55	2.7	0.00052	0.185
	Hicks and Mason 1991	107	48	2.46	0.91	55	2.7		0.1853
	Hicks and Mason 1991	120	49.4	2.61	0.93	55	2.7	0.00052	0.183
	Hicks and Mason 1991	144	54.9		0.96	55			
viill Creek at Papanui	Hicks and Mason 1991	0.01	2.9		0.02	10			
	Hicks and Mason 1991	0.02	4.4		0.02	10			
	Hicks and Mason 1991	0.05	3.1		0.08	10			
	Hicks and Mason 1991	0.26	3.8		0.19	10			
	Hicks and Mason 1991	0.29	4		0.19	10			
	Hicks and Mason 1991	0.47	4		0.29	10			
	Hicks and Mason 1991 Hicks and Mason 1991	0.69	4.3		0.35	10 10			
	Hicks and Mason 1991	2.06	4.7 5.8		0.47	10			
	Hicks and Mason 1991	2.00	5.9		0.64 0.64	10			0.273
	Hicks and Mason 1991	2.14	6.1			10			0.275
	Hicks and Mason 1991	8.52	10.1		1.05	10			
	Hicks and Mason 1991	0.38	7.3		0.12	16			
	Hicks and Mason 1991	0.61	7.5		0.19	16			
	Hicks and Mason 1991	2.19	8.1		0.46	16			
	Hicks and Mason 1991	5.03	8.2		0.8	16			0.291
	Hicks and Mason 1991	7.72	10.1		0.94	16	1.2	0.00611	0.333
	Hicks and Mason 1991	12.2	12.3		1.08	16	1.2		
	Hicks and Mason 1991	12.3	12.3	0.91	1.1	16	1.2	0.00611	0.368
	Hicks and Mason 1991	17.9	13.7	1	1.31	16	1.2	0.00611	0.418
	Hicks and Mason 1991	20.2	14.2	1.05	1.34	16	1.2	0.00611	0.415
	Hicks and Mason 1991	25.1	15.7		1.36	16			
	Hicks and Mason 1991	29.3	16.3		1.47	16			0.423
lutchers Creek at Lake Kaniere		0.02	4.1		0.05	9			0.050
	Hicks and Mason 1991	0.05	5.4		0.07	9			0.059
	Hicks and Mason 1991	0.29	5.3		0.34	9			0.271
	Hicks and Mason 1991	1.75	6.6		0.85	9			0.487
	Hicks and Mason 1991	1.95	6.8		0.84	9			0.460
	Hicks and Mason 1991	1.99	6.7		0.9	9			
	Hicks and Mason 1991	4.31	7.4		1.33	9			
	Hicks and Mason 1991	4.8	7.4	0.45	1.44	9	0.7		0.685

		Discharge	Width	Mean Depth	Mean Velocity	Maximum Width	Maximum Depth	Siope	Froude Number
River Name	Source	(m ³ /s) Q	(m) W	(m) Y	(m) V	(m) Wm	(m) Ym	(m/m) Sa	F
	Hicks and Mason 1991	12.6		9 0.67					
	Hicks and Mason 1991	14.5							0.8948:
	Hicks and Mason 1991	16.7							0.93362
	Hicks and Mason 1991	18.9				9			
Waihua at Gorge	Hicks and Mason 1991	0.42							
	Hicks and Mason 1991	1.1							
	Hicks and Mason 1991	6.55							
	Hicks and Mason 1991	19.2				18			
	Hicks and Mason 1991	19.8							
	Hicks and Mason 1991	20.3							
	Hicks and Mason 1991	30.2							
Manager int ToDonos	Hicks and Mason 1991	32.1							0.6704
Wanganui at TePorere	Hicks and Mason 1991 Hicks and Mason 1991	0.93							0.1515
	Hicks and Mason 1991	1.17							0.14960
	Hicks and Mason 1991	1.17		B 0.43					0.1675
	Hicks and Mason 1991	2,66							0.2588
	Hicks and Mason 1991	13,1							0.4931
	Hicks and Mason 1991	15.8							
	Hicks and Mason 1991	16.2							
	Hicks and Mason 1991	29.3							0.7071
Opahi at Pond	Hicks and Mason 1991	0.25							0.0296
opulliur one	Hicks and Mason 1991	0.31							0.0329
	Hicks and Mason 1991	0.38							
	Hicks and Mason 1991	1.03							
	Hicks and Mason 1991	5.8						0.00113	0.1834
	Hicks and Mason 1991	5.88	9.	7 1.01	0.6	11	1.1	0.00113	0.1907
	Hicks and Mason 1991	7.46	10.	5 1.09	0.65	11	1.1	0.00113	0.1988
Huka Huka at Lathams Bridge	Hicks and Mason 1991	0.09	5.	5 0.11	0.15	9	0.5	5 0.04042	0.1444
	Hicks and Mason 1991	0.48	6.8	B 0.24	0.3	9	0.5	5 0.04042	0.1956
	Hicks and Mason 1991	0.63		7 0.25	0.36	9	0.5	5 0.04042	0.2299
	Hicks and Mason 1991	1.08	7.	4 0.29	0.5	9	0.6		
	Hicks and Mason 1991	1.63	7.	5 0.33	0.66			5 0.04042	0.3670
	Hicks and Mason 1991	1.93		7 0.34	0.74			5 0.04042	0.4053
	Hicks and Mason 1991	3.55							
	Hicks and Mason 1991	4.17							
	Hicks and Mason 1991	5.09							
	Hicks and Mason 1991	8.17							
Clark Fork	Barnes 1967	1950.75							
Clark Fork	Barnes 1967	891.85							
Blackfoot	Barnes 1967	232.16							
Coer d'Alene	Barnes 1967	319.93							
Salt	Barnes 1967	36.24							
Clearwater	Barnes 1967 Bornes 1967	2802.96							
Etowah WF Bitterroot	Barnes 1967 Bornes 1967	63.99							
	Barnes 1967 Barnes 1967	109.85							
Yakima MF Vermillion	Barnes 1967 Barnes 1967	784.26						5 0.00462 1 0.00295	
Wenatchee	Barnes 1967 Barnes 1967	642.7							
Movie	Barnes 1967	227.35							0.4971
Spokane	Barnes 1967	1121.18							0.4239
Bull	Barnes 1967	91.17							
MF Flathead	Barnes 1967	410.53							0.5385
M Oconee	Barnes 1967	172.99							0.2121
Chiwawa	Barnes 1967	166.48							
Grande Ronde	Barnes 1967	130.8							
Deep	Barnes 1967	235							
Chattahoochee	Barnes 1967	144.39							
sf Clearwater	Barnes 1967	356.74							
MB Westfield	Barnes 1967	96.26							

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Table A2 - Slope Control Flow Measurement Data

River Name	Source	Discharge (m ³ /s)	Width (m)	Mean Depth (m)	Mean Velocity (m/s)	Maximum Width (m)	Maximum Depth (m)	Siope (m/m)	Froude number
	11000.184/0							0.000.00	
Connecticut at Thompsonville	USGS-NWIS	2712.3							
	USGS-NWIS	2273.5		3.83		310			
	USGS-NWIS	1976.2		3.71					
	USGS-NWIS	1387.3		3.11					
	USGS-NWIS	1177.8		3.11					
	USGS-NWIS	1010.2		2.92		310			
	USGS-NWIS	925.8		2.87					
	USGS-NWIS	792.8		2.78					
	USGS-NWIS	628.5							
	USGS-NWIS	535.1		2.56					
	USGS-NWIS	407.7		2.43					
	USGS-NWIS	351.1							
	USGS-NWIS	257.1		2.22		310			
	USGS-NWIS	241.5		2.18		310			
	USGS-NWIS	209.5	298.7	2.04		310			
	USGS-NWIS	195.4		2.09		310			
	USGS-NWIS	165.1	282.8	1.93	0.3	310	4.1	0.00042	0
	USGS-NWIS	129.7	296.6	1.8			4.1		
	USGS-NWIS	94.6	297.5	1.8	0.18	310	4.1	0.00042	0
	USGS-NWIS	66.3	214.9	1.93	0.16	310	4.1	0.00042	0
ndroscoggin at Auburn	USGS-NWIS	1874.3	125	6.94	2.16	126	6.9	0.00051	C
	USGS-NWIS	1466.6	124	6.19	1.91	126	6.9	0.00051	C
	USGS-NWIS	1313.7	126.5	5.91	1.76	126	6.9	0.00051	C
	USGS-NWIS	1073		5.3		126	6.9	0.00051	C
	USGS-NWIS	823.9		4.92				0.00051	C
	USGS-NWIS	526.6		4.26			6.9	0.00051	
	USGS-NWIS	489.8		4.23					
	USGS-NWIS	359.6		3.83					
	USGS-NWIS	336.9		3.8		126			
	USGS-NWIS	245.2		3.46					
	USGS-NWIS	204.1	119.5	3.37		126			
	USGS-NWIS	188.8		3.36					
	USGS-NWIS	158.3		3.22		126			
	USGS-NWIS	131.1	114.3	3.15		126			
	USGS-NWIS	101.4		3.02		120			
		80.1		2.9					
	USGS-NWIS	1	115.8						
	USGS-NWIS	48.7		2.73					
	USGS-NWIS	40.5	112.8	2.65					
	USGS-NWIS	27.5	106.7	2.71					
elaware at Port Jervis	USGS-NWIS	1959.2		4.09					
	USGS-NWIS	1758.2		4.1					
	USGS-NWIS	1220.3		3.22					
	USGS-NWIS	1030.6		3.04					
	USGS-NWIS	775.8	192	2.62		196			
	USGS-NWIS	577.6	190.8	2.26					
	USGS-NWIS	474.5		2.04					
	USGS-NWIS	438.8	191.7	1.97					
	USGS-NWIS	419	191.1	1.92					
	USGS-NWIS	365.2	191.7	1.77					
	USGS-NWIS	334.1	189	1.75	1.01	196	4.1	0.00098	
	USGS-NWIS	308.6	189.6	1.72	0.94	196	4.1	0.00098	; (
	USGS-NWIS	281.7	190.5	1.6	0.92	196	4.1	0.00098	
	USGS-NWIS	241.2	189	1.5	0.85	196	4.1	0.00098	. (
	USGS-NWIS	186.6	186.2	1.38	0.73	196	4.1	0.00098	(C
	USGS-NWIS	163.6	186.2	1.3					
	USGS-NWIS	131.7	190.8	1.14					
	USGS-NWIS	103.6	185.3	1.1		196			
	USGS-NWIS	75	182.3	0.97					
	USGS-NWIS	47.8	179.2	0.79		196			
	USGS-NWIS	2686.9	324.6	3.95		325			
•	USGS-NWIS	863.5	193.5	3.71					
	USGS-NWIS	7.3	193.8	0.19					
		7.3	193.8	0.19		325			
	USGS-NWIS	1							
	USGS-NWIS	8.1	192	0.23		325			
	USGS-NWIS	10.4	203.6	0.21		325			
	USGS-NWIS	2194.2	262.1	4.32		325			
	USGS-NWIS	8.4	195.7	0.15					
	USGS-NWIS	10.3	203	0.19	0.27	325	4	0.00068	

PiverNemo	Source	Discharge (m ³ /s)	Width (m)	Mean Depth (m)	Mean Velocity (m/s)	Maximum Width (m)	Maximum Depth (m)	Slope (m/m)	Froude number
River Name	Source	(m /s)	(m)	(m)	(mvs)	(m)	(11)	(111/11)	
	USGS-NWIS	145		2.36	0.33	325			
	USGS-NWIS	222.5				325			
	USGS-NWIS	175.8				325			
	USGS-NWIS	564.8				325			
	USGS-NWIS	509.6				325			
	USGS-NWIS	223.4				325			
	USGS-NWIS	897.5				325			
Penobscot at W. Enfield	USGS-NWIS	3086.1				272			
	USGS-NWIS	2534				272 272			
	USGS-NWIS USGS-NWIS	1480.7				272			
	USGS-NWIS	959.8							
	USGS-NWIS	639.9				272			
	USGS-NWIS	475.7				272			
	USGS-NWIS	444.5				272			
	USGS-NWIS	402							
	USGS-NWIS	342.6				272			
	USGS-NWIS	334.1							
	USGS-NWIS	302.9				272			
	USGS-NWIS	276.9							
	USGS-NWIS	245.2				272			
	USGS-NWIS	235.8				272			
	USGS-NWIS	223.7				272			
	USGS-NWIS	196.8							
	USGS-NWIS	167							
	USGS-NWIS	137.3				272			
	USGS-NWIS	112.1							
aku near Juneau	USGS-NWIS	2992.6				240			
	USGS-NWIS	2304.6				240			
	USGS-NWIS	2024.3				240			
	USGS-NWIS	1809.2				240			
	USGS-NWIS	1492.1				240	5.2	0.0006	
	USGS-NWIS	1183.5				240			
	USGS-NWIS	917.3				240	5.2	0.0006	0.
	USGS-NWIS	461.5				240	5.2	0.0006	
	USGS-NWIS	393.5	50.6	7.36	1.06	240	5.2	0.0006	0.
	USGS-NWIS	334.1	196.6	1.98	0.86	240	5.2	0.0008	0.
	USGS-NWIS	278.9	196.9	1.77	0.8	240	5.2	0.0006	0.
	USGS-NWIS	204.4				240	5.2		
	USGS-NWIS	156	195.1	1.24	0.64	240	5.2	0.0006	0.
	USGS-NWIS	95.7	112.8			240	5.2	0.0006	0.
	USGS-NWIS	66.8				240	5.2	0.0006	0.
	USGS-NWIS	40.9	190,5	0.82	0.26	240	5.2	0.0006	0.
	USGS-NWIS	37.7	181.3	0.66	0.31	240	5.2	0.0006	0.
	USGS-NWIS	35.1	144.8	1.17	0.21	240	5.2	2 0.0006	0.
	USGS-NWIS	27.1	79.9	0.93	0.37	240	5.2	0.0006	0.
anana near Nenana	USGS-NWIS	2273.5	263	4.49	1.93	314	4.5	5 0.000203	
	USGS-NWIS	2015.9							
	USGS-NWIS	1823.3	251.1			314	4.5	0.000203	0.
	USGS-NWIS	1585.5				314	4.5	5 0.000203	0
	USGS-NWIS	1299.5	265.8	3.6	1.36	314	4.5	5 0.000203	0.
	USGS-NWIS	1101.4	271	3.01	1.35	314	4.5	5 0.000203	0.
	USGS-NWIS	724.8	209.1	3.11	1.12	314	4.6	5 0.000203	0.
	USGS-NWIS	546.4	227.1	2.33	1.03	314	4.5	5 0.000203	0.
	USGS-NWIS	461.5	208.8	2.69	0.82	314	4.5	5 0.000203	0.
	USGS-NWIS	242.9	228.6	2.4	0.44	314	4.5	5 0.000203	0.
	USGS-NWIS	222.3	248.4	3.48	0.26	314	4.5	5 0.000203	0.
	USGS-NWIS	195.1	201.2	2.22	0.44	314	4.5	5 0.000203	0
	USGS-NWIS	193.1	196.6			314	4.3	5 0.000203	
	USGS-NWIS	192.2	214.9	2.19	0.41	314	4.5	5 0.000203	0
	USGS-NWIS	186				314	4.5		
	USGS-NWIS	180.1				314			
	USGS-NWIS	175				314			
	USGS-NWIS	169.3				314			
	USGS-NWIS	165.9	210.3			314		5 0.000203	
	USGS-NWIS	156.9				314			
Susquehanna at Marietta	USGS-NWIS	6427							
	USGS-NWIS	5124.6	323.1			299			
	USGS-NWIS	3708.9	270			299			
	USGS-NWIS	2658.6							

River Name	Source USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS	(m ³ /s) 2064 1891.3 1449.6 1160.8 897.5 620 506.8 413.4 345.4 261.3 206.1 153.2 87.8	243.8 271.3 240.8 237.7 226.8 230.1 225.5 228.3 203 228.6	5.94 3.85 4.48 4.1 3.12 2.7 2.7 1.42	1.31 1.39 1.08 0.92 0.88 0.81 0.68	299 299 299 299 299 299 299	9.3 9.3 9.3 9.3	0.00044 0.00044 0.00044 0.00044	0.1 0.2 0.1
Mississippi at St. Cloud	USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS	1891.3 1449.6 1160.8 897.5 620 506.8 413.4 345.4 261.3 206.1 153.2	243.8 271.3 240.8 237.7 226.8 230.1 225.5 228.3 203 228.6	5.94 3.85 4.48 4.1 3.12 2.7 2.7 1.42	1.31 1.39 1.08 0.92 0.88 0.81 0.68	299 299 299 299 299 299 299	9.3 9.3 9.3 9.3 9.3 9.3	0.00044 0.00044 0.00044 0.00044	0.1 0.2 0.1
Mississippi at St. Cloud	USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS	1449.6 1160.8 897.5 620 506.8 413.4 345.4 261.3 206.1 153.2	271.3 240.8 237.7 226.8 230.1 225.5 228.3 203 228.6	3.85 4.48 4.1 3.12 2.7 2.7 1.42	1.39 1.08 0.92 0.88 0.81 0.81	299 299 299 299 299 299	9.3 9.3 9.3 9.3	0.00044 0.00044 0.00044	0.2 0.1
Vississippi at St. Cloud	USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS	1160.8 897.5 620 506.8 413.4 345.4 261.3 206.1 153.2	240.8 237.7 226.8 230.1 225.5 228.3 203 228.6	4.48 4.1 3.12 2.7 2.7 1.42	1.08 0.92 0.88 0.81 0.68	299 299 299 299	9.3 9.3 9.3	0.00044 0.00044	0.1
Vississippi at St. Cloud	USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS	897.5 620 506.8 413.4 345.4 261.3 206.1 153.2	237.7 226.8 230.1 225.5 228.3 203 228.6	4.1 3.12 2.7 2.7 1.42	0.92 0.88 0.81 0.68	299 299 299	9.3 9.3	0.00044	
Vississippi at St. Cloud	USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS	620 506.8 413.4 345.4 261.3 206.1 153.2	226.8 230.1 225.5 228.3 203 228.6	3.12 2.7 2.7 1.42	2 0.88 0.81 0.68	299 299	9.3		
Mississippi at St. Cloud	USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS	506.8 413.4 345.4 261.3 206.1 153.2	230.1 225.5 228.3 203 228.6	2.7 2.7 1.42	0.81 0.68	299			
Mississippi at St. Cloud	USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS	413.4 345.4 261.3 206.1 153.2	225.5 228.3 203 228.6	2.7 1.42	0.68		9.3		
Mississippi at St. Cloud	USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS	345.4 261.3 206.1 153.2	228.3 203 228.6	1.42					
Mississippi at St. Cloud	USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS	261.3 206.1 153.2	203 228.6		4 00				
Mississippi at St. Cloud	USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS	206.1 153.2	228.6			299 299			
Vississippi at St. Cloud	USGS-NWIS USGS-NWIS USGS-NWIS USGS-NWIS	153.2							
Mississippi at St. Cloud	USGS-NWIS USGS-NWIS USGS-NWIS								
Mississippi at St. Cloud	USGS-NWIS USGS-NWIS	1 07.0							
Mississippi at St. Cloud	USGS-NWIS	30.9		0.45					
		1282.6		5.24					
	USGS-NWIS	891.8							
	USGS-NWIS	651.2							
	USGS-NWIS	535.1							
	USGS-NWIS	501.1	190.5						
	USGS-NWIS	433.2							
	USGS-NWIS	393.5							
	USGS-NWIS	346							
	USGS-NWIS	317.1	187.4					0.00053	
	USGS-NWIS	302.9							
	USGS-NWIS	273.8	189,9	4.11	0,35	195	5.2	0.00053	3 0.0
	USGS-NWIS	221.1	184.7				5.2	0.00053	8 0.0
	USGS-NWIS	188.8	170.7	3.97	0.28	195	5.2	0.00053	S 0.0
	USGS-NWIS	132.2	169.2	1.36	0.57	195	5.2	0.00053	0 .
	USGS-NWIS	73.9	152.4	0.92	0.53	195	5.2	0.00053	3 O.
	USGS-NWIS	55.8	149	0.79	0.47	195	5.2	0.00053	3 0 .1
	USGS-NWIS	52.7	147.8	1.09	0.33	195	5.2	0.00053	30.
	USGS-NWIS	35.4	120.4	0.6	0.49	195	5.2	0.00053	B 0,1
	USGS-NWIS	25.7	90.2	0.62	0.46	195	5.2	0.00053	3 O.
Kennebec at Sydney	USGS-NWIS	5549.3	207.3	10.4	2.58	207	10.4	0.000061	I 0.:
	USGS-NWIS	2944.5	196.6	7.04	2.13	207	10.4	0.000061	I 0.:
	USGS-NWIS	2491.5	195.1	6.33		207	10.4		
	USGS-NWIS	2154.6							
	USGS-NWIS	1837.5	190.5						
	USGS-NWIS	1511.9	189				10.4		
	USGS-NWIS	1169.3							
	USGS-NWIS	872						0.000061	
	USGS-NWIS	586.1	176.8						
	USGS-NWIS	526.6	182.9						
	USGS-NWIS	396.4						0.000061	
	USGS-NWIS	351.1	179.8						
	USGS-NWIS	268.1						0.000061	
	USGS-NWIS	219.7							
	USGS-NWIS	164.5						0.000061	
	USGS-NWIS	105	173.7					0.000061	
	USGS-NWIS	75.9	179.8					0.000061	
	USGS-NWIS	62.9							
Markony of Salamanar	USGS-NWIS	26.8							
Allegheny at Salamanca	USGS-NWIS USGS-NWIS	852.2							
		637 472.8							
	USGS-NWIS								
	USGS-NWIS	342.6	114.9						
	USGS-NWIS	317.1	114						
	USGS-NWIS USGS-NWIS	302.9 276.9	108.8 114.6						
	USGS-NWIS	278.9	114.8						
	USGS-NWIS	201	105.8						
	USGS-NWIS	194.2	105.8	1.7					
	USGS-NWIS	194.2	106.1						
	USGS-NWIS	142.7							
	USGS-NWIS	142.7	106.7						
		115.2							
	USGS-NWIS USGS-NWIS		108.2						
		110.1							
	USGS-NWIS USGS-NWIS	89.2 81.3							
	USGS-NWIS	01.3	109.7			115	4.1		

River Name	Source	Discharge (m ³ /s)	Width (m)	Mean Depth (m)	Mean Velocity (m/s)	Maximum Width (m)	Maximum Depth (m)	Siope (m/m)	Froude number
Kiver Name	Source	(11.75)	(11)	(111)	(11/5)	(11)	(11)	(mvm)	
	USGS-NWIS	49			0.71	115			
	USGS-NWIS	32.6			0.67				
Hudson at Hadley	USGS-NWIS	487			1.18				
	USGS-NWIS	314.3			1	118			
	USGS-NWIS	240.7	112.8		0.87	118			
	USGS-NWIS	214.3				118			
	USGS-NWIS	184			0.77	118			
	USGS-NWIS	161.4	112.8		0.71	118			
	USGS-NWIS	141	112.8		0.64	118			
	USGS-NWIS	101.6		1.79	0.52				
	USGS-NWIS USGS-NWIS	87.2			0.48 0.47				
	USGS-NWIS	73.6	106.4		0.47	118 118			
	USGS-NWIS	59.5		1.57	0.41	118			
	USGS-NWIS	54.4	90.1		0.36	118			
	USGS-NWIS	43.3	88.4		0.38	118			
	USGS-NWIS	39.6	85.9			118			
	USGS-NWIS	38.5	85.9		0.23	118			
	USGS-NWIS	35.1	83.8		0.27	118			
	USGS-NWIS	33.1	85.3		0.25				
	USGS-NWIS	29.7	65.3 76.5		0.25	118			
	USGS-NWIS	18.8	76.3		0.25				
/atanuska at Palmer	USGS-NWIS	878.8	70.2 93.9		2.8				
hatanuska at i annoi	USGS-NWIS	693.7	84.7		2.81	94			
	USGS-NWIS	651.2	93.3		2.9				
	USGS-NWIS	617.2	93.6	2.59	2.54	94			
	USGS-NWIS	506.8	85.3	2.33	2.54	94			
	USGS-NWIS	450.2	68.6		2.61	. 94			
	USGS-NWIS	419	85.3	2.26	2.16	94			
	USGS-NWIS	390.7	90.5		2.32	94			
	USGS-NWIS	356.7	70.7	2.4	2.02	94			
	USGS-NWIS	331.3	68.6		2.25	94			
	USGS-NWIS	297.3	80.2		2.21	94			
	USGS-NWIS	272.4	87.8	1.72	1.8	94	3.3		
	USGS-NWIS	235.8	83.2		1.74	94			
	USGS-NWIS	212.6	84.7	1.32	1.91	94			
	USGS-NWIS	188.6	74.1	1.73	1.47	94			
	USGS-NWIS	132.5	67.1	1.43	1.38	94			
	USGS-NWIS	101.9	92		1.34	94			
	USGS-NWIS	47.3	83.5	0.59	0.96	94			
	USGS-NWIS	24.3	38.1	0.7	0.91	94			
/lerrimack at Franklin	USGS-NWIS	458.7	83.8	4.72	1.17	84			
	USGS-NWIS	421.9	80.8	4.69	1.11	84			
	USGS-NWIS	295	79.2	4.09	0.91	84			
	USGS-NWIS	185.4	77.1	3.54	0.68	84			
	USGS-NWIS	163.1	78	3.36	0.62				
	USGS-NWIS	145	77.7	3.27	0.58	84			
•	USGS-NWIS	113.8	74.7	3.11	0.49	84			
	USGS-NWIS	95.7	77.7	2.98	0.41	84			
	USGS-NWIS	90.3	77.4	2.94	0.4	84	4.7		
	USGS-NWIS	84.4	76.8	2.85	0.39	84			
	USGS-NWIS	74.2	76.2		0.34	84			
	USGS-NWIS	51	75.3		0.25				
	USGS-NWIS	32.8	75.6	2.43	0.18	84	4.7		
	USGS-NWIS	27.4	73.1	2.45	0.15	84			
	USGS-NWIS	25.9	74.7	2.44	0.14	84	4.7	0.0002	0
	USGS-NWIS	24	85.3	0.87	0.33	84	4.7	0.0002	0
	USGS-NWIS	20.8	73.1	2.39	0.12	84	4.7	0.0002	0
	USGS-NWIS	20.2	84.7	0.76	0.31	84	4.7		
Vhite at W. Hartford	USGS-NWIS	308.6	47.9	3.24	1.98	48			
	USGS-NWIS	233	42.7	3.12	1.75	48			
	USGS-NWIS	145.2	38.7	3.12	1.2				
	USGS-NWIS	109.6	38.1	2.97	0.97	48			
	USGS-NWIS	89.5	37.8	2.8	0.84	48			
	USGS-NWIS	82.4	36.9	2.85	0.79	48			
	USGS-NWIS	62.3	72.2	1.58	0.55	48			
	USGS-NWIS	47.3	35.4	2.61	0.51	48			
	USGS-NWIS	32.6	84.4	0.81	0.48	48			
	USGS-NWIS	30.9	95.4	0.44	0.73	48			
					00	,0			

		Discharge		Mean Depth	Mean Velocity	Maximum Width	Maximum Depth	Slope	Froude number
River Name	Source	(m³/s)	(m)	(m)	(m/s)	(m)	(m)	(m/m)	
	USGS-NWIS	28	93.9	0.38	0.79	48	3.2	0.0012	2 0.4*
	USGS-NWIS	24.7	75.3	0.71	0.46	48	3.2	0.0012	2 0.17
	USGS-NWIS	22.3	76.8	0.74	0.39	48	3.2	0.0012	2 0.14
	USGS-NWIS	19.6	57.6	0.52	0.65	48	3.2	0.0012	0.29
	USGS-NWIS	16.7	80.5	0.6	0.34	- 48	3.2	0.0012	2 0.14
	USGS-NWIS	13.5	74.4	0.65	0.28	48	3.2	0.0012	2 0.1
	USGS-NWIS	10.4	79.9	0.49	0.27	48	3.2	0.0012	2 0.12
	USGS-NWIS	8.1	81.1	0.49	0.2	48	. 3.2	0.0012	2 0.09
	USGS-NWIS	5.3	82	0.43	0.15	i 48	3.2	0.0012	2 0.07
Columbia	Barnes 1967	11494.96	529.4	8.53	2.55	529	8.5	0.00019	0.28
Columbia	Barnes 1967	28312.7	510.8	16.79	3.3	511	16.8	0.000266	6 0.26
Amazon at Obidos	Oltman 1968	216000	2290	48.03	1.96	2300	50	7.3E-06	5 0.09
	Oltman 1968	72500	2260	40.88	0.78	2300	50	7.3E-08	6 0.04
	Dury 1976	283170	2300	50.33	2.45	2300	50	7.3E-06	6 0.1 [.]
	Oltman 1968	165000	2280	46.49	1.56	2300	50	7.3E-06	6 0.07

Appendix 3 - Bankfull River Discharge and Channel Geometry Data

Table A3 - Bankfull River Discharge and Channel C	Geometry	Data Base
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River Name	Discharge (m ³ /s)	Water surface slope (m/m)	Channel width (m)	Average Channel Depth (m)	Average Channel Velocity (m/s)	Froude Number	Source
NA	36				3.25	0.67	Church and Room
NA	11				2.10		1983
NA	91.				0,85		(includes data fror
NA	58				0.84		Bray, 1979; and
NA	66				1.63		Williams, 1978).
NA	. 69.				0.68		
							1
NA	17.				2.08		
NA	23.				0.75		
NA	5.						
NA		3 0.0412			3.03		
NA	7.	5 0.0146	7	0.44	2.44	1.17	
NA	89.	2 0.0063	. 26	1.58	2.17	0.55	
NA	60.	9 0.0061	34	0.76	2.36	0,86	
NA	27.	5 0.0036	24	0.7	1.64	0.62	
NA	8.				0.46		
NA	36				2.92		
NA	1.				0.72		
NA	10.				0.48		
NA	10.				1.27		
NA	27				3.81		
NA	10.		20		0.60		
NA	18		32	2.51	2.33		
N A	22	6 0.0008	58	1.78	2.19	0.52	
NA	73.	6 0.0135	16	1.07	4.30		
NA	13.				2.04		
NA	12.				1.24		
NA	79.				1.91		
NA .	22.						1
					2.37		
NA	1				0.76		
VA	12.				0.93		
NA A	24.	5 0.0092	13	0.83	2.27	0.80	
٨٨	6.1	2 0.0237	4	0.41	3.78	1.89	
NA		5 0.0066			1.83		
NA	4.				3.00		
NA	5.				0.50		
NA	45.				0.91		
NA	85				2.18		
NA	22		90		2.04		1
NA	35				1.01		
NA	10.				1.55		
NA	5.				1.49		
NA	5.	5 0.0162	- 8	0.37	1.86	0.98	1
NA	4.	9 0.0263	8	0.46	1.33	0.63	
A	28.		24		1.55		
NA	13.						
NA	20.				1.29		
NA	10		35		1.25		
	1						
NA	6.2		9		1.89		
NA	2.7				1.99		
NA	5.8				1.58		
NA	1.5		5		1.27		
JA	7.73	3 0.0148	12	0.34	1.89	1.04	l I
NA	9.2	9 0.0107	9	0.64	1.61	0.64	4
A A	1.5				2.80		
A	12.				1.96		
IA	14				2.23		
NA NA	8.8		9		1.79		
A	3.5		4				
NA	6.1						
NA	25.3	3 0.0106	15	0.73	2.31		
NA	11.0	5 0.013	10	0.61	1.90	0.78	31
NA	1.5		2		2.57		
NA .	7.6		10		1.04		
NA	8.9		13		1.72		
NA	5.24		13		1.34		
NA	2.2		3		1.88		
NA	6.8		7		2.13		
NA	1.00	6 0.0165	3	0.4	0.88	0.45	

River Name	Discharge	Water surface slope	Channel width	Average Channel Depth	Average Channel Velocity	Froude Number	Source
	(m ³ /s)	(m/m)	(m)	(m)	(m/s)		
NA	0.538					0.35	
NA	2.42	0.0356	7	0.37	0.93	0.49	
NA	156	0.0018	50	1.5	2,08	0.54	
NA	17						
NA	3,4						1
NA	31						
NA	1.7						
NA	6.3						
NA	16.7						5
NA	510						
NA	666						
NA	9621						
	3						
NA	10895						
NA	16696						5
NA	9055						
NA	10753						•
NA	107		59				2
NA	68		51	2.16			
NA	101		58	2.52			
NA	141		54				
NA	127		54				3
NA	848						
NA	396						
NA	679						
NA	1058						
NA	2263	0.00036	253	4.78	1.87	0.27	
NA	565	0.00033	79	4.14	1.73	0.27	
NA	96	0.00033	51	2.01	0.94	0.21	
NA	152	0.00033	55	2.46	1.12	0.23	
NA	396	0.0005	39	7.19	1.41	0.17	
NA	79	0.0005	30	1.79	1.47	0.35	
NA	141						
NA	148						
NA	148						
NA	130		46				
NA	48		33				
NA	121		46				
NA	2605		442	5.33			
NA	7782		479	5.82			8
NA	6084		475	4.99			
NA	6933		475	5.36			
	121		-75	1.15			
NA							9
NA	48		40	0.85			1
NA	84		47	-1.03			
NA	52		35	0.76			
NA	40		28	0.73			1
NA	20		24	0.58			1 · · ·
NA	13		23	0.55			
NA	2829		201	3.47	4.06		
NA	690		151	1.79			
NA	933		162				
NA	165		49				
NA	84		40				
NA	155		48				
NA	18		21		0.88		
NA	37		25		1.36		
NA	169	0.0049	72	1.21	1.94	0.56	
NA	311	0.0049	73	1.64			
NA	152		95	1.24			
NA	52	0.0077	35	0.73			
NA	101	0.0077	41	0.82			
NA	65	0.0038	45	0.76			
NA	181	0.0038	62	1.03			
NA	486	0.0038	121	1.03			
							1
NA	299	0.0017	110	1.58			
NA	9004	0.00069	549	7.6			
NA	7220	0.00069	546	6.8			
NA	8212	0.00069	548	7.22			
NA	1416	0.0012	82				
NA	159	0.0012	67	2.1	1.13	0.25	

River Name	Discharge	Water surface slope	Channel width	Average Channel Depth	Average Channel Velocity	Froude Number	Source
	(m ³ /s)	(m/m)	(m)	(m)	(m/s)	NUMBER	
NA	278				1.57	0.32	
NA	55						2
NA	18				0.82		
NA	39				1.11		1
	2378				4.69		
NA							
NA	439				2.05		8
NA	595				2.41		
VA	178				1.38		1
NA .	68				1.32		
NA .	101				1.36		
JA	749				1.88		
NA .	1358				2.42		3
NA .	14998				2.83		6
NA .	8206				2.53		1
NA	9763		472		2.62		
A	2405	0.00052	271	4.2	2.11	0.33	8
A	3395	0.00052	274	5.18	2.39	0.34	
A	5433	0.00052	280	6.94	2.80	0.34	
A	834	0.00051	168	2.65	1.87	0.37	
IA	1584	0.00051	182		2.47		2
A	1839		111		4.31		
IA.	679		99		2.45		1
IA	919				2.91		1
1A	905				1.83		1
IA	1075				1.96		
IA	1075				2.60		
IA	481	0.0012		1.58	1.90		8
A	792	0.0012		2.07	2.32		
IA	1811	0.00084	133	4.57	2.96		
							9
IA	305	0.00084	111	1.67	1.65		3
A	543	0.00084	117	2.19	2.12		1
1A	113	0.00055	38		1.46		
A	28	0.00055	29	1.34	0.72		
A	56	0.00055	32		1.05		
A	203	0.0033	80	1.37	1.85		3
NA	79	0.0033	61	1	1.30		2
VA	166	0.0033	78	1.24	1.72	0.49	
A	155	0.0018	69	1.7	1.32	0.32	
A.	288	0.0018	75	2.31	1.66	0.35	
A	481	0.0026	95	2.37	2.14	0.44	
JA	594	0.0026	98	2.56	2.37	0.47	
A	367	0.004	94	1.58	2.47	0.63	
A	509	0.004	97	1.7	3.09	0.76	
JA.	314	0.002	104	1.43	2.11		
IA	404	0.002	107	1.61	2.35		
Ā	933	0.0012	85	3.29	3.34	0.59	
A	124	0.0012	63	1.4	1.41		
IA	234	0.0012	69	1.79	1.89		
IA	62	0.0036	31	1.21	1.65		
IA	28	0.0036	26	0.88	1.00		
iA IA	20 59	0.0036	20	1.18	1.22		
	53	0.0035	44	0.79	1.52		
IA IA	53	0.0035	44	0.79	1.52		
IA	130	0.0012	28	3.41	1.36		
A	10	0.0012	16	1.12	0.56		1
A	18	0.0012	19	1.4	0.68		
IA	339	0.0012	112	1.82	1.66		
A	650	0.0012	130	2.46	2.03		
IA	1018	0.00035	123	3.84	2.16		
A	350	0.00035	111	2.28	1.38	0.29	
A	693	0.00035	119	3.2	1.82	0.32	-
IA	141	0.0036	41	1.67	2.06	0.51	
IA	28	0.0036	27	0.88	1.18	0.40	
IA	59	0.0036	30	1.31	1.50	0.42	
IA	891	0.00044	180	2.62	1.89	0.37	
	1584	0.00044	192	3.56	2.32	0.39	
	42	0.0024	30	0.94	1.49	0.49	
IA	32	0.0024	28	0.91	1.26	0.42	1
IA IA	65	0.0032	31	1.24	1.69	0.49	
	33	0.0032	27	0.88	1.39	0.47	

River Name		Water surface	Channel	Average Channel	Average Channel	Froude	Source
	Discharge		width	Depth	Velocity	Number	
	(m ³ /s)	(m/m)	(m)	(m)	(m/s)		
NA	5				1.65	0.49	1
NA	277	7 0.0037	77	1.49	2.41	0.63	
NA	12	7 0.0037	57	1.12		0.60	
NA	203		69	1.34	2.20	0.61	
NA	1:				0.87	0.32	
NA	2				1.35		
NA	10						4
NA	21				1.14		
NA	6				1.32		
NA	203				1.88		
NA	42				1.02		8
NA	96				1.45		
NA	40				1.59		
NA	6					0.52	
					1.80		
NA	110				1.62		1
NA	16				1.89		
NA	42				0.98	0.25	3
NA					0.29		
NA	16				0.58	0.16	
NA	1				1.07	0.45	
NA							
NA	8				0.98		1
NA	11(2.26		
NA	16'				2.73		
NA	489				1.78		3
NA	99				1.06	0.33	
NA	175				1.26	0.35	
NA	10				1.23		
NA	39				2.33		
NA	268	0.00059			2.28	0.49	
NA	46				1.11	0.34	
NA	77	0.00059	41	1.43	1.31	0.35	
NA	23	0.0025	26	0.7	1.26	0.48	
NA	79	0.0025	31	0.85	3.00	1.04	ł
NA	13				1.22		1
NA	29				1.87		
NA	73				2.10		
NA	124				2.72		
NA	93				2.10		1
NA	164				3.27	1.11	
NA	110				1.39	0.40	
NA	220				1.00		
NA	186			1.28	2.46		9
NA	45		43		1.28		ê
NA	142				2.15	0.43	
NA NA	104				2.15		1
NA	169						1
					2.46		
NA	101				2.24		
NA	174				2.59	0.76	
NA	509				1.70	0.35	
NA	226				1.09	0.26	
NA	382				1.45		
NA	413				1.97	0.45	1
NA	549			2.22	2.23		1
NA	127				1.78		
NA	510				3.32		
NA	630				2.54		
NA	199				2.88		
NA	153			1.25	2.23	0.64	(
NA	8.5				1.45		
NA	161	0.0007	90	1.19	1.50	0.44	· ·
NA	141			3.51	1.49		
NA	11553		594	8.84	2.20	0.24	
NA	368			2.79	1.33		
NA	163			1.23	1.95		1
NA	73			3.51	0.56	0.10	
NA	2265		197	5.15	2.23	0.31	£
NA	4474		229	7.01	2.23	0.34	
	258						
NA			85	2.45	1.24		1
NA	1954	0.0015	161	4.57	2.66	0.40	1

River Name		Water surface	Channel	Average Channel	Average Channel	Froude	Source
	Discharge		width	Depth	Velocity	Number	
	(m ³ /s)	(m/m)	(m)	(m)	(m/s)		
NA	40						
NA	733				1.71	0.28	
NA	255				1.80	0.49	1
NA	354	0.00159			2.06		1
NA	9628	0.00056					
NA	6478						
NA	2890						
NA	2661	0.00056					
NA	2280						
NA NA	1402 2718	0.0004 0.00032				0.28 0.28	
NA NA	55						1
NA	425	0.0042					
NA	481	0.0019			1.85		
NA	198	0.00092					
NA	23	0.0007					
NA	64	0.0044					
NA	10	0.0064			0.98		1
NA	10.7	0.013			1.47		
NA	29.5						1
NA	66	0.0048					
NA	66	0.0105	19	1.36	2.55	0.70	
NA	140	0.0017					
NA	58	0.0057			2.37		8
NA	67	0.0018			1.22		
NA	25	0.0052			1.28		
NA	66	0.0024					
NA	81	0.0014					
NA	170	0.0074			2.25		
NA	260	0.0007			1.68		
NA	14.2	0.0032			1.21		
NA	36.5	0.0137			2.46		
NA	370	0.0015					
NA	66 2.7	0.0014			1.41 0.83		
NA	2.7	0.0023 0.0036			2.36		
NA	157	0.0009					
NA	550	0.0007			2.22		
NA	38	0.002			1.20		
NA	24	0.0037					
NA	40	0.0028			1.55		
NA	45	0.00066			1.97		
NA	66	0.00069					
NA	68	0.00062					
NA	13	0.003					
NA	4.8	0.0094				0.40	
NA	1.1	0.0193	3	0.4	0.92	0.46	
NA	3.8	0.0115			0.90		
NA	3.5	0.0125					
NA	2270	0.0015					
NA	0.61	0.0286			1.53		
NA	4.9	0.0175					
NA	3.6	0.0151					
NA	0.06	0.032					
NA	2096	0.00121					
NA	1042	0.00189					
NA .	1700	0.00221					
NA	3820 16300	0.01003			3.78		8
NA	16300	0.00007 0.00009					
NA	1500	0.00013					1
NA	4000	0.00013					2
NA	3426	0.00013			3.10		
NA	5154	0.0013			3.57		5
NA	2662	0.00055					
NA	3341	0.00065					
NA	1133	0.00027					
NA	850	0.0003					
		0.00031					
NA	850	0,00031	14.11	6.6	2.07	0	1

River Name	Discharge		Channel width	Average Channel Depth	Average Channel Velocity	Froude Number	Source
	(m ³ /s)	(m/m)	(m)	(m)	(m/s)		
NA	159				2.52		
NA	184				2.62		
NA	221				2.92		
NA	317						
NA	428						1
NA	3.4						
NA	2832		900				
NA	6.7						
NA	0.7						
NA	85.2						1 · · · ·
NA	7.1						8
NA	9.8						1
NA	12.2						1
NA	2.2						3
NA	2.7						
NA	1.9		5		1.23		1
NA	8.4						
NA .	22.6						
NA	4.5						1
NA	3.2		6				
NA	2.5				1.02		1
NA	49						8
NA	37.5	0.0067	26	0.91	1.58		
NA	7.1						3
NA	42	0.0058	25	0,88	1.91	0.65	
NA	101	0.0037	37	1.45	1.88	0.50	
NA	167	0.0018	53	1.63	1.93	0.48	
NA	46.7	0.002	24	1.62	1.20	0.30	
NA	255	0.00088	84	1.85	1.64	0.39	
NA	72.2	0.0071	31	1.13	2.06	0.62	
NA	114	0.0024	37	1.65	1.87	0.46	
Diamond	132.9	0.0196	33.2	. 1	4.00	1.28	Dingman and
Wild	205	0.0198	43.5	0.8	5.89	2.10	and Palaia, 1999
Ellis	33.7	0.049	20.2	1.3	1.28	0.36	
Lucy	16.1	0.039	14.4	1	1.12	0.36	
Saco	462.6	0.0018	69,8	2.2	3.01	0.65	
Oyster	8.5	0.0022	12.1			0.38	
Dudley	4.6	0.0015	8	0.6	0.96	0.40	
Pemii W	302.9						1 · · · ·
Stevens	5.8						
Baker	144.8						1
Pemi P	588.3						
Smith	49.1						•
Beards	33.9						
W Br War	8.9						
Warner	60.3						
Soucock	32.2						1
S Br Pisc	57.6						3
Stony	5.2						1
Halls	89.4						1
E Br Pass	35.7						
Moose	58.6						
Moose St	75.2						
Ammon	119						£
E Orange	6.9						
Mink	6.1						1
Ayers	20.1						
White	487.8						1
Williams	113.8						1
Saxtons	73.3						1
Cold	55.5						•
	26.6						•
S Br Ashu	4						1
Batten	90.5						1
Dog	88.7						
Mad	161.7						
Missisq	287.9						1
Black	58.2						
Paradise Creek near Paradise KS	36.81						Schumm, 1960
North Fork Solomon River near Downs, KS	226.5						
Prairie Dog Creek at Norton KS	73.61	0.0005	13.72	1.89	2.84	0.66	i l

River Name	Discharge		Channel width	Average Channel Depth	Average Channel Velocity	Froude Number	Source
Sappa Creek at Stamford NE	(m³/s) 50.96	<u>(m/m)</u> 0.0013	<u>(m)</u> 13.11	(m) 1.83	(m/s) 2.12	0.50	
Sappa Creek at Beaver city NE	38.22				2.12		3
Beaver Creek at Beaver City NE	28.31	0.000			1.70		5
Beaver Creek at Ludell KS	12.74						
Frenchman Creek at Hamlet, NE	24.07				1.11		
Blackwood Creek at Culbertson, NE	19.54				0.93		
Red Willow Creek near Red Willow, NE	62.85						
South Loup River near Cumro, NE	58.89	0.003	43.58			0.13	
White River at Interior, SD	308.61	0.002	89.3	1.77	1.95	0.47	
Cheyenne River at Edgemont, SD	103.62	0.0025	67.36	1,52	1.01	0.26	
Smoky Hill River near Russel, KS	226.5	0.00066	35.05	1.07	6.04	1.87	
Smoky Hill River near Danopolis, KS	260.48				5.53		
Smoky Hill River near Junction city , KS	368.06	0.0004			5.19		1
Kansas River at Wamego, KS	1104.19	0.0008			1.87)
Kansas River near topeka, KS	1359	0.0005	243.83		1.02		
Arikaree River at Haigler, NE	99.09	0.002			5.25		
S. F. Republican River near Benkleman, NE	127.41	0.002			5.97		
Republican river near Benkleman, NE Republican River near Bostwick, NE	61.58 339.75	0.003 0.0008			2.16 4.76		
Republican River at Concordia, KS	368.06	0.0008	46.94 76.2				
Republican River at Junction City, KS	424.69	0.0007	76.2 91.44		2.35		
South Fork powder River near Kaycee, WY	110.42	0.0007			4.35		,
Middle Fork Powder River above Kaycee WY	16.25	0.005			2.00		
Middle Fork Powder River near Kaycee, WY	46.15	0.0015	14.32		2.41		
Owl Creek near Thermopolis , WY	16.56	0.0015			1.30		
Gooseberry Creek at Pulliam, WY	8.81	0.006	17.98		0.67		
Greybull River near Basin,WY	88.9	0.0015			2.32		
Bates Creek near Alcova, WY	14.16	0.0035	21.03	0.85	0.79	0.27	
Powder River at Moorhead, MT	210.93	0.0016	64.61	1.22	2.68	0.77	
Red Fork at Barnum WY	18.12	0.005	10,67	0.76	2.23	0.82	
Tongue River near Acme, WY	97.68	0.002	30.48	1.31	2.45	0.68	
Horseshoe Creek near Glendo, WY	14.86	0.0025	19.51	0.82	0.93		
Smoky Hill River near Elkader, KS	84.94	0.006	152.39		0.46		
Republican River near Naponee, KS	321.35	0.0007	38.71	1.37	6.06		
Powder River near Sussex, WY	165.63	0.0008	53.95		2.72		
Powder River near Arvada WY	243.49	0.0007	51.81	1.37	3.43		
Missouri Landusky	850 683	0.00049	190		0.71		Osterkamp and
Missouri Culbertson Yellowstone Corwin	487	0.00016 0.0023	320 82.3		0.20 1.94		Hedman, 1982.
Yellowstone Livngston	584	0.0023	88.4		1.81		
Bighorn Bighorn	407	0.00045	82.3		1.55		
Yellowstone Miles City	1544	0.00068	219		0.97		
Missouri at Sioux City	963	0.00021	350		0.16		
Missouri Omaha	1811	0.00016	290	11.6	0.54	0.05	
Middle Loup St. Paul	235	0.001	134	1.07	1.64	0.51	
North Loup Ord	75.1	0.0013	75.6		1.01		
North Loup St. Paul	181	0.0011	85.3		1.40		
Elkhorn Norfolk	108	0.00069	80.8		1.32		3
Missouri Nebraska City	2554	0.00024	270	10	0.95		1
Missouri St. Joseph	2790	0.00021	270	10	1.03		
Kansas Wamego	1080	0.00025	223	11	0.44		
Kansas Topeka	1312	0.00027	159	8	1.03		
Kansas Lecompton	1561	0.00027	171	8.5	1.07		4
Kansas DeSoto	1420	0.00034	165	8.5	1.01		
Missouri Waverly	3200 640	0.00015	320 82.3	13 2.13	0.77 3.65		3
Thompson Trenton Missouri Booneville	2148	0.00076	82.3 430	2.13	3.65 0.29		3
Missouri Herman	4941	0.00018	430	17.2	0.29		
ColumbiaVenita	11494.9	0.00013	424 529.4		2.55		Barnes, 1967
Indian fork	21.7	0.00028	15.8	1.65	0.83		the second states
Champlin	67.7	0.0035	23.8	1.37	2.08		
Clark Fork	1950.7	0.00073	130.8	5	2.98		1
Clark Fork	891.8	0.00125	88.4	3.9	2.59		
Columbia	28312.6	0.00026	510.8	16.79	3.30		2
Esopus	393.5	0.0034	89.3	1.68	2.62		
Salt Cr.	52.7	0.00056	22.9	2.07	1.11	0.25	
Blackfoot	232.2	0.0023	59.1	1.86	2.11	0.49	
Coer d'Alene	319.9	0.0025	49.4	2.41	2.69	0.55	
Rio Chama	30	0.0012	27.4	1.04	1.05		
Salt	36.2	0.0019	57.9	0.67	0.93	0.36	1

		Water		Average	Average		
River Name]	surface	Channel	Channel	Channel	Froude	Source
	Discharge	slope	width	Depth	Velocity	Number	
	(m ³ /s)	(m/m)	(m)	(m)	(m/s)		
Beaver Kill	438.8						
Clearwater	2802.9						
Etowah	64			2.96	1.11	0.21	
WF Bitteroot	109.9	0.0046	32	1.46	2.35	0.62	
Yakima	784.3	0.003	67.4	3.57	3.26	0.55	
MF Vermilion	45.9	0.0031	35.7	1.01	1.27	0.40	
Weneatchee	642.7	0.0024	70.1	3.26	2.81	0.50	
Moyie	227.3	0.0036	44.8	2.16	2.35	0.51	
Spokane	1121.2						8
Tobesfokee	71.9	0.00077	25	2.74	1.05	0,20	
Bull Cr	91.2	0.0012	32.9	2.19	1.27	0.27	
NF Flathead	410.5	0.0036	55.5	2,68	2.76	0.54	
Middle Oconee	173	0.00047	43	3.32	1.21	0.21	
Beaver Cr	45.3	0.0012	14.9	2.65	1.15		
Catherine Cr	49.3	0.0067	17.4	1.28	2.21		
Chiwawa	166.5	0.0052	41.8	1.71	2.33	0.57	
Esopus	393.5	0.0045	54.3	2.53	2.86	0.58	
Grande Ronde	130.8	0.0053	34.7	1.62	2.33	0.58	1
Murder Cr	23.8	0.0027	13.7	1.4	1.24	0.34	
Provo	34	0.0089	15.5	1.07	2.05	0.63	
S Beaverdam	23.2	0.0016	18.6	1.4	0.89	0.24	
Deep	235	0.00077	66.7	3.17	1.11	0.20	
Clear Cr	39.1	0.0168	15.2	1.16	2.22	0.66	
Chattahoochee	144.4	0.0024	44.8	2.35	1.37	0.29	· ·
SF Clearwater	356.7	0.0063	46.3	2.71	2.84	0.55	
EB Ausable	220.6	0.0056	46.6	2.16	2.19	0.48	
MB Westfield	96.3	0.0087	36.3	1.34	1.98	0.55	
Mission Cr	3.5	0.0169	6.4	0.43	1.27	0.62	
NF Cedar	28.2	0.0237	18.6	0.79	1.92	0.69	
Merced	55.2	0.013	21.6	1.31	1.95	0.54	
Pond Cr	41.9	0.00064	31.4	2.47	0.54	0.11	1
Boundary	71.6	0.0187	25.6	1.34	2.09	0.58	
Amazon	283170	0.000013	3870	33	2.22	0.12	Dury, 1976*
Reference							-

Reference | *Discharge Prediction, Present and Future from Channel Dimensions, Journal of Hydrology vol. 30 pg. 219-245.

Table A4 - Prandt	l von-Karmen	Synthetic	River	Channel	Data	Base
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Channel I	lafa			Scalars			Estimated	Data			
Channel	Channel	Channel	Roughness					Тор	Mean	Mean	Froude
Width	Slope	Max Depth	Height	Max Y ¹	Distance ²	Integral ³	Discharge		Depth	Velocity	Number
Wm	S	Ym	ks	Y	x	Q'	Q	W	Y	V	F
(m)	(m/m)	(m)	(m)	(m)	(m)	(m ²)	(m ³ /s)	(m)	(m)	(m/s)	
30				1	4.7	0.31	0.14	9.5	0.10		
30					6.7	1.79		13.4			
30					8.2	4.27	3.24	16.4	0.29		0.39
30					9.5 10.6	7.64 11.84	6.70 11.60	19.0 21.2			
30 30		1.47		*	11.6	16.81	18.04	23.2			
30		1.47			12.5	22.51	26.10	25.1			
30		1.47		1	13.4	28.90		26.8			
30	0.008	1.47		1.32	14.2	35.97	47.29	28.5			0.64
30		1.47		1	15.0	43.67	60.53	30.0			
30		1.74			4.7	0.62	0.21	9.5			0.18
30		1.74 1.74		8	6.7 8.2	2.83 6.36	1.35 3.71	13.4 16.4			
30 30	-	1.74			9.5	11.05	8				
30		1,74		1	10.6	16.82	12.66	21.2			
30		1.74			11.6	23.58	8	23.2			0.46
30) 0.004	1.74		1	12.5	31.28	27.87	25.1			
30		1.74			13.4	39.87	37.98	26.8			
30		1.74		1	14.2	49.32	8				
30		1.74 2.05			15.0 4.7	59.60 1.08	63.47 0.28	30.0 9.5			
30 30		2.05		1	6.7	4.32		13.4			
30		2.05		8	8.2	9.31		16.4			
30		2.05		8	9.5	15.82		19.0			
30		2.05		9	10.6	23.73	13.74	21.2			
30		2.05		8	11.6	32.94	20.88	23.2			
30		2.05			12.5	43.36	29.69	25.1			
30 30		2.05		1.64 1.85	13.4 14.2	54,93 67.61	40.21 52.49	26.8 28.5			
30		2.05		9	15.0	81.34					
50		2.01		0.20	7.9	1.15	\$ ·	15.8			
50	0.005	2.01	0.0189	0.40	11.2	5.35	3	22.4			
50		2.01			13.7	12.06					
50		2.01		1	15.8	21.01	17.05				
50 50		2.01 2.01		1	17.7 19.4	32.01 44.93	29.04 44.64				
50		2.01			20.9	59.65	1	41.8			
50		2.01			22.4	76.08	1				
50	0.005	2.01			23.7	94.16	114.58	47.4	1.21		
50		2.01		1	25.0	113.82	1				
50		2.51		0.25	7.9	2.38	3				
50 50		2.51			11.2 13.7	9.34 19.94	3.78 9.89				
50		2.51		1	15.8	33.74					
50		2.51			17.7	50.46	1	35.4			
50		2.51	0.0115	1.51	19.4	69.88	49.01	38.7	1.00) 1.26	0.40
50		2.51			20.9						
50		2.51			22.4						
50 50		2.51 2.51			23.7 25.0						
50		2.51			25.0 7.9						
50		2.96			11.2						
51		2.96			13.7					0.67	0.28
5(0.001	2.96			15.8						
5		2.96			17.7						
50		2.96			19.4						
51 51		2.96 2.96			20.9 22.4		•				
50		2.90			22.4 23.7		•				
51		2.96			25.0		1				
10		3.30			15.8		1				
10	0.002	3.30			22.4						
10	0.002	3.30	0.0125	0.99	27.4	55.99	31.86	54.8	0.66	6.88	0.35

	81 IN	ata			Scalars	6			Estimated				
Channe		Channel	Channel	Roughness				•		Тор	Mean	Mean	Froude
Width		Slope	Max Depth	Height	Max Y ¹	I	Distance ²	Integral [®]	Discharge		Depth	Velocity	Number
Wm		S	Υ _m	k _s	Y)	ĸ	Q'	Q	W	Y	V	F
(m)		(m/m)	(m)	(m)	(m)		(m)	(m ²)	(m ³ /s)	(m)	(m)	(m/s)	
1	100	0.002	3.30			.32	31.6	94.22					0.38
1	100	0.002		0.0125	1.	.65	35.4	140.36	103.10	70.7			
	100	0.002			1	.98	38.7	193.84	8				
1	100	0.002	3.30			.31	41.8	254.21	1				0.44
	100	0.002	3.30		1	64	44.7	321.10	1				0.46
	100	0.002	3.30		1	97	47.4	394.22	1				
	100	0.002	3.30		1	.30	50.0	473.32					
	100	0.001	3.90		1	.39	15.8	10.97					0.21
	100	0.001	3.90		1	.78	22.4	39.08	1				
	100 100	0.001 0.001	3.90 3.90		1	.17 .56	27.4 31.6	80.46 133.34	1				
	100	0.001	3.90		4	.95	35.4	196.61	1				
	100	0.001	3.90			.34	38.7	269,46		70.7			
	100	0.001	3.90			.73	41.8	351.30	9				
	100	0.001	3.90		4	.12	41.0	441.64	1	89.4			
	100	0.001	3.90		5	.51	47.4	540.06	,				
	100	0.001	3.90		ž –	.90	50.0	646.22	8				
	100	0.0005	4.61	0.0041	6	46	15.8	16.72	8				
	100	0.0005	4.61	0.0041	4	92	22.4		1				
	100	0.0005	4.61	0.0041	ä	38	27.4	114.61					
	100	0.0005	4.61	0.0041		84	31.6	187.64					
	100	0.0005	4.61	0.0041		30	35.4	274.34	1				
	100	0.0005	4.61	0.0041	÷	76	38.7	373.65					
	100	0.0005	4.61	0.0041	3.	22	41.8	484.71	248.75	83.7	2.15	1.38	0.30
	100	0.0005	4.61	0.0041		69	44.7	606.88	332.95	89.4	2.48		
	100	0.0005	4.61	0.0041	4.	15	47.4	739.60	430.38	94.9	2.76	1.64	0.3
	100	0.0005	4.61	0.0041	4.	61	50.0	882.41	541.26	100.0	3.07	1.76	0.3
2	200	0.0012	4.91	0.0091	0.	49	31.6	28.05	8.70	63.2	0.33		
2	200	0.0012	4.91	0.0091	0.	98	44.7	99.64	43.73			0.75	0.29
;	200	0.0012	4.91	0.0091		47	54.8	204.88	1				
2	200	0.0012	4.91	0.0091		97	63.2	339.27					
	200	0.0012	4.91	0.0091		46	70.7	499.98	1				
	200	0.0012	4.91	0.0091		95	77,5	685.00					
	200	0.0012	4.91	0.0091		.44	83.7	892.77	1				
	200	0.0012	4.91	0.0091		93	89.4	1122.05					
	200	0.0012	4.91	0.0091	1	42	94.9	1371.82	1				
	200	0.0012	4.91	0.0091	1	91	100.0	1641.21	1				
	200	0.0006	5.80	0.0049	1	58	31.6	42.70	1				
	200	0.0006	5.80	0.0049		16	44.7	144.76					
	200	0.0006	5.80	0.0049		74	54.8	291.72	1				
	200	0.0006 0.0006	5.80 5.80	0.0049 0.0049		32 90	63.2 70.7	477.28 697.54	£ .				
	200	0.0006	5.80	0.0049					1				
	200 200	0.0006	5.80 5.80	0.0049		48 06	77.5 83.7	949.72 1231.71	•				
	200	0.0006	5.80	0.0049		64	89.4	1541.82	1				
	200	0.0006	5.80	0.0049		22	94.9	1878.67					
-	200	0.0006	5.80	0.0049		80	100.0	2241.08					
	200		6.85	0.0023		69	31.6	63.52					
	200	0.0003	6.85	0.0023		37	44.7	207.98					
	200	0.0003	6.85	0.0023		06	54.8	412.52	1				
	200	0.0003	6.85	0.0023		74	63.2	668.37	1				
	200	0.0003	6.85	0.0023		43	70.7	970.12					
	200	0.0003	6.85	0.0023		11	77.5	1313.97					
	200	0.0003	6.85	0.0023	1	80	83.7	1697.04					
	200	0.0003	6.85	0.0023		48	89,4	2117.03					
	200	0.0003	6.85	0.0023		17	94,9	2572.08	1				
	200	0.0003	6.85	0.0023		85	100.0	3060.59					
	300	0.0008	6.36	0.0065		64	47.4	66.39					
	300	0.0008	6.36	0.0065		27	67.1	227.21					
	300	0.0008	6.36	0.0065		91	82.2	459.78					
	300	0.0008	6.36	0.0065		54	94.9	754.15					
	300	0.0008	6.36	0.0065		18	106.1	1104.13					
	300	0.0008	6.36	0.0065		82	116.2	1505.29	6				
	300	0.0008	6.36	0.0065		45	125.5	1954.29	8				
;	000					09	134.2	2448.45					

Chann			Channel	Roughness	Scalars			Estimated		Moor	1000	Eroudo
Chann Width	ei	Channel Slope		Roughness Height	Max Y ¹	Distance ²	Integral ³	Discharge	Top	Mean Depth	Mean Velocity	Froude Number
wiath W _m		Siope	Ym	k _s	Y	X	Q'	Q	W	Y	Velocity	F
		(m/m)	(m)	(m)	(m)	^ (m)	(m ²)	(m ³ /s)	(m)	' (m)	(m/s)	t
(m)	300	0.0008			5.72	142.3	2985.54	2581.76	284.6			0.39
	300	0.0008			6.36	150.0	3563.70	3248.43	300.0			
	300	0.0004	7.51	0.0032		47.4	99.22	21.98				
	300	0.0004			3		327.19					
	300	0.0004				82.2		1	164.3			
	300	0.0004			3.00	94.9	1057.12		189.7			
	300	0.0004 0.0004			3.75 4.51	106.1 116.2	1536.61 2083.56	761.08 1130.48				
	300 300	0.0004	7.51	0.0032	5.26	125.5	2693.39	1				
	300	0.0004	7.51	0.0032	6.01	134.2			268.3			
	300	0.0004	7.51	0.0032	6.76	142.3	4087.74	2716.34				
	300	0.0004	7.51	0.0032	7.51	150.0	4866.74	3408.93	300.0	5.01		
	300	0.0001	10.47	0.0005	1.05	47.4	210.85	27.58	94.9	0.70	0.42	0.10
	300	0.0001	10.47		2.09	67.1	661.30					
	300	0.0001	10.47		3.14	82.2	1284.67	291.04				
	300	0.0001	10.47		4.19	94.9	2054.11	537.34				
	300	0.0001 0.0001	10.47 10.47	0.0005 0.0005	5.24 6.28	106.1 116.2	2953.34 3971.02	863.77 1272.26	212.1 232.4			
	300 300	0.0001	10.47	0.0005	7.33		5098.62	1764.41	252.4			
	300	0.0001	10.47	0.0005	8.38	134.2	6329.39	2341.56	268.3			
	300	0.0001	10.47	0.0005	9.43	142.3	7657.83	3004.87	284.6			
	300	0.0001	10.47	0.0005	10.47	150.0	9079.32	3755.36	300.0	6,98	1.79	0.2
	400	0.0007	7,36	0.0057	0.74	63.2	110.84	32.15	126.5			
	400	0.0007	7.36	0.0057	1.47	89.4	374.34	153.55				
	400	0.0007	7.36	0.0057	2.21	109.5	753.07	378.33	219.1			
	400 400	0.0007 0.0007	7.36 7.36	0.0057 0.0057	2.94 3.68	126.5 141.4	1230.83 1797.54	714.01 1165.84	253.0 282.8			
	400	0.0007	7.36	0.0057	4.41	154.9	2446.07	1737.87	309.8			
	400	0.0007	7.36	0.0057	5.15	167.3	3170.97	2433.40	334.7			
	400	0.0007	7.36	0.0057	5.89	178.9	3967.92		357.8			
	400	0,0007	7.36	0.0057	6.62	189.7	4833.36	4205.75	379.5	4.41	2.51	0.3
	400	0.0007	7.36	0.0057	7.36	200.0	5764.26	5287.09	400.0			
	400	0.0003	9.02	0.0022	0.90	63.2	179.26	37.68	126.5			
	400	0.0003	9.02	0.0022	1.80	89.4	581.58	172.89	178.9			
	400 400	0.0003 0.0003	9.02 9.02	0.0022	2.71 3.61	109.5 126.5	1148.53 1855.79	418.16 780.19	219.1 253.0			
	400	0.0003	9.02	0.0022	4.51	141.4	2688.40	1263.63	282.8			
	400	0.0003	9.02	0.0022	5.41	154.9	3635.87	1872.09	309.8			
	400	0.0003	9.02	0.0022	6.31	167.3	4690.27	2608.49	334.7			
	400	0.0003	9.02	0.0022	7.21	178.9	5845.29	3475.31	357.8	4.81	2.02	0.2
	400	0.0003	9.02	0.0022	8.12	189.7	7095.77	4474.69	379.5			
	400	0.0003	9.02	0.0022	9.02	200.0	8437.34	5608.51	400.0			
	400	0.0001	11.74	0.0005	1.17	63.2	323.41	44.78	126.5			
	400 400	0.0001 0.0001	11.74 11.74	0.0005 0.0005	2.35 3.52	89.4 109.5	1011.76 1963.00	198.12 470.78	178.9 219.1			
	400	0.0001	11.74	0.0005	4.70	126.5	3136.14		253.0			
	400	0.0001	11.74	0.0005	5.87	141.4	4506.36					
	400		11.74	0.0005	7.04	154.9	6056.38	2054.11	309.8			
	400		11.74	0.0005	8.22	167.3	7773.20	2847.64	334.7			
	400	0.0001	11.74	0.0005	9.39	178.9	9646.57	3777.93	357.8			
	400	0.0001	11.74	0.0005	10.56	189.7	11668.09					
	400	0.0001	11.74	0.0005	11.74	200.0	13830.73	6055.93	400.0			
	500	0.0006 0.0006	8.34 8.34	0.0049 0.0049	0.83 1.67	79.1 111.8	169.55 565.76	48.48 228.75	158.1 223.6			
	500 500	0.0006	8.34 8.34	0.0049	2.50	136.9	1131.99	228.75	223.0			
	500	0.0006	8.34	0.0049	3.34	158.1	1843.98	1054.41	316.2			
	500	0.0006	8.34	0.0049	4.17	176.8	2686.72	1717.62	353.6			
	500	0.0006	8.34	0.0049	5.00	193.6	3649.58	2555.87	387.3			
	500	0.0006	8.34	0.0049	5.84	209.2	4724.49	3573.76	418.3			
	500	0.0006	8.34	0.0049	6.67	223.6	5905.05	4775.19	447.2			
	500	0.0006	8.34	0.0049	7.51	237.2	7185.96	6163.50	474.3			
	500	0.0006	8.34	0.0049	8.34	250.0	8562.77	7741.69	500.0			
	500	0.0003	9.85	0.0022	0.99	79.1	250.27	54.98	158.1			
	500	0.0003	9.85	0.0022 0.0022	1.97	111.8	809.66	251.56	223.6			
	500	0.0003	9.85	0.0022	2.96	136.9	1596.81	607.64	273.9	1.97	⁷ 1.13	0.

	el D		Channel	Roughneen	Scalars			Estimated		Mean	5/ean	Froude
Chann Width		Channel Slope	Channel Max Depth	Roughness Height	Max Y ¹	Distance ²	Integral ³	Discharge	Top Midth	Mean Depth	Mean Velocitv	Froude Number
		Siope	•	neigni k _s	Y		ntegrai Q'		W	Y	Velocity	F
Wm			Y _m]	x		Q				r
(m)		(m/m) 0.0003	(m) 9.85	(m) 0.0022	(m) 3.94	(m) 158.1	(m ²) 2577.94	(m ³ /s) 1132.75	(m) 316.2	(m) 2.63	(m/s) 1.36	0.2
	500 500	0.0003	9.85 9.85					5				
	500	0.0003	9.85			193.6						
	500	0.0003	9.85		6.90			1				0.2
	500	0.0003	9.85		7.88			3				
	500	0.0003	9.85		8.87	237.2						0.3
	500	0.0003	9.85	0.0022		250.0	11694.56	8124.82				
	500	0.0001	12.82	0.0005	1.28	79.1	450.61	65.21	158.1	0.85	0.48	0,1
	500	0.0001	12.82		2.56		1407.01	6				
	500	0.0001	12.82		,		2727.22					
	500	0.0001	12.82		5.13		4354.37	8				
	500	0.0001	12.82		6.41	176.8	6254.02					
	500	0.0001	12.82		7.69	193.6	8402.20					
	500	0.0001 0.0001	12.82 12.82	0.0005 0.0005	8.98 10.26	209.2 223.6		1				
	500 500	0.0001	12.82	0.0005	11.54	223.0						
	500	0.0001	12.82	0.0005	12.82	250.0	19170.34	3				
	750	0.0005	10.23	0.0040	1.02	118.6	345.65		237.2			
	750	0.0005	10.23	0.0040	2.05	167.7	1136.22					
	750	0.0005	10.23	0.0040	3.07	205.4	2257.79					
	750	0.0005	10.23	0.0040	4.09	237.2						
	750	0.0005	10.23	0.0040	5.12	265.2	5319.96	3438.68	530.3	3.41	1.90	0.3
	750	0.0005	10.23	0.0040	6.14	290.5	7210.02	5105.18	580.9			0.3
	750	0.0005	10.23	0.0040	7.16	313.7	9316.60	6				
	750	0,0005	10.23	0.0040	8.19	335.4						
	750	0.0005	10.23	0.0040	9.21	355.8	14131.20					0.:
	750	0.0005	10.23	0.0040	10.23	375.0	16820.17	15375.49				0.:
	750	0.0002	12.75	0.0012	1.27	118.6	571.12					
	750	0.0002 0.0002	12.75 12.75	0.0012 0.0012	2.55	167.7	1812.96					
	750 750	0.0002	12.75	0.0012	5.10	205.4 237.2	3542.96 5686.70					
	750	0.0002	12.75	0.0012	6.37	265.2	8198.85					
	750	0.0002	12.75	0.0012	7.65	290.5	11047.73					
	750	0.0002	12.75	0.0012	8.92	313.7	14209.47	7672.00				
	750	0.0002	12.75	0.0012	10.20	335.4	17665.16		670.8			
	750	0.0002	12.75	0.0012	11.47	355.8	21399.33		711.5			
	750	0.0002	12.75	0.0012	12.75	375.0	25399.00					
	750	0.0001	15.06	0.0004	1.51	118.6	823.07	129.07	237.2	1.00	0.54	0.
	750	0.0001	15.06	0.0004	3.01	167.7	2561.35	568.03	335.4			
	750	0.0001	15.06	0.0004	4.52	205.4	4956.27	1346.19				
	750	0.0001	15.06	0.0004	6.02	237.2	7904.62					
	750	0.0001	15.06	0.0004	7.53	265.2	11343.99					
	750	0.0001	15.06	0.0004	9.03	290.5	15231.00					
	750	0.0001 0.0001	15.06 15.06	0.0004 0.0004	10.54 12.04	313.7 335.4	19533.07 24224.47	8104.19 10744.58				
	750 750	0.0001	15.06	0.0004	12.04	335.8 355.8	29284.19					
	750	0.0001	15.06	0.0004		375.0	34694.63					
	000	0.0003	12.96	0.0021	1.30	158.1	704.90					
	000	0.0003	12.96	0.0021	2.59	223.6	2261.63					
	000	0.0003	12,96	0.0021	3.89	273.9	4442.74					
	000	0.0003	12.96	0.0021	5.18	316.2	7154.50					
1	000	0.0003	12.96	0.0021	6.48	353.6	10339.58	5826.69	707.1	4.32	1.91	
1	000	0.0003	12.96	0.0021	7.78	387.3	13957.85	8616.45	774.6	5.18	2.15	0.
	000	0.0003	12.96	0.0021	9.07	418.3	17978.97		836.7			
	000	0.0003	12.96	0.0021	10.37	447.2	22378.92		894.4			
	000	0.0003	12.96	0.0021	11.67	474.3	27137.99		948.7			
	000	0.0003	12.96	0.0021	12.96	500.0	32239.62		1000.0			
	000	0.0001	16.87	0.0004	1.69	158.1	1261.79		316.2			
	000	0.0001	16.87	0.0004	3.37	223.6	3917.55		447.2			
	000	0.0001	16.87	0.0004	5.06	273.9	7571.71					
	000	0.0001	16.87	0.0004	6.75	316.2	12066.73		632.5			
	000	0.0001	16.87	0.0004	8.44	353.6	17307.48					
	000	0.0001 0.0001	16.87 16.87	0.0004 0.0004	10.12 11.81	387.3 418.3	23227.84	9445.23 13079.02	774.6 836.7			
		0.0001	10.07	0.0004	11.01	410.3	20110.10	100/8.02	000.7	1.01	1.99	0.

1000 0.0001 16.87 50.00 52261.34 2774.496 1000.0 10.25 2.4 1.33 0.6 1000 0.00005 18.93 0.0001 1.96 186.1 1790.6 223.66 6501.67 922.64 447.2 230.96 647.7 3.90 16.2 1.36 1000 0.00005 18.63 0.0001 7.97 315.2 16725.79 4267.30 632.5 3.31 15 116.0 1000 0.00005 18.93 0.0001 13.96 4183 4087.6 9802.54 774.5 0.64.7 19.64 147.2 6000.2 18.92 9802.54 774.5 0.63.7 19.62 10.000.1 19.94 2336.2 2383.2 948.7 11.89 20.001 18.93 0.0001 17.93 474.3 11404 2383.2 948.7 11.89 248.7 11.89 248.7 11.89 11.89 11.89 11.89 11.89 11.89 11.89 11.89 11.89 11.89 11.89 1	el Data	2			Scal	ars			Estimated	Data			
Wm S Y x C Q W Y V (m) (m)<				•		4							Froude
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$			-	-	1			-	÷.				Number
1000 0.0001 16.87 0.0004 16.87 0.000 10.12 22.22 1000 0.00005 19.93 0.0001 16.87 0.000 22.56 55.1 7744.96 000.0 11.25 2.4 1000 0.00005 19.93 0.0001 15.86 273.9 10549.74 233.09 647.7 30.99 1.5 1000 0.00005 19.93 0.0001 7.97 3187.81 2487.30 832.5 53.31 17.27 16.64 14.1 1000 0.00005 19.93 0.0001 11.96 33.73 3175.29 18257.85 836.4 10.63.1 15.64 1000 0.00005 19.93 0.0001 15.94 447.3 4469.2 2382.2 948.7 10.00.0 13.98 20.001 13.88 478.3 475.4 27.99 0.00 1000 0.00005 19.93 0.0001 19.93 2.0001 19.93 2.0001 19.82 2.0001 19.82			Υm	K _s	Y		х				Y	V	F
1000 0.0001 16.87 0.0004 16.87 50.0 52851.34 277.44 96 1000.0 1.25 2.2 1.33 0.6 1.25 2.2 1.33 0.6 1.25 2.2 1.35 0.6 1.2 1.35 0.6 1.2 2.33 0.6 4.77 3.96 1.6 1.35 1.35 2.336 0.36 2.336 0.36 2.336 0.36 2.336 0.36 2.336 0.36 2.336 0.36 7.7 3.6 1.4 <						45.40					and the second	main mark	
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notes:

maximum depth of flow (prescribed)
 distance from center of channel to bank
 value of equation (6) divided by 2.5V*