

Eastern South Dakota
Water Conference
Brookings, SD
Oct 13, 2011



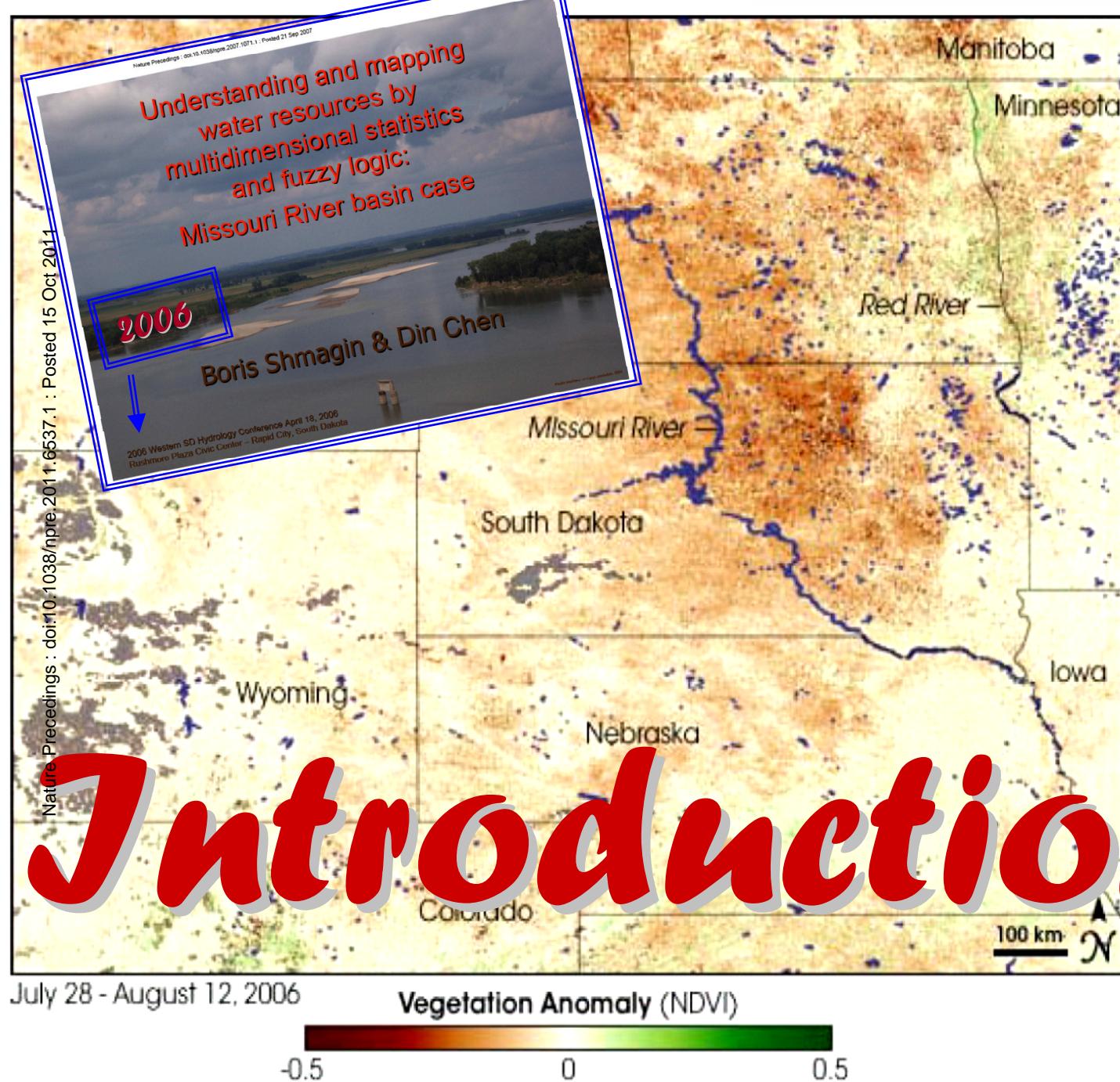
Missouri River Watershed: the Object for Hydrological Study & Uncertainty of Models

Boris Shmagin
WRI
SDSU

The goal:

To talk on my experience of study river flow variability with use of science

Nature Precedings : doi:10.1038/npre.2011.6537.1 : Posted 15 Oct 2011

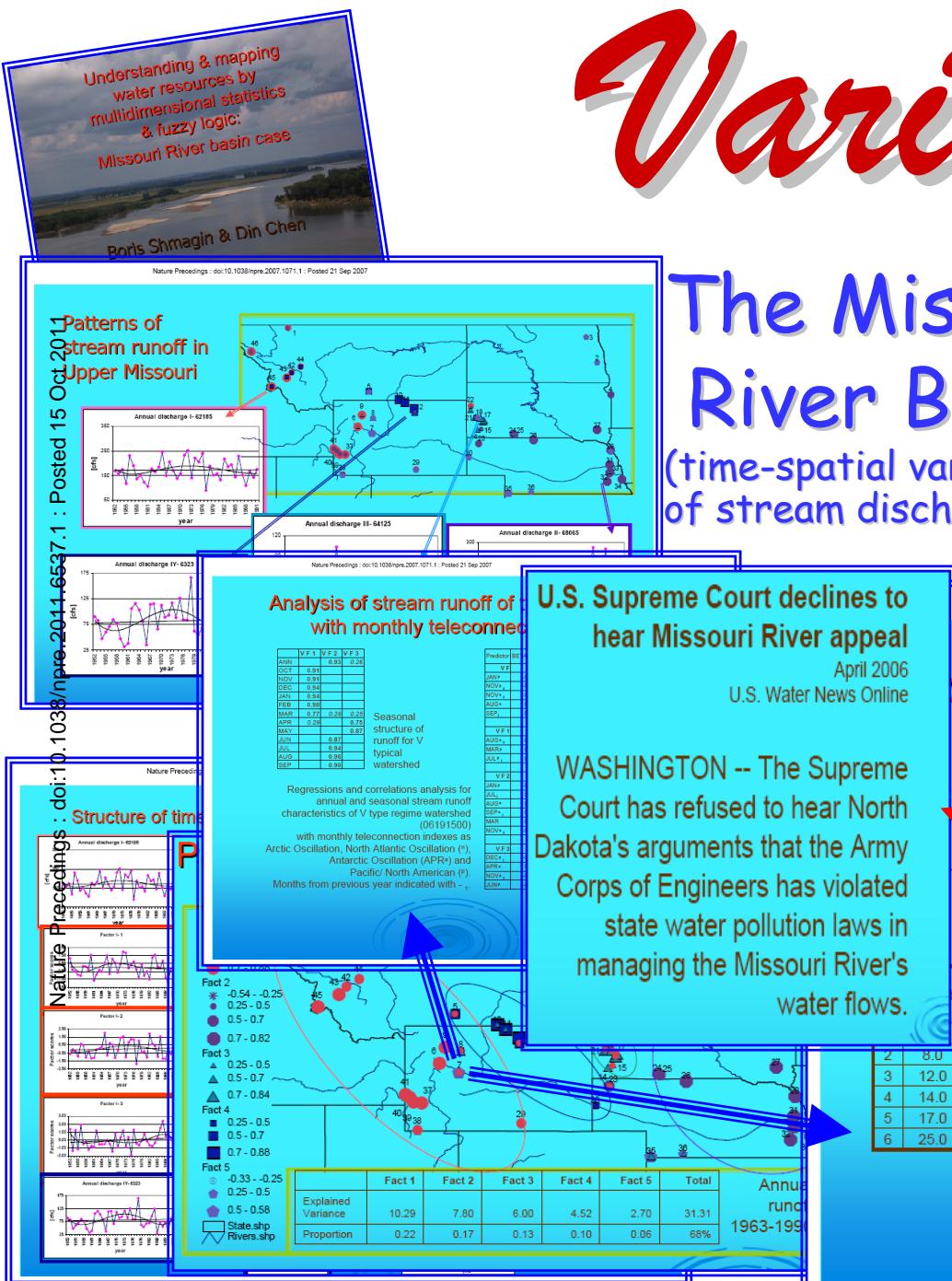


The topics:

- * Introduction:
Philosophy
for water
development
- * Knowledge &
Uncertainty
- * Results as
Math Models
- * Communicating
the
Knowledge
- * Conclusions

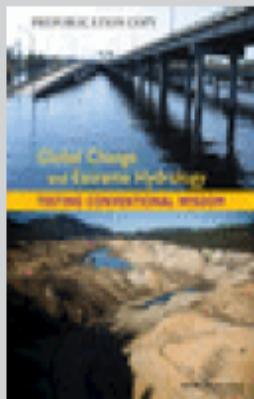
Variability as Math Models

The Missouri River Basin (time-spatial variability of stream discharge)



From the National Research Council

Nature Precedings : doi:10.1038/npre.2011.6537.1 : Posted 15 Oct 2011



Global Change and Extreme Hydrology: Testing Conventional Wisdom

GeoJournal (2011) 76:401–415
DOI 10.1007/s10708-009-9257-x

Democracy or expertise? objectivity as an elusive ideal in the resolution of a Vermont land use dispute

Thomas H.N. Young

ISBN
978-0-309-21768-2

60 pages
6 x 9
PAPERBACK (2011)

Committee on Hydrologic Science; National Research Council

... our relation to the hydrologic system requires a modicum of reverence for rivers

"The management of resources cannot be carried out successfully if it is looked upon as just another facet of economics, administration, & politics."

"The great geographer, William Morris Davis, viewed the river system as having a life of its own. Its *youthful headwaters*, he said, are steep and rugged. It rushes toward the sea, eroding bed & bank on its way. In its *central part, it is mature*, winding sedately through wide valleys adjusted to its duty of transporting water & sediment. Near its mouth it has reached, in its *old age*, a nearly level plain through which it wanders in a somewhat aimless course toward final extinction as it joins the ocean that had provided the sustaining waters through its whole life span."

"Man's engineering capabilities are nearly *limitless*. Our economic views are too *insensitive* to be the only criteria for judging the health of the river organism.

What is needed is a gentler basis for perceiving the effects of our engineering capabilities. This more humble view of our relation to the hydrologic system requires a modicum of reverence for rivers."

A reverence for rivers

Luna B. Leopold

Keynote address to the Governor's Conference on the California Drought
Los Angeles, California, March 7, 1972

Lula

Leopold

Posted 15 Oct 2011

The Concept of Entropy in Landscape Evolution

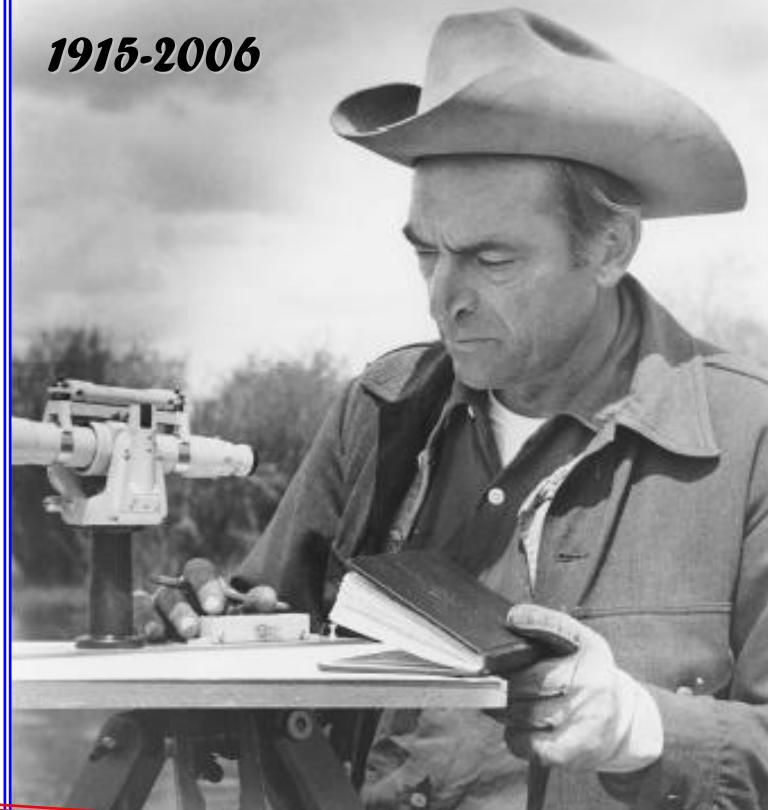
By LUNA B. LEOPOLD and WALTER

THEORETICAL PAPERS IN THE
GEOMORPHIC SCIENCES

GEOLOGICAL SURVEY PROFESSIONAL P



1915-2006



PHILOSOPHY FOR WATER DEVELOPMENT

Chief Hydraulic Engineer
Room 2227 GS Bldg.
Geological Survey

DO
DEPARTMENT OF THE INTERIOR

INFORMATION SERVICE

For Release to PM's, MARCH 28, 1961

ADDRESS BY LUNA B. LEOPOLD AND E. L. HENDRICKS, GEOLOGICAL SURVEY, DEPARTMENT OF
THE INTERIOR, AT THE NATIONAL WATER RESEARCH SYMPOSIUM, WASHINGTON, D. C.,
MARCH 28, 1961

UNITED STATES GOVERNMENT PRINTING OFFICE, WASHINGTON : 1

PHILOSOPHY FOR WATER DEVELOPMENT

The Knowledge & the Uncertainty



Toward a generalized theory of uncertainty (GTU)—an outline

Lotfi A. Zadeh

Berkeley initiative in Soft Computing (BISC),
and the Electronics Research Laboratory, Department
615 Soda Hall, Berkeley, CA 94720

Received 21 December 2004; accepted

Dedicated to Didier Dubois, Henri Prade and the
Richard Bellman and Herbert

Abstract
doi:10.1038/npl.2005.11
Published online 15 Oct 2005

INFORMATION SCIENCES 8, 199–249 (1975)

199

The Concept of a Linguistic Variable and its Application to Approximate Reasoning—I

L. A. ZADEH

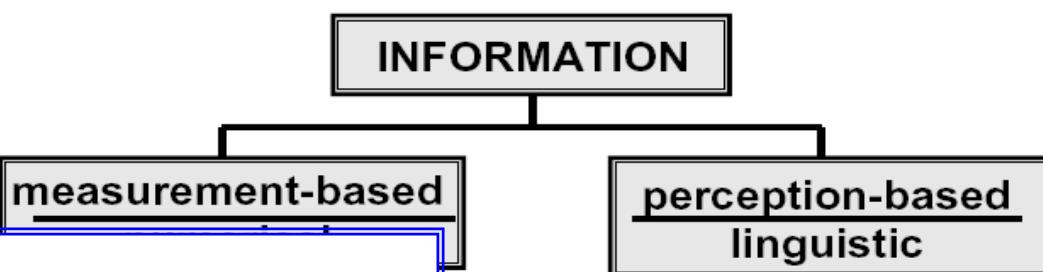
Computer Sciences Division, Department of Electrical Engineering and Computer Sciences,
and the Electronics Research Laboratory, University of California, Berkeley, California
94720

ABSTRACT

By a *linguistic variable* we mean a variable whose values are words or sentences in a natural or artificial language. For example, *Age* is a linguistic variable if its values are linguistic rather than numerical, i.e., *young, not young, very young, quite young, old, not very old and not very young*, etc., rather than 20, 21, 22, 23, In more specific terms,

The Uncertainty & Information

L. A. Zadeh / Information Sciences 172 (2005) 1–40



- It is very warm
- Most Swedes are tall
- probability is high
- it is cloudy
- traffic is heavy
- it is hard to find parking near the campus

tion may be viewed as a special case of

n is intrinsically imprecise

ed vs. perception-based information.

Information in the Language

4 Jürgen Van de Walle

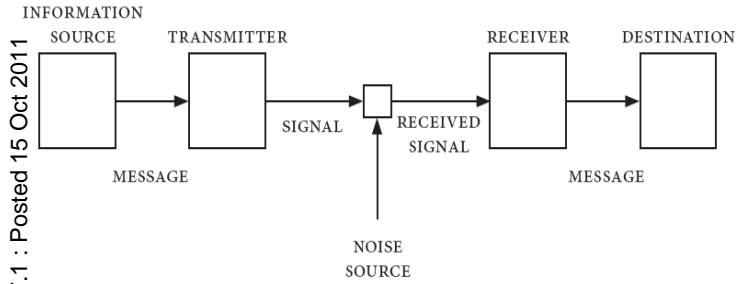
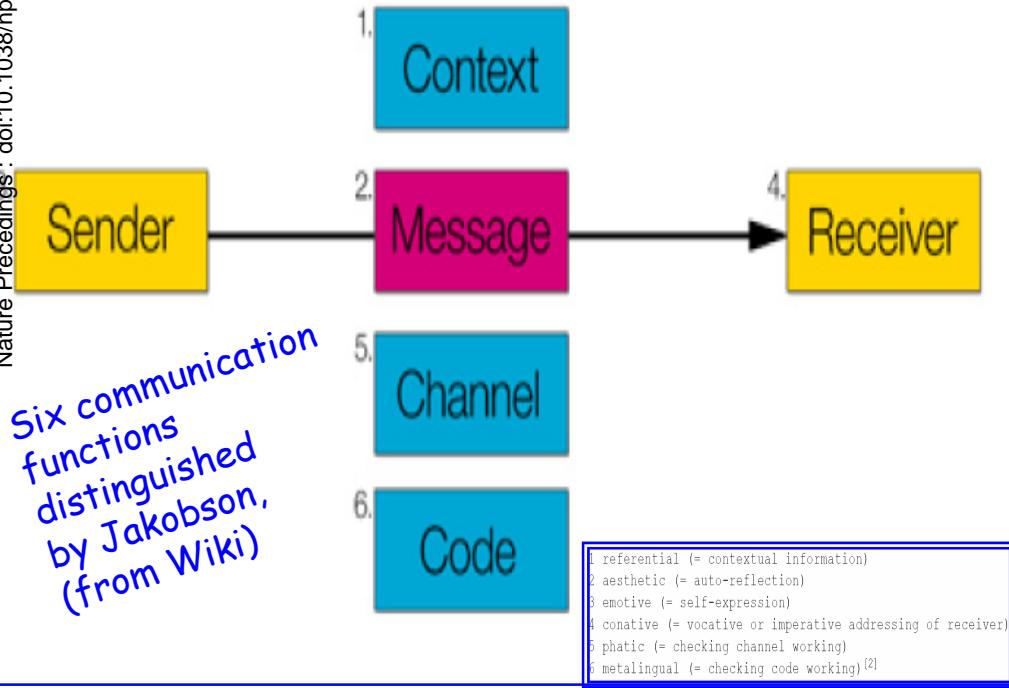


Figure 1. Shannon's communication channel

Nature Precedings| doi:10.1038/npre.2011.6537.1 : Posted 15 Oct 2011



Roman Jakobson, cybernetics and information theory
A critical assessment¹

Walle

Folia Linguistica Historica 29/1-2 (2008), 1-37.

"In cognitive linguistics as in cognitive science, the human mind is considered to be an information-processing device (Stillings 1995), & language is viewed as a vehicle for communicating information."

From: J. Van de Walle, 2008

The Science & the Language

15 Oct 2011 : Posted by John R. Searle
What is Language for Landau FNLSavas

Page 1

6 November, 2006

What is Language: Some Preliminary Remarks¹

By John R. Searle

Copyright John R. Searle

Naturalizing Language

I believe that the greatest achievements in philosophy over the past hundred or one hundred and twenty five years have been in the philosophy of language. Beginning with Frege, who invented the subject, and continuing through Russell, Wittgenstein, Quine, Austin and their successors, right to the present day, there is no branch of philosophy with so much high quality work as the philosophy of language. In my view, the only achievement comparable to those of the great philosophers of language is Rawls's reinvention of the subject of political philosophy (and therefore implicitly the subject of ethics). But with this one possible exception, I think that work in the philosophy of language is at the top of our achievements.

"Chomsky argues that no science has a mechanical procedure for discovering the truth anyway"

"the proper object of study was the speaker's underlying knowledge of the language, his "linguistic competence" that enables him to produce & understand sentences he has never heard before"

From: "Chomsky's Revolution in Linguistics" by John R. Searle
The New York Review of Books, June 29, 1972

Statistics & Uncertainty

The Statistician (2000)
49, Part 3, pp. 293–337

The philosophy of statistics

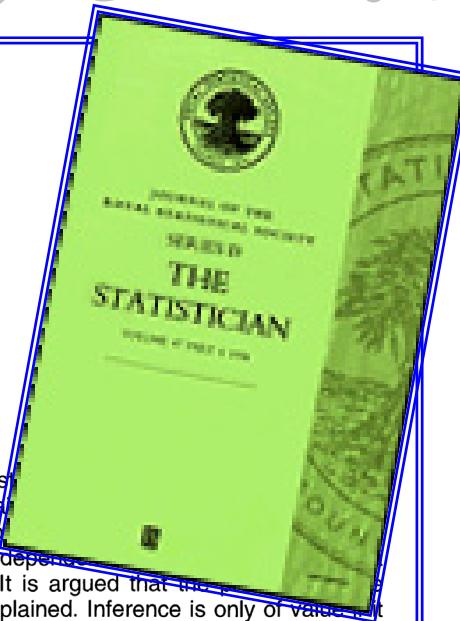
Dennis V. Lindley

Minehead, UK

[Received June 1999]

Summary. This paper puts forward an overall view of statistics as the study of uncertainty. The many demonstrations that uncertainty can be quantified by probability are summarized. The concept of probability is firmly based on probability alone. Progress is therefore dependent on a probability model; methods for doing this are considered. It is argued that the statistician's task is to help the client to handle uncertainty. The personal nature of probability is emphasized. The roles of likelihood and exchangeability are explained. Inference is only of value if it can be used, so the extension to decision analysis, incorporating utility, is related to risk and to the use of statistics in science and law. The paper has been written in the hope that it will be intelligible to all who are interested in statistics.

Keywords: Conglomerability; Data analysis; Decision analysis; Exchangeability; Law; Likelihood; Models; Personal probability; Risk; Scientific method; Utility



The statistician's task is
to articulate the
scientist's uncertainties
in the language of
probability...

A model is merely your
reflection of reality &
like probability,
it describes neither you
nor the world,
but only a relationship
between you &
that world." (p. 303)

"... data analysis assists in the formulation of a model & is an activity that precedes the formal probability calculations that are needed for inference." (p. 305)

"Statisticians are not masters in their own house.

Their task is to help the client to handle the uncertainty that they encounter. The 'you' of the analysis is the client, not the statistician." (p. 318)

From Data Analysis to Statistical Learning

THE PHILOSOPHY OF EXPLORATORY DATA ANALYSIS*

I. J. GOOD†

Statistics Department
Virginia Polytechnic Institute

J. R. Statist. Soc. A (1995)
158, Part 3, pp. 419–466

This paper
cisenly than us
and somewhat

A *data set* is
descriptive stat
resented in a m
finding patterns
probability of be
pattern is one fo
“cavity”, which i
to be probably pos
theory of probabi
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Introduction. Bo
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nized much more c
fluence of Tukey (19

The Elements of Statistical Learning

Data Mining, Inference, and Prediction

Second Edition

Model Uncertainty, Data Mining and Statistical Inference

By CHRIS CHATFIELD†

University of Bath, UK

[read before The Royal Statistical Society on Wednesday, January 18th, 1995, the President,
Professor D. J. Bartholomew, in the Chair]

SUMMARY

This paper takes a broad, pragmatic view of statistical inference to include all aspects of model formulation. The estimation of model parameters traditionally assumes that the model has a *prespecified known form* and takes no account of possible uncertainty regarding the model structure. This implicitly assumes the existence of a ‘true’ model, which many would regard as a fiction. In practice *model uncertainty* is a fact of life and is often more serious than other sources of uncertainty which have received far more attention from statisticians. This is true whether the model is specified on subject-matter grounds or, as is increasingly the case, when a model is formulated, fitted and checked using the data set in an iterative, interactive way. Modern computing power allows a large number of models to be considered and data-dependent specification searches have become a norm in many areas of statistics. The term *data mining* may be used in this context if the analyst goes to great lengths to obtain a good fit. This paper reviews the effects of model uncertainty, such as too narrow prediction intervals, and the non-trivial parameter estimates which can follow data-based modelling. Ways of assessing and reducing the effects of model uncertainty are discussed, including the use of simulation methods, a Bayesian model averaging approach and collecting additional information where possible. Perhaps the main aim of the paper is to ensure that statisticians are aware of the problems and start addressing the issues even if there is no simple, general theoretical fix.

ADDRESSING THE COMPLEXITY OF THE EARTH SYSTEM

Nature Precedings | doi:10.16988/npre.2011.6557.1 | Posted 15 Oct 2011

BY CARLOS NOBRE, GUY P. BRASSEUR, MELVYN A. SHAPIRO, MYANNA LAHSEN, GILBERT BRUNET, ANTONIO J. BUSALACCHI, KATHY HIBBARD, SYBIL SEITZINGER, KEVIN NOONE, AND JEAN P. OMETTO

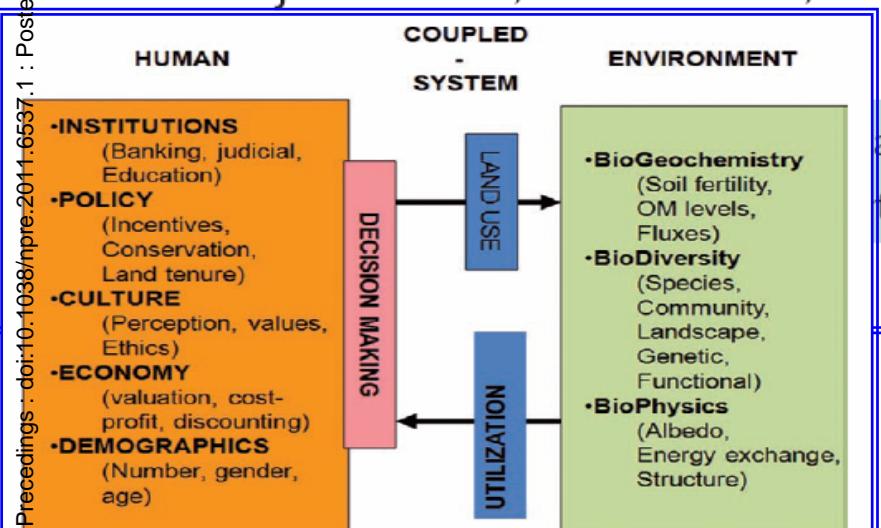


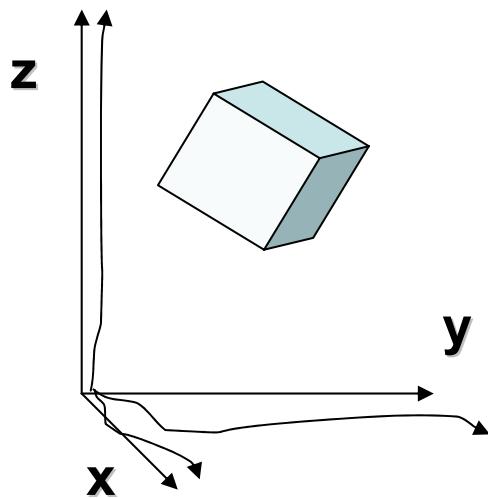
FIG. 4. An example of a model of a coupled human-environmental system that accounts for the influences of one subset of human actions (land use) on the natural systems and for the role of environmental goods and services for human welfare (utilization). [While "culture" is listed as a separate factor in this list, it is worth emphasizing that culture is a pervasive factor that also shapes institutions, economy, science, etc. (Proctor 1998).]

al, and societal processes would accelerate
with system prediction.

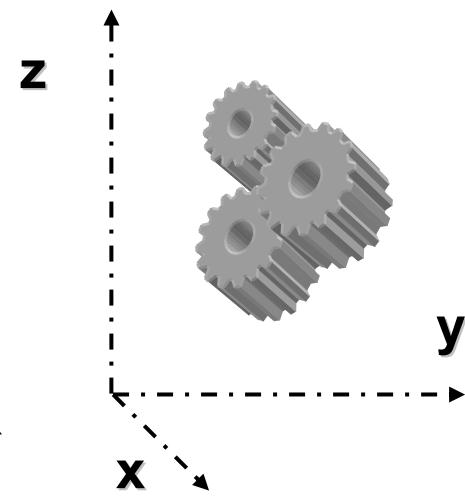
OCTOBER 2010 **BAMS**

& of the
Watershed

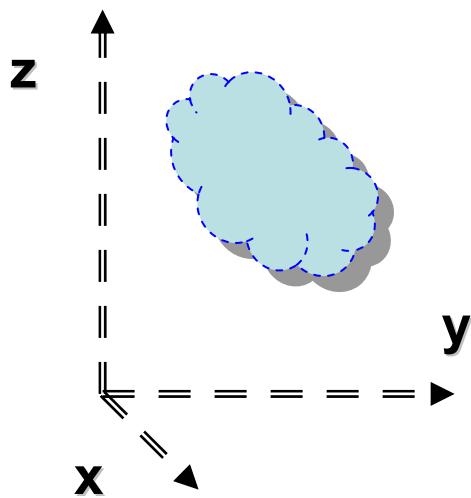
The Uncertainty & Different Systems of coordinates



Mathematical & physical objects are abstractions & "have" the principle of uncertainty



Technological objects have the errors of measurement



Natural objects have fuzzy boundaries in their own coordinates of multi-dimensional process & nonstationary axes

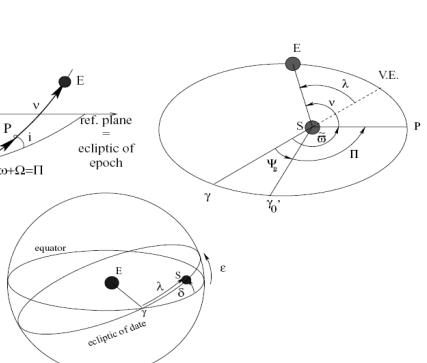
THE CLIMATE RESPONSE TO THE ASTRONOMICAL FORCING

M. CRUCIFIX*, M. F. LOUTRE and A. BERGER

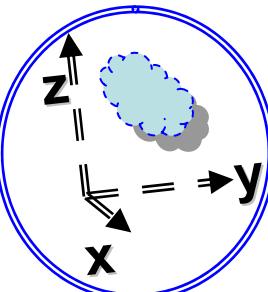
Institut d'Astronomie et de Géophysique G. Lemaître, Louvain-la-Neuve, Belgium

(*Author for correspondence: E-mail: michel.crucifix@uclouvain.be)

(Received 30 August 2005; Accepted in final form 22 February 2006)



years. Two
theory of
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be essential
2) to infer
the impact
important



coordinates for the Earth



National Science Foundation
WHERE DISCOVERIES BEGIN

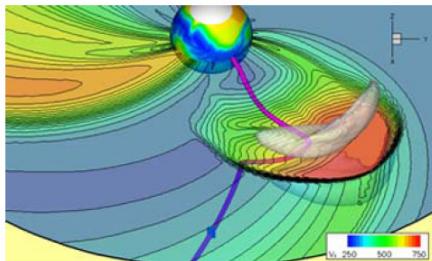
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Nature Precedings : doi:10.1038/npre.2011.6357.1

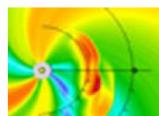
Press Release 11-016 First Large-Scale, Physics-Based Space Weather Model Transitions Into Operation

Provides forecasters with one-to-four-day advance warning of 'solar storms'

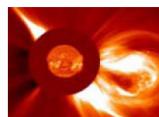


A coronal mass ejection (CME) in a model; the CME is the gray cloud toward the lower right.

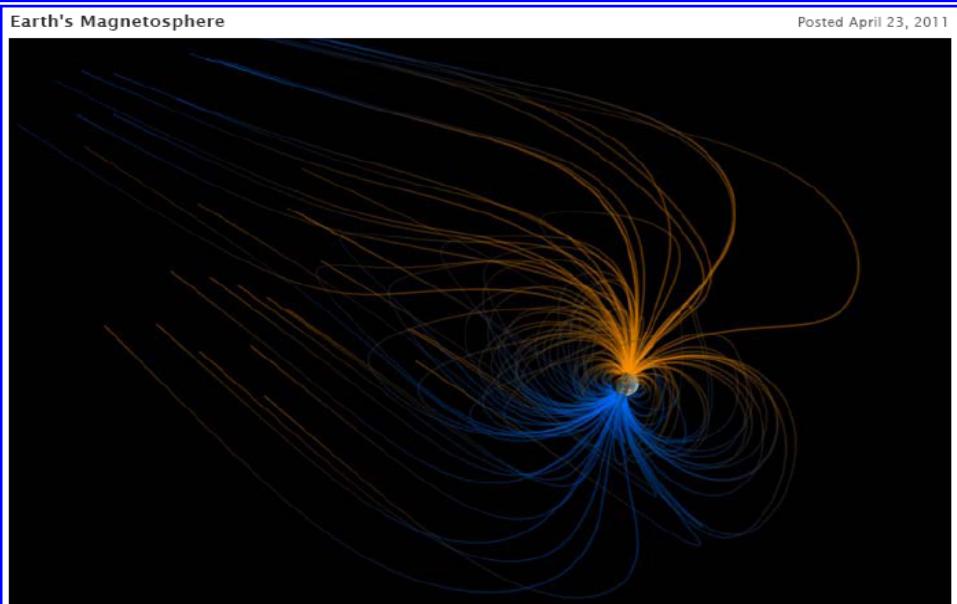
[Credit and Larger Version](#)



This Center for Integrated Space Weather Modeling display will be used to predict space weather.
[Credit and Larger Version](#)



A large coronal mass ejection observed by the SOHO spacecraft December, 2003.
[Credit and Larger Version](#)



[Next Image](#)

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You've seen the [pattern](#) in science class when you laid bits of iron around a bar magnet. The invisible force field around the magnet becomes suddenly visible when the iron filings fall into line.

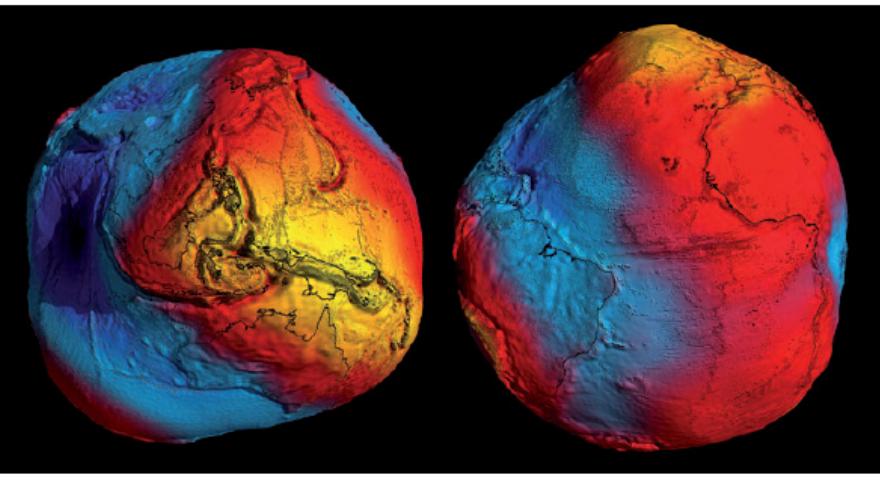
The iron-cored Earth is also a great magnet, and scientists have spent a century

The coordinates on the Earth

Taking the “Boulder” Step From Static to Dynamic Geoid

2009 Workshop on Monitoring North American Geoid Change;
Boulder, Colorado, 21–23 October 2009

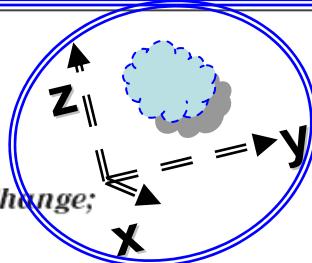
PAGE 46



The pull of the planet

The most detailed map of Earth's gravity ever made was unveiled last week in Munich, Germany, when researchers presented eight months' worth of data from the European Space Agency's Gravity Field and Steady-State Ocean Circulation Explorer (GOCE), a satellite launched in 2009. GOCE maps subtle variations in Earth's gravitational field that arise

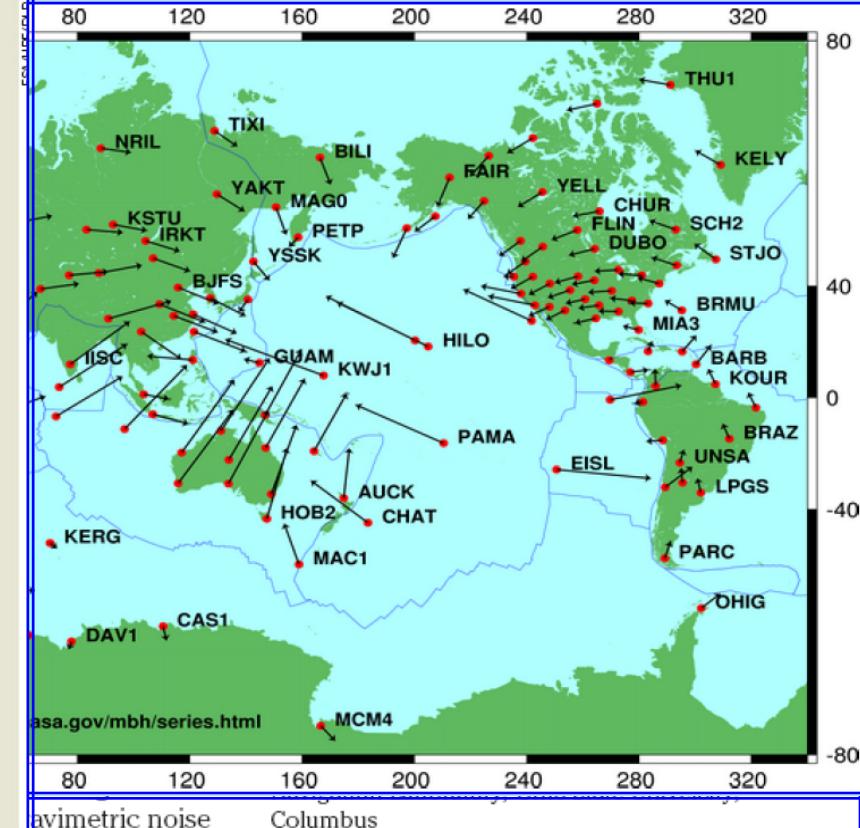
from the planet's uneven distribution of mass. The result is a 'geoid' (pictured — variations exaggerated 10,000 times), showing the world if it were covered by an ocean whose height was influenced only by gravity. This reference allows geoscientists to precisely measure the heights of shifting oceans and continents. GOCE will continue mapping until the end of 2012.



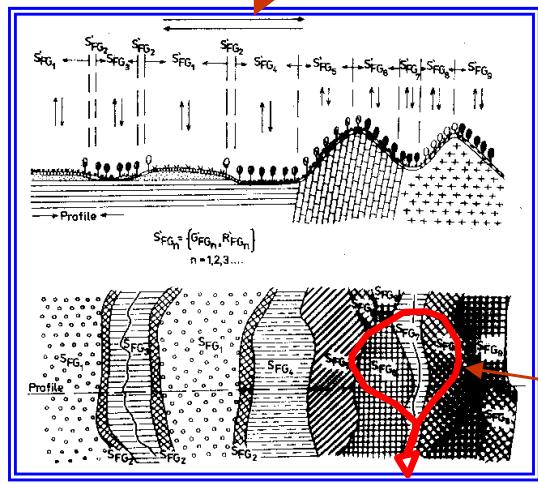
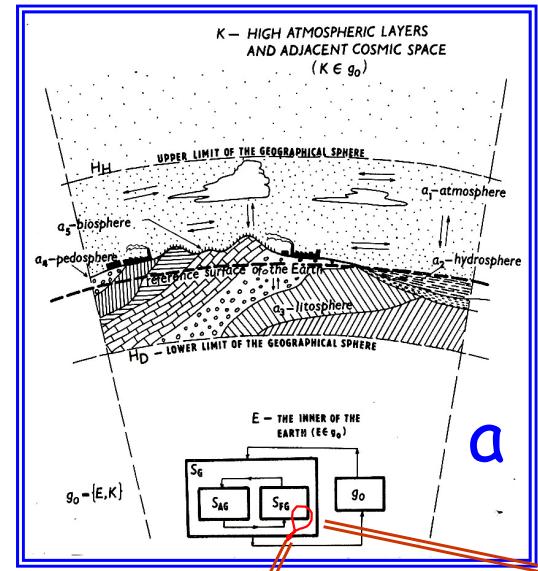
the Earth

on the Earth would average out over the long-term span of the project. The primary goal will be to model the dynamic geoid, taking into account deep tectonic mass changes (such as the continental uplift seen in the region of Hudson Bay and southern Canada) while separating out the small-scale or episodic geophysical (e.g., water table or magma) changes and weather phenomena. The means by which absolute and relative

and Mexico) and Europe specializing in satellite monitoring of the environment as well as in Colorado.

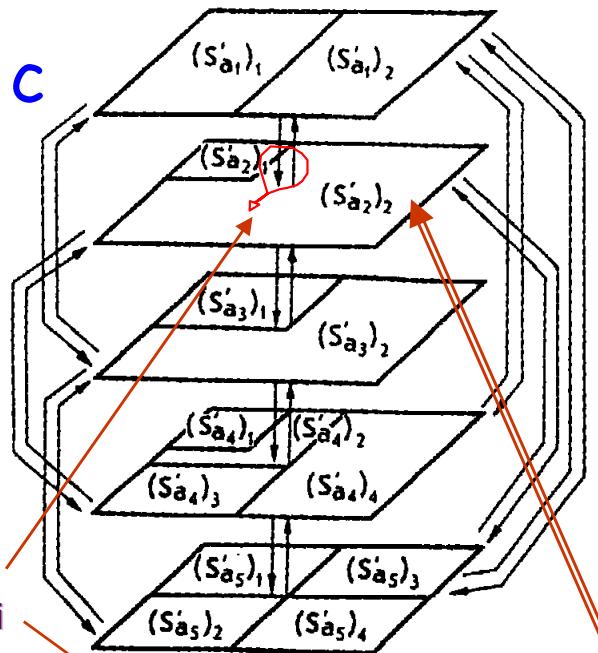


Cybernetic Model (a) for Watershed in Landscape, with Map of Conditions (b) & Multilayer Map (c)

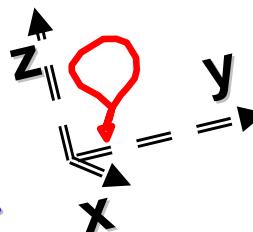
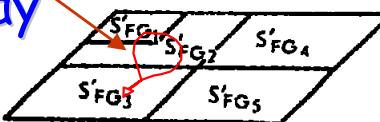


Any watershed g_{ji} for territory may be considered as a part of stream runoff system Sg_2 .

Each of these components may be characterized by matrix of input $\{W_i\}$, matrix of output $\{Q_i\}$, & matrix of states $\{H_i\}$.



System of Physical Geography Sphere (S_{FG}) with five independent elements:
 a_1 - atmosphere,
 a_2 - hydrosphere,
 a_3 - lithosphere,
 a_4 - pedosphere,
 a_5 - biosphere
(after Krcho, 1978)



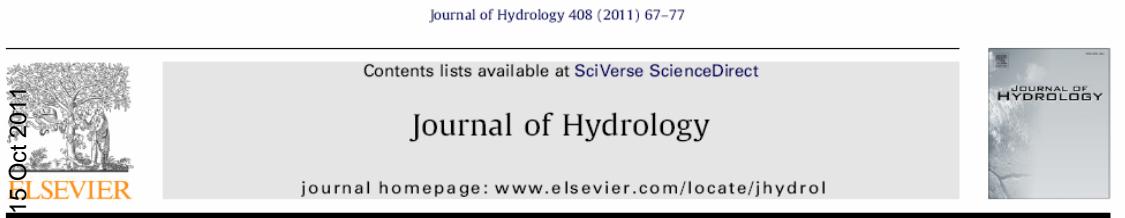
$Sg_2 = \{g_{ji}, R_{ji}\}$, where g_{ji} - watershed in specific coordinates

Math Models for the Analysis

The Duration Curves



Posted 15 Oct 2011
bysciencedirect.com
by Elsevier



Spatially smooth regional estimation of the flood frequency curve (with uncertainty)

Laio ^{a,*}, D. Ganora ^a, P. Claps ^a, G. Galeati ^b

Hydrol. Earth Syst. Sci., 15, 2805–2819, 2011
 www.hydrol-earth-syst-sci.net/15/2805/2011/
 doi:10.5194/hess-15-2805-2011
 © Author(s) 2011. CC Attribution 3.0 License.

Towards reconstruction of the flow duration curve: development of a conceptual framework with a physical basis

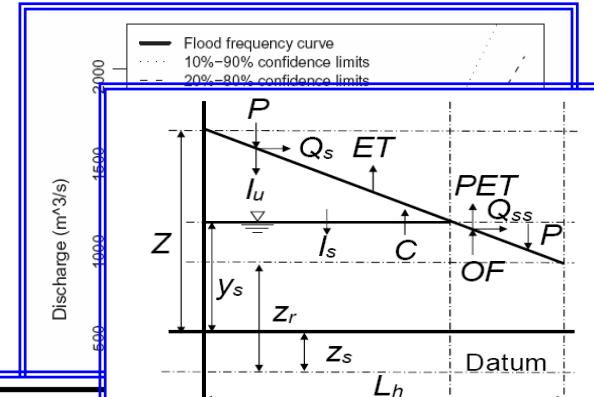
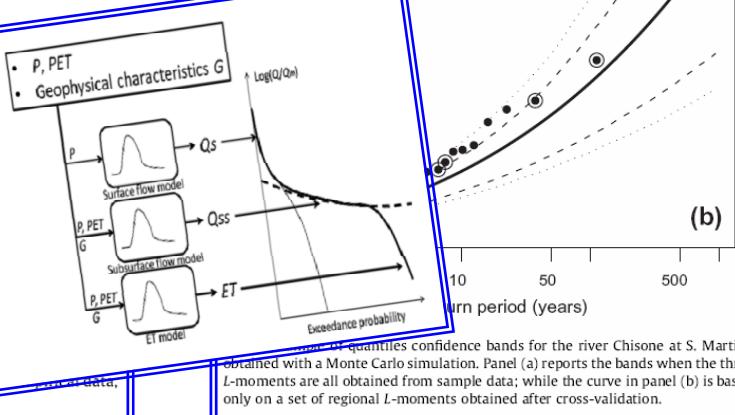
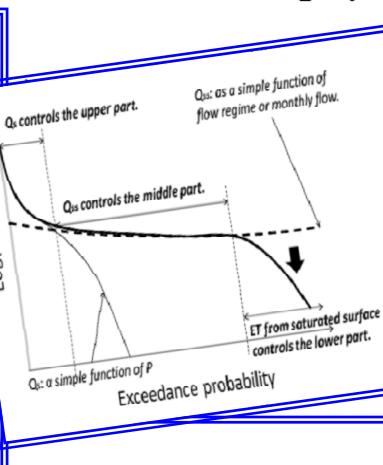
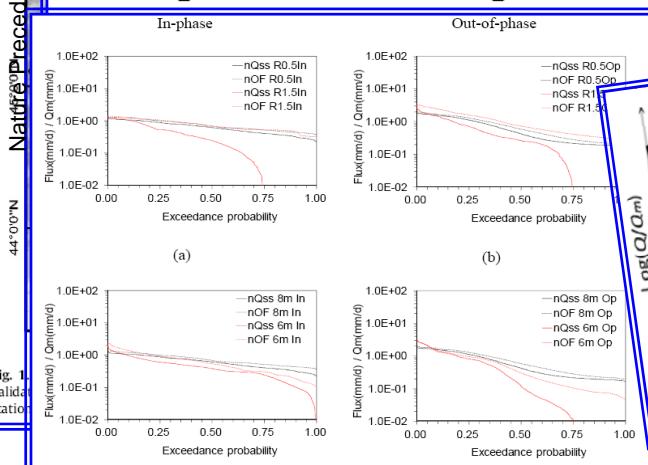


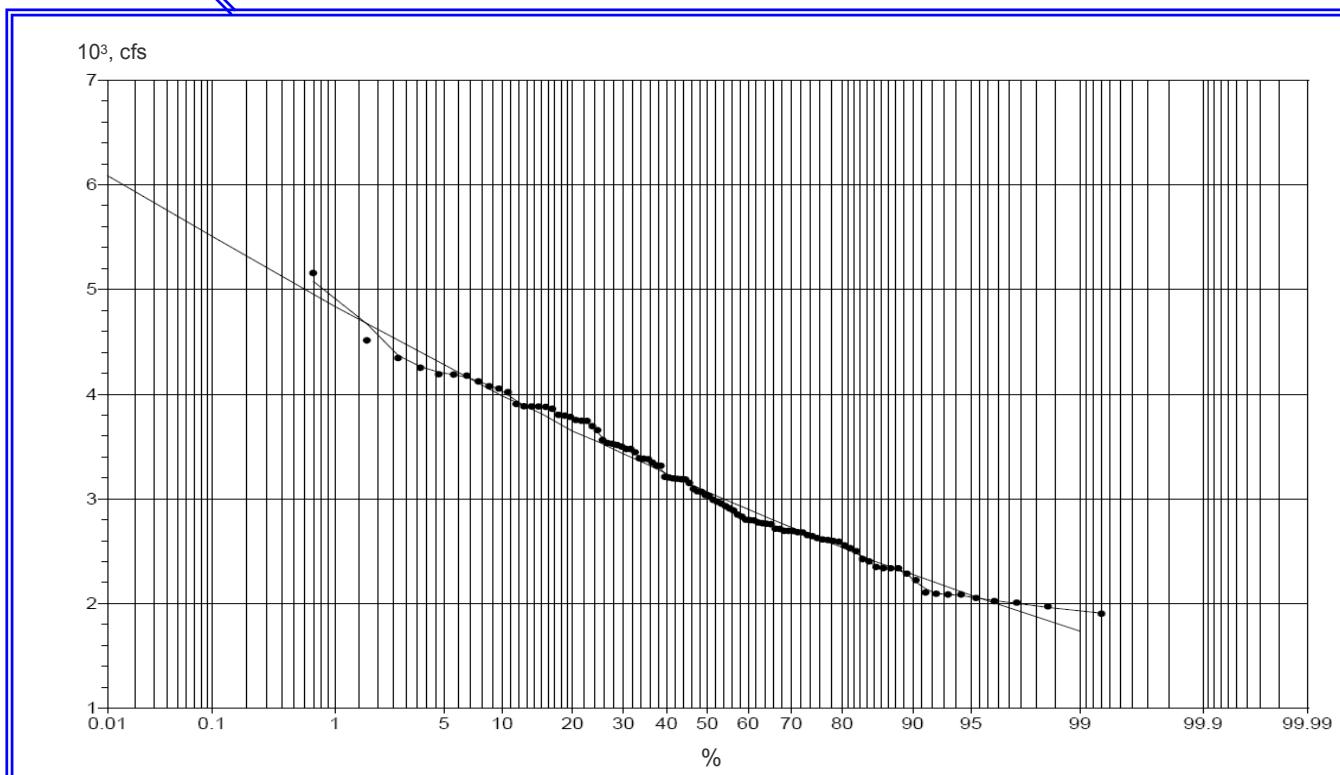
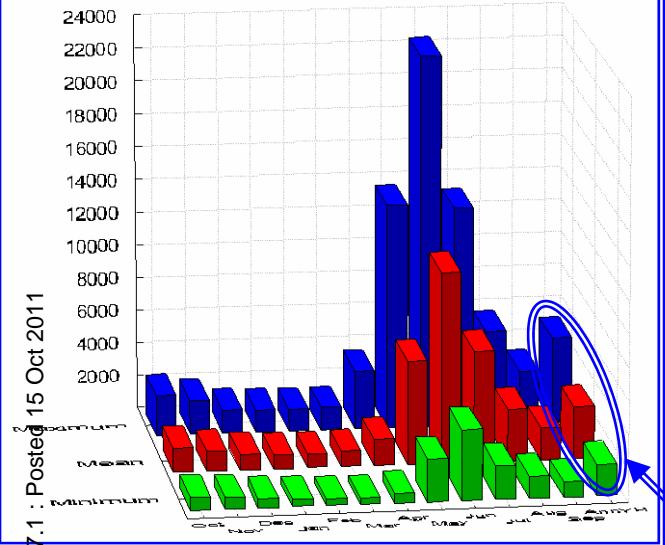
Fig. 1. Conceptual drawing of Reggiani et al.'s (2000) REW-scale water balance model: P : Precipitation, ET: Evapotranspiration, PET: Potential evapotranspiration, Q_s : Surface runoff, Q_{ss} : Subsurface runoff, I_u : Infiltration from the ground surface, I_s : Infiltration to the saturated zone, C: Capillary rise, OF: Outflow from saturated zone, Z: Average elevation of ground surface from datum, z_r : Average elevation of channel bed with respect to datum, z_s : Average elevation of the bottom surface of the REW with respect to datum, y_s : Average thickness of saturated zone, L_h : Averaged horizontal length of one side of REW.

The Duration curve

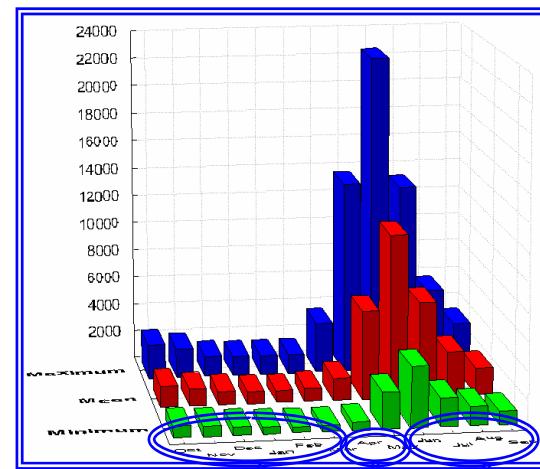
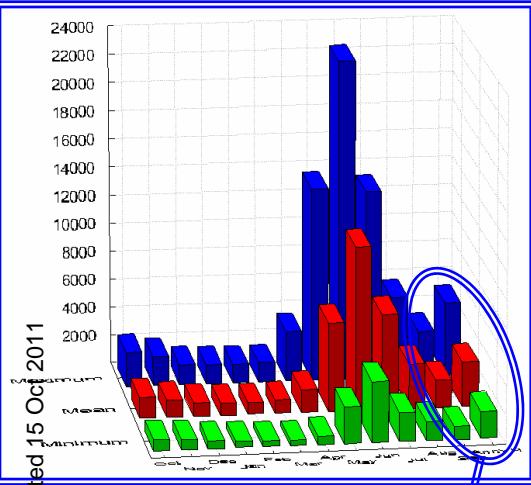
Nature Precedings : doi:10.11038/npre.2011.6537.1 : Posted 15 Oct 2011

The hydrograph of
hydrological year for
USGS 06191500
1911-2010

Empirical
durational curve
1911-2010 for
USGS 06191500
Yellowstone
River at Corwin
Springs, MT

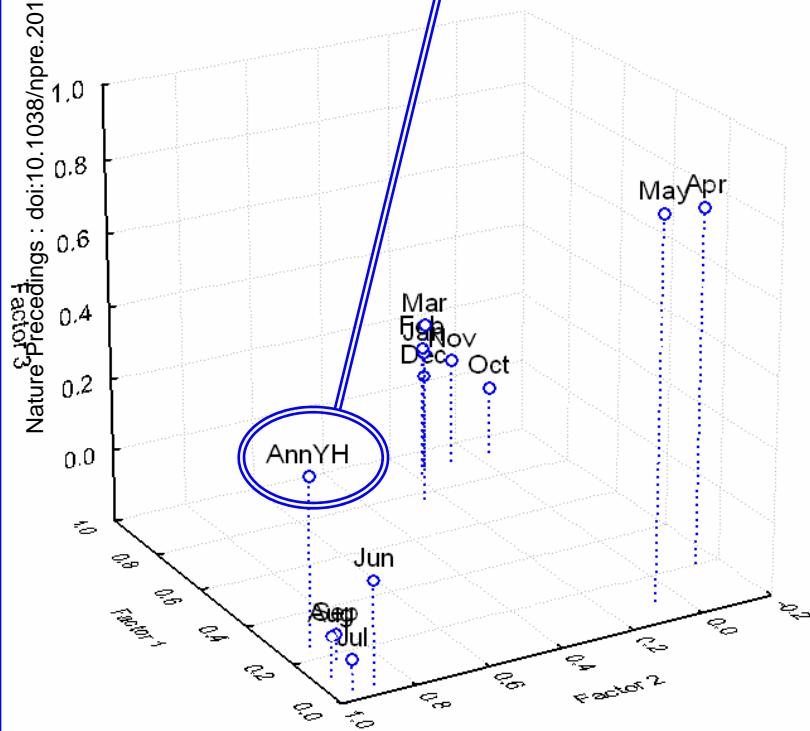


Annual distribution



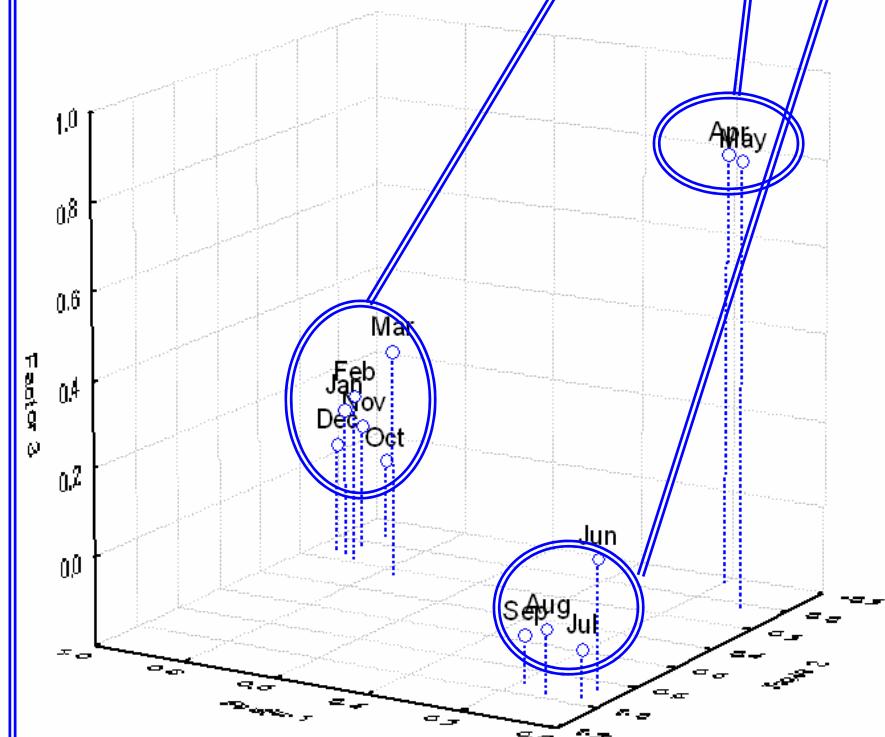
Factor Loadings, Factor 1 vs. Factor 2 vs. Factor 3

Rotation: Varimax normalized
Extraction: Principal components



Factor Loadings, Factor 1 vs. Factor 2 vs. Factor 3

Rotation: Varimax normalized
Extraction: Principal components

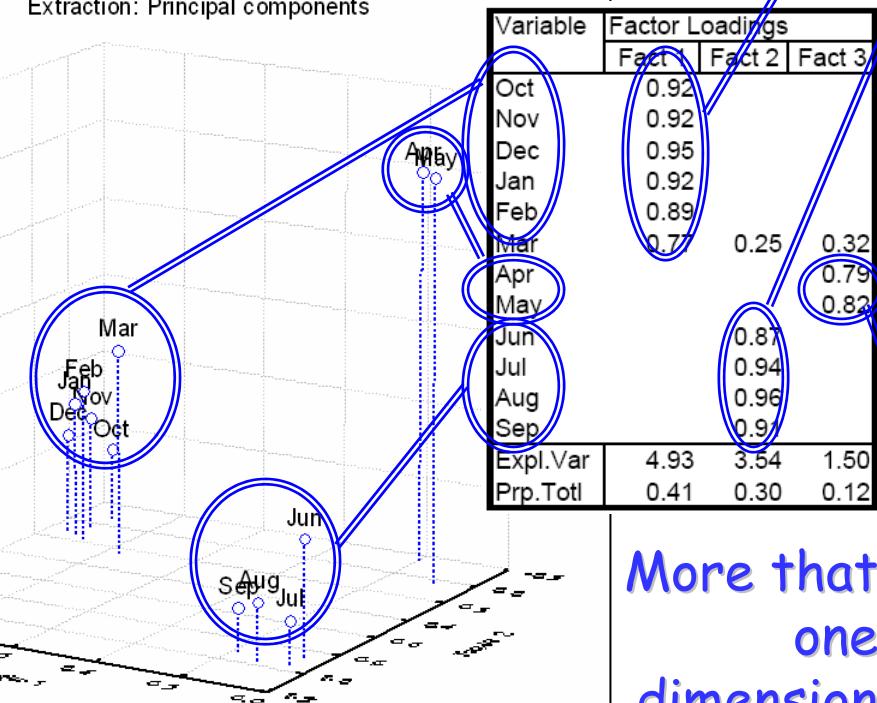


Structure of the seasonal variability

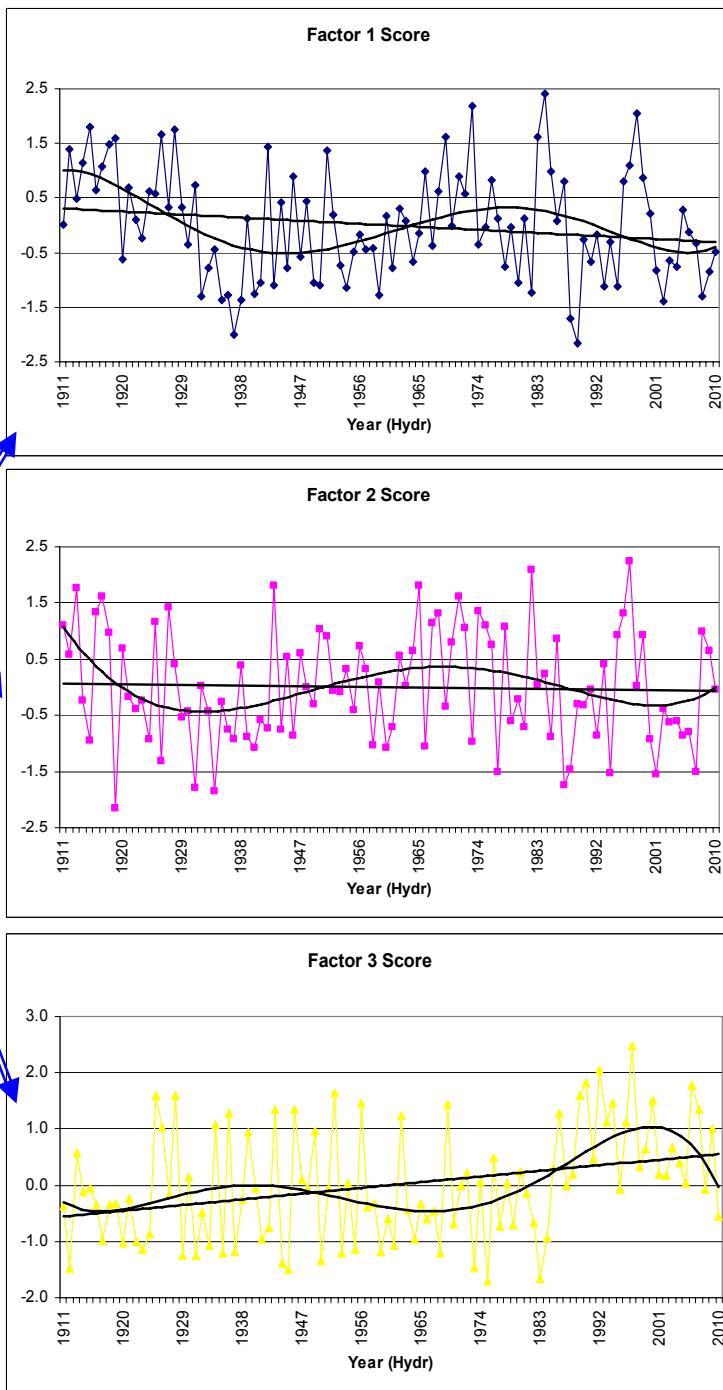
Factor Loadings, Factor 1 vs. Factor 2 vs. Factor 3

Rotation: Varimax normalized

Extraction: Principal components

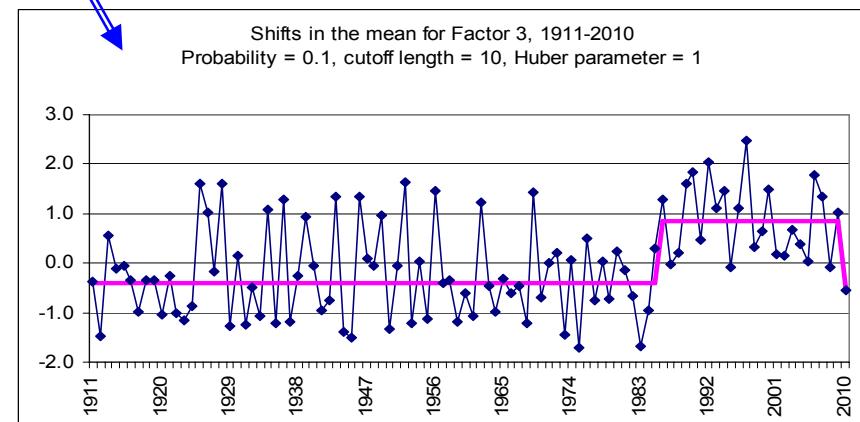
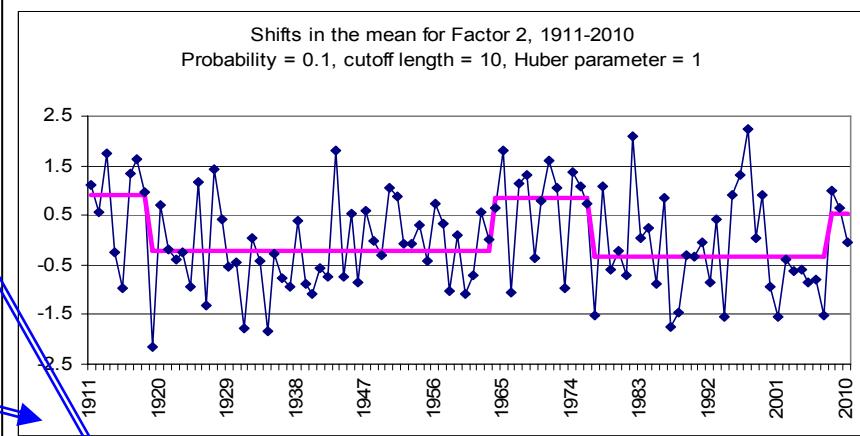
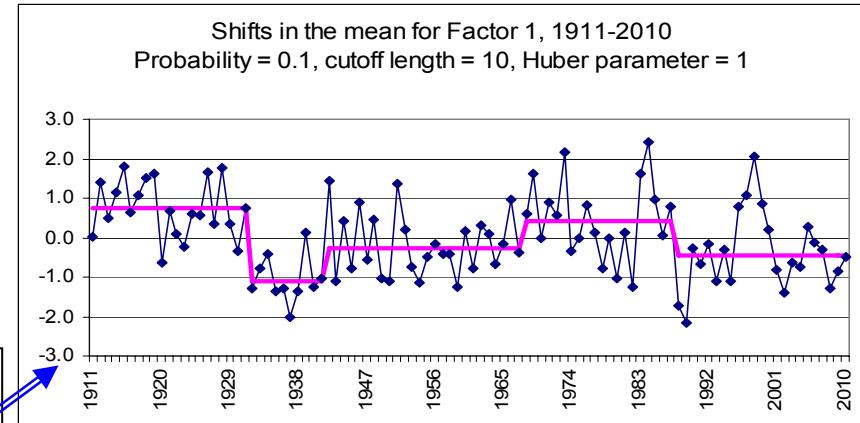
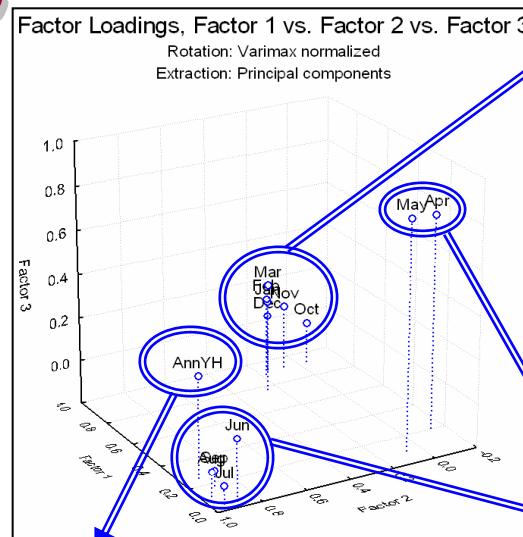


More than
one
dimension



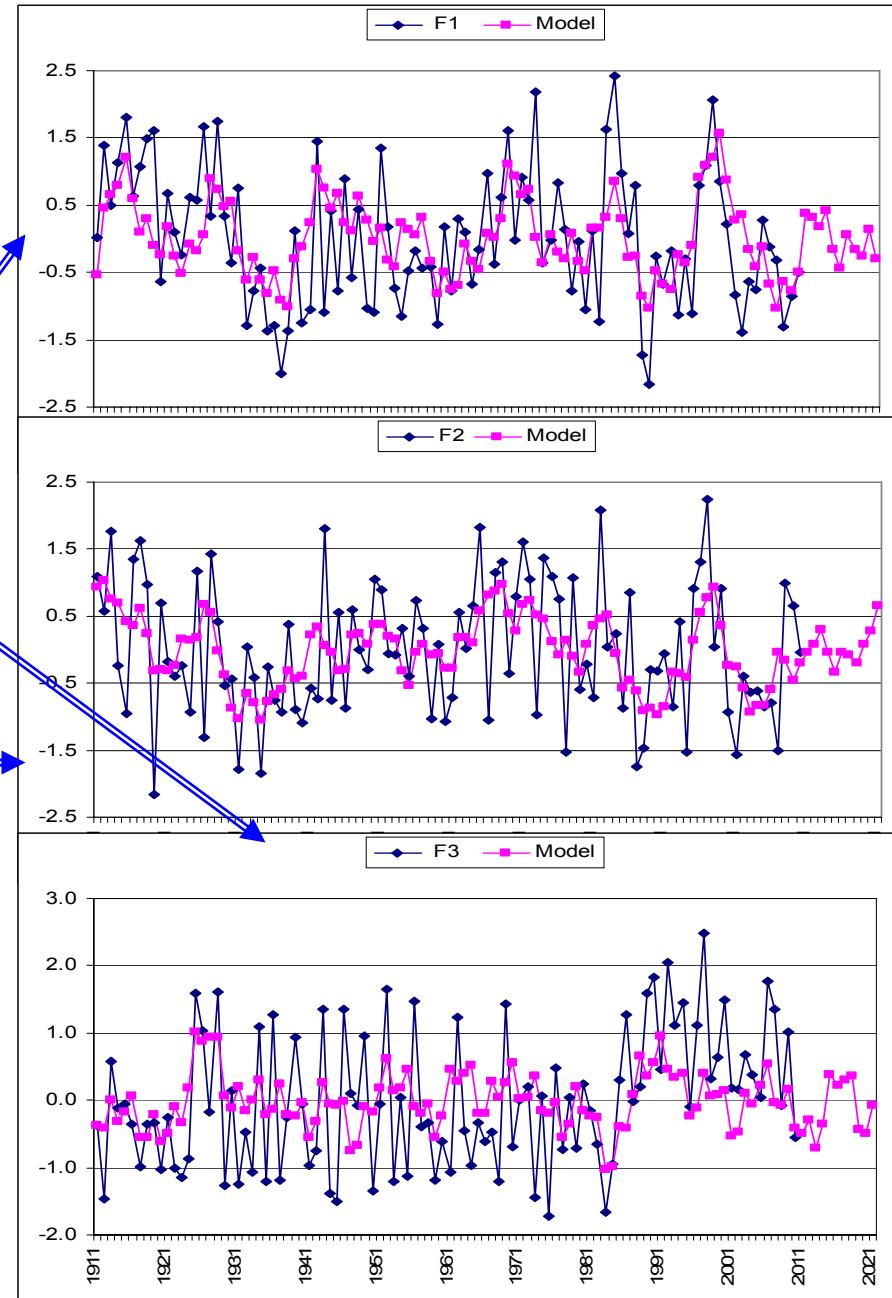
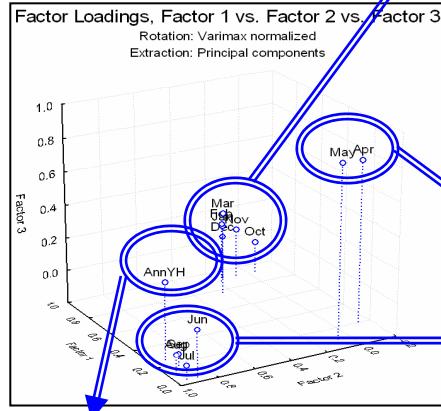
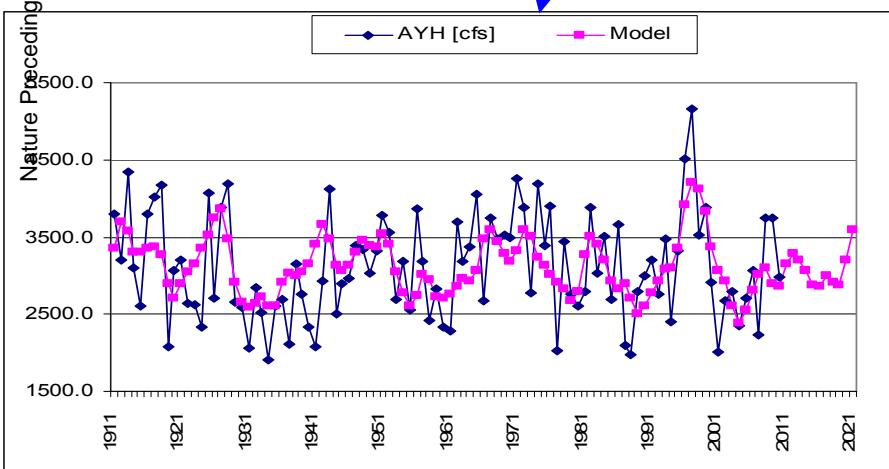
Interannual seasonal regime: shifts

Nature Precedings : doi:10.1038/npre.2011.65371 : Posted 15 Oct 2011



Modeling the Regimes 1911-2010 (2021)

doi:10.1038/npre.2011.6537;1 : Posted 15 Oct 2011



The Wavelets

Global and Planetary Change 78 (2011) 1–13



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A synthesis of the time-scale v
continuous wavelet transform

A. Rossi*, N. Massei, B. Laignel



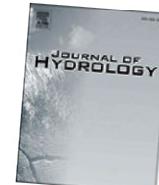
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Journal of Hydrology 376 (2009) 295–306

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Journal of Hydrology

journal homepage: www.elsevier.com/locate/jhydrol



A coupled approach of surface hydrological modelling and Wavelet Analysis for understanding the baseflow components of river discharge in karst environments

F. Salerno *, G. Tartari

Water Research Institute, National Research Council (IRSA-CNR), Località Occhiate, 20047, Brugherio, Milan, Italy
Contents lists available at ScienceDirect

Journal of Hydrology

journal homepage: www.elsevier.com/locate/jhydrol

Nature Precedings | doi:10.1608/npre.2011.6337.1 | Posted 15 Oct 2011



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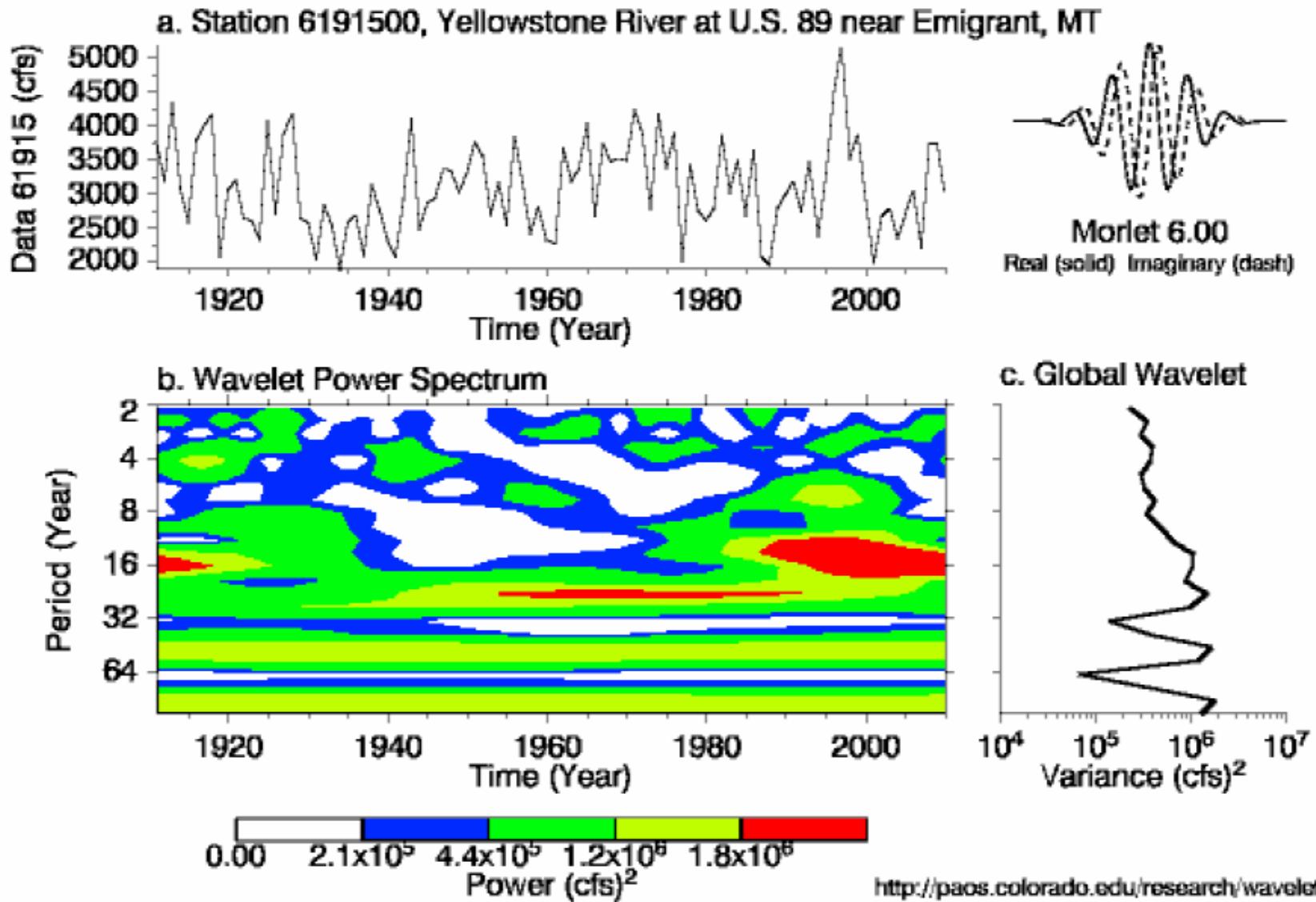
A wavelet-support vector machine conjunction model for monthly streamflow forecasting

Ozgur Kisi ^{a,*}, Mesut Cimen ^b

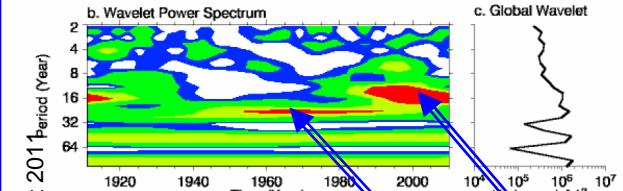
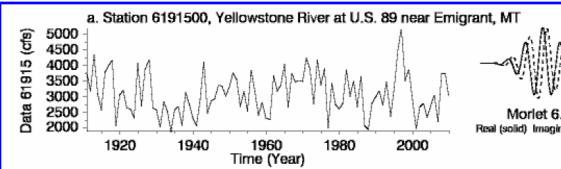
^a Erciyes University, Engineering Faculty, Civil Eng. Dept., Hydraulics Division, 38039 Kayseri, Turkey

^b Suleyman Demirel University, Engineering-Architecture Faculty, Civil Eng. Dept., Isparta, Turkey

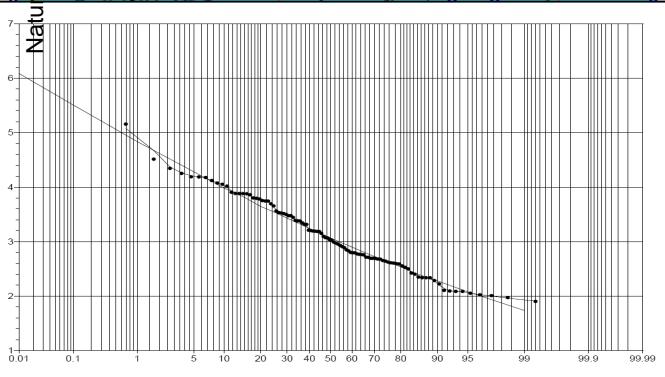
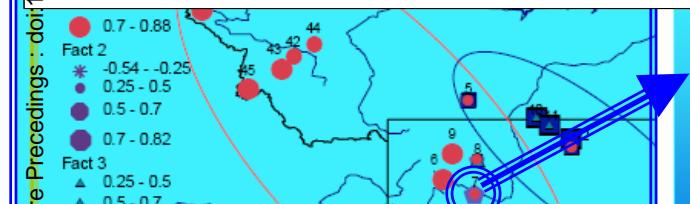
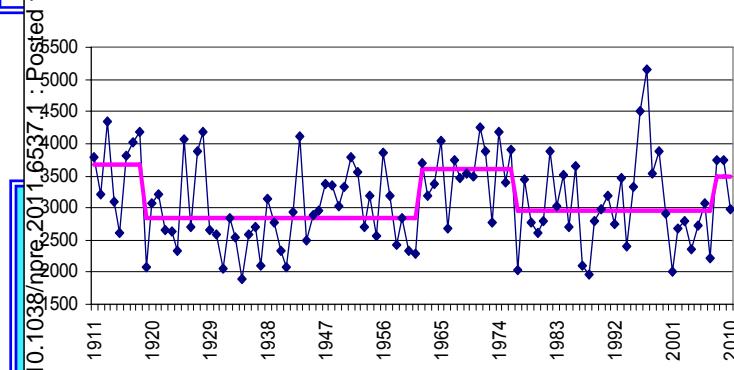
The Wavelets



Math Models &



Shifts in the mean for Annual Hydrologic Year, 1911-2010
Probability = 0.1, cutoff length = 10, Huber parameter = 1



Nature Precedings : doi:10.1038/npre.2007.1071.1 : Posted 21 Sep 2007

Chart of annual runoff 1911-2005

Model:

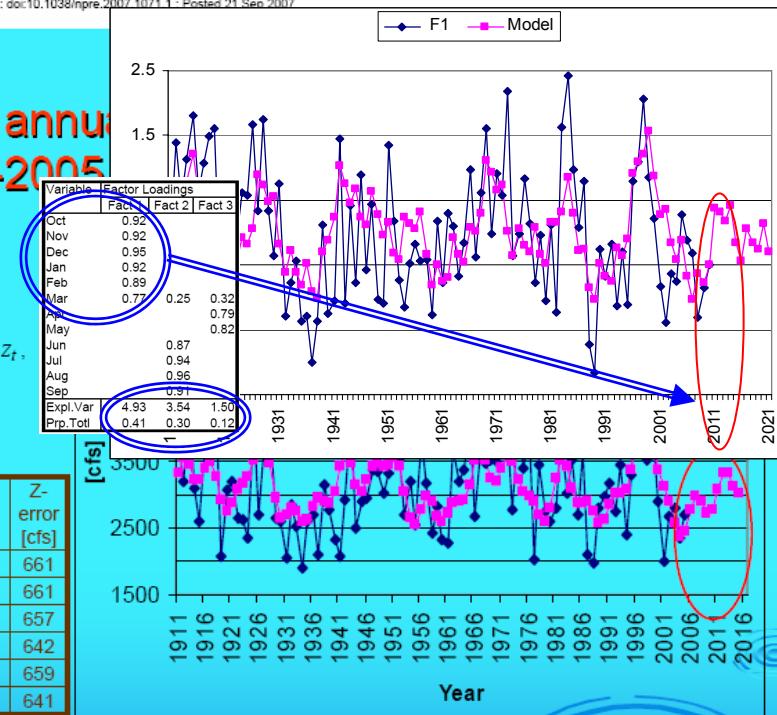
$$X_t = X_0 + \sum_{i=1}^K A_i \cos\left(\frac{2\pi}{T_i} t - \varphi_i\right) + Z_t,$$

Variable	Factor 1	Fact 2	Fact 3
Oct	0.92		
Nov	0.92		
Dec	0.95		
Jan	0.92		
Feb	0.89		
Mar	0.77	0.25	
Apr		0.32	
May		0.79	0.82
Jun		0.87	
Jul		0.94	
Aug		0.96	
Sep		0.91	
Expl.Var	4.93	3.54	1.50
Prp.Totl	0.41	0.30	0.12

No.	T-period [year]	A-amplitude [cfs]	φ -faze	Z-error [cfs]
1	5.5	175.9	2.54	661
2	8.0	187.3	0.06	661
3	12.0	165.5	1.78	657
4	14.0	245.7	1.27	642
5	17.0	171.1	1.36	659
6	25.0	321.4	2.18	641

4	Fact 5	Total
02	2.70	31.31
00	0.06	68%

Annual runoff 1963-1990



Time Variability

The Knowledge of Variability for Watershed

- * The Knowledge about watershed comes only from the analysis of the empirical data (instrumental observations)
- * Variability has to be defined as annual & seasonal structure in coordinates of the hydrologic time & space for particular watershed (coordinates have nonstationary axes, the factor's axes for multidimensional process are considered as the basis for coordinate system)
- * The math model does not have criteria to verify itself (Gödel's incompleteness theorems) & multi models & scales studies with use of empirical data have to be completed

communicating the Knowledge

The Uncertainty in Hydrology: the Usual Approach

Climatic Change (2011) 105:387–408
DOI 10.1007/s10584-010-9896-4

The role of uncertainties in the design of international water treaties: an historical perspective

Alena Drieschova · Itay Fischhendler · Mark Giordano

Table 1 Uncertainty language in transboundary water agreements, 1900–2007

Nature of uncertainty	% of sample which mentioned
Exogenous resource uncertainty	
Flow variability	49%
General environmental	13%
Scientific	4%
Explicit climate change uncertainty	0.69%
Exogenous background	
International relations	
Demand uncertainty	
Induced endogenous uncertainty	
Treaty implementation	
Data	
Treaty finance	
Treaty effectiveness	
Treaty created infrastructure	

Table 2 Changes in types of uncertainty mentioned in transboundary water agreements, 1900–2007

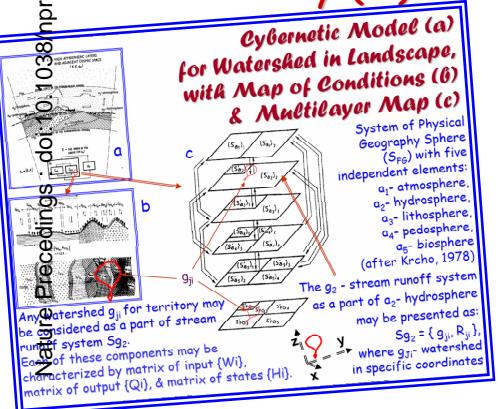
	1900–1949	1950–1969	1970–1989	1990–2007
Exogenous resource uncertainties				
Flow variability	44%	56%	41%	51 %
General environmental uncertainty	2%	6%	19%	24%
Scientific uncertainty	4%	1%	6%	6%
Explicit climate change uncertainty	0%	0%	0%	3%
Exogenous background uncertainties				
International relations	17%	4%	7%	4%
Induced endogenous uncertainties				
Implementation uncertainty	6%	7%	6%	7%
Data uncertainty	2%	0%	0%	1 %
Financial uncertainty	6%	6%	7%	4%
Effectiveness uncertainty	4%	1%	7%	4%
Infrastructural uncertainty	10%	13%	15%	28%

The Uncertainty & The Knowledge through Modeling: Object, Data, Analysis & Results

Posted 15 Oct 2011 08:37:11

Photo picture as presentation of the natural object

The knowledge (K) = 0,
about a new object for
the consideration
the uncertainty (U) = 1



K_p = 1 & we have the direction for the research, the task, U = 0, but the Knowledge is previous (K_p)

From Data Analysis to Statistical Learning

THE PHILOSOPHY OF EXPLORATORY DATA ANALYSIS*
I. L. GOOD!
Statistics Department
Virginia Polytechnic Institute and State University

**J. R. Statist. Socie. A (1995)
158, Part 3, pp. 419-466**

Model Uncertainty, Data Mining and Statistical Inference
By CHRIS CHAFFIELD!
University of Bath, UK

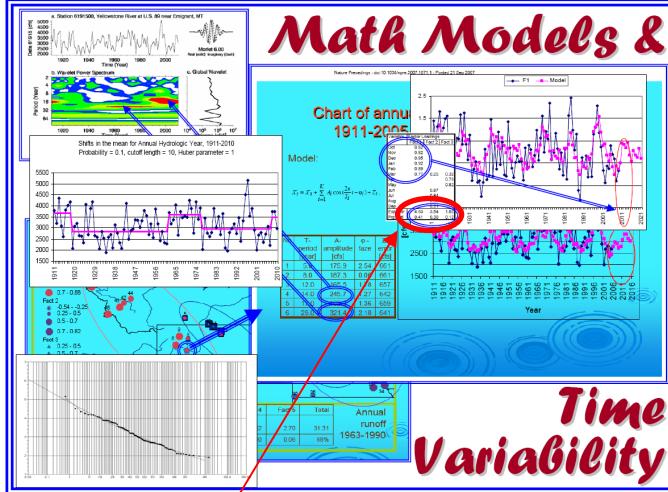
Read Before the Royal Statistical Society on Wednesday, January 18th, 1995, the President, Professor D. J. Hand, President of the Chair

Summary
This paper takes a broad, pragmatic view of statistical inference to include all aspects of information. The estimation of model parameters traditionally assumes that there is a true model and that the data set is representative of it. This is implicitly assumed in the model structure. This implicitly assumes the existence of a 'true' model, which may not represent it as it is. In other words, uncertainty is not part of the model structure. This is true whether the model is specified on subject-matter knowledge or from data. The term *data mining* has been used to describe the process of extracting useful information from data sets. It is a broad term used in many areas of statistics. This is true whether the model is specified on subject-matter knowledge or from data. The term *data mining* has been used to describe the process of extracting useful information from data sets. It is a broad term used in many areas of statistics. The term *data mining* may be used in this paper to mean the process of extracting useful information from data sets. The term *model uncertainty*, such as too narrow prediction intervals, and the non-trivial parameter estimates which can follow data-based modelling. Ways of assessing the uncertainty of the model are discussed. These include cross-validation methods, a Bayesian model averaging approach and collecting additional data. Finally, the main aim of the paper is to ensure that statisticians are aware of the problems and start addressing the issues even if there is no simple general theoretical fix.

The Elements of Statistical Learning
Data Mining, Inference, and Prediction
Second Edition
Trevor Hastie, Robert Tibshirani, Jerome Friedman

The Statistical Learning is the way to obtain ("extract") the structure of a natural object

The conceptual model (Cybernetic Model) is the way to use previously obtained Knowledge



After Statistical Learning
 $K > U$

The Uncertainty from Analysis obtained for every model.
For Factor Analysis
 $U=1$ - explained variability

Communicating the Knowledge for the Watershed

Scientist working in Hydrology have to handle the Uncertainty & communicate the Knowledge about time-spatial variability of the Watershed characteristics

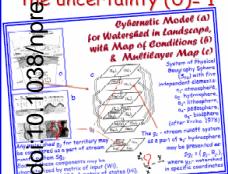
Nature Precedings doi:10.1038/npre2011-6537-1 Posted 15 Oct 2011

The Uncertainty & The Knowledge through Modeling: Object, Data, Analysis & Results



Photo picture as presentation of the natural object

The knowledge ($K=0$, about a new object for the consideration) the uncertainty ($U=1$)

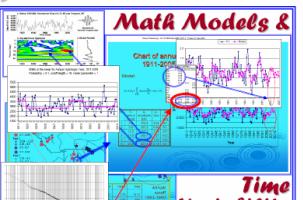


$K=1$ & we have the direction for the research, the task, $U=0$, but the Knowledge is previous (K_p)

From Data Analysis to Statistical Learning



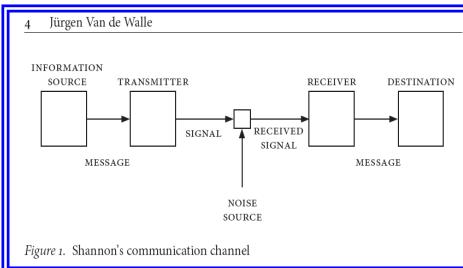
The Statistical Learning is the way to obtain ("extract") the structure of a natural object



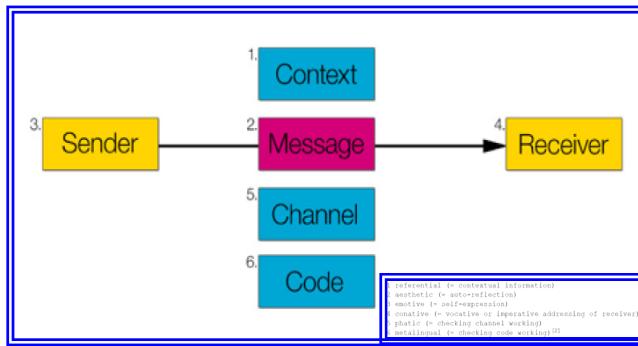
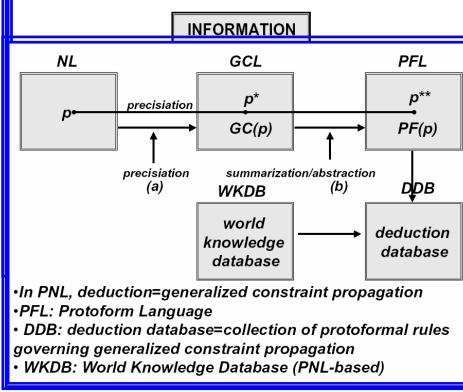
After Statistical Learning
 $K > U$

The conceptual model (Cybernetic Model) is the way to use previously obtained Knowledge

The Uncertainty from Analysis obtained for every model.
For Factor Analysis
 $U=1$ - explained variability



L.A. Zadeh / Information Sciences 172 (2005) 1–40



Conclusions

- * The Knowledge & the Uncertainty obtained with math models help to determine:
 - the scope of practical applications to be developed (like water balance estimations for conservation &/or management of water resources in different scales),
 - the tasks to educate the public/communities about water resources & environmental issues.
- * Communication of the knowledge of hydrologic objects & processes is successful with colleagues with equal experience.
 - * Education has be based on research.