Sub-grid parameterization of snow distribution for an energy and mass balance snow cover model

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Abstract:

Representation of sub-element scale variability in snow accumulation and ablation is increasingly recognized as important in distributed hydrologic modelling. Representing sub-grid scale variability may be accomplished through numerical integration of a nested grid or through a lumped modelling approach. We present a physically based model of the lumped snowpack mass and energy balance applied to a 26-ha rangeland catchment with high spatial variability in snow accumulation and melt. Model state variables are snow-covered area average snow energy content (U), the basin-average snow water equivalence (W_n) , and snow-covered area fraction (A_t) . The energy state variable is evolved through an energy balance. The snow water equivalence state variable is evolved through a mass balance, and the area state variable is updated according to an empirically derived relationship, $A_i(W_a)$, that is similar in nature to depletion curves used in existing empirical basin snowmelt models. As snow accumulates, the snow covered area increases rapidly. As the snowpack ablates, A_{ℓ} decreases as W_a decreases. This paper shows how the relationship $A_1(W_a)$ for the melt season can be estimated from the distribution of snow water equivalence at peak accumulation in the area being modelled. We show that the depletion curve estimated from the snow distribution of peak accumulation at the Upper Sheep Creek sub-basin of Reynolds Creek Experimental Watershed compares well against the observed depletion data as well as modelled depletion data from an explicit spatially distributed energy balance model. Comparisons of basin average snow water equivalence between the lumped model and spatially distributed model show good agreement. Comparisons to observed snow water equivalence show poorer but still reasonable agreement. The sub-grid parameterization is easily portable to other physically based point snowmelt models. It has potential application for use in hydrologic and climate models covering large areas with large model elements, where a computationally inexpensive parameterization of sub-grid snow processes may be important. Copyright © 1999 John Wiley & Sons, Ltd.

KEY WORDS sub-grid parameterization; snow distribution; snow water equivalence; energy balance model

INTRODUCTION

Within the last 20 years, interest in scaling has increased within the hydrologic research community. The increase in interest has been driven in part by a desire to apply physically based hydrologic models to catchments and global circulation model (GCM) grid cells. Snowmelt has been an important hydrologic process examined with respect to scaling. In mountainous regions, snowmelt is one of the largest surface water inputs controlling runoff. Snow cover affects the atmosphere through its strong influence on the surface radiation and energy balance.

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At the catchment and GCM scale, interest lies in determining the effects of changing land use. Some of the interest of atmospheric modellers relates to the effects of changing climate. Therefore, empirical hydrologic models, may not be useful at these larger scales (Seyfried and Wilcox, 1995). Empirical models require calibration under particular conditions. If land use or general climate conditions change, the correlations may not necessarily be the same. Physically based models use parameters that are, at least in principle, related to physical conditions and can sometimes be measured.

Physically based models tend to have a foundation in point-scale research. Much research of snowmelt processes has been conducted at the plot or point scale (Hathaway *et al.*, 1956; Anderson, 1976; Morris, 1986; Morris, 1990; Jordan, 1991; Tarboton *et al.*, 1995; Tarboton and Luce, 1997). Point scale models are often not applicable for larger areas even using effective parameters calibrated for the catchment (Arola and Lettenmaier, 1996; Luce *et al.*, 1997, 1998). Snowmelt shares this characteristic with other hydrologic processes (Beven, 1995; Kalma and Sivapalan, 1995). Generalized solutions available to solve the problem are numerical integration (Abbot and Refsgaard, 1996), spatial distribution functions (Moore, 1985), and parameterizations (Beven, 1995; Blöschl, 1996).

For snowmelt, several solutions have been applied. Numerical integration has been and continues to be a popular method through use of distributed hydrologic models (Blöschl et al., 1991; Wigmosta et al., 1994; Kirnbauer et al., 1994; Liston, 1997). The distribution function approach has also been applied to create lumped models (Horne and Kavvas, 1997; Anderson, 1973; Rango and Van Katwijk, 1990; Martinec et al., 1994). The areal depletion curve concept, which amounts to a distribution function approach, has been applied to empirical models including the National Weather Service River Forecasting System (NWSRFS) and the Snowmelt Runoff Model (Martinec et al., 1994). This approach can be adopted into a physically based modelling framework by developing a relationship between the state variable of interest, snow water equivalence over the basin, and the areal extent of snowcover. This is similar to the approach of TOPMODEL (Beven and Kirkby, 1979) where saturated area is estimated as a function of the basinaveraged stored water. In TOPMODEL, the parameterization relating the stored water state variable to the saturated area is derived based on an analysis of the topography. With the lumped snowmelt model, the parameterization relating the basin averaged snow water equivalence state variable to the fractional snow coverage is derived based on a probability distribution/density function (pdf) of peak snow water equivalence. Other remote sensing and modeling tools (Elder et al., 1989, 1991, 1995, 1998; Elder, 1995; Rosenthal and Dozier, 1996) may be used to relate topography to the pdf of peak snow water equivalence.

The objectives of this paper are to present and test a physically based lumped model of snowpack evolution for a small watershed (26 ha) that uses a depletion curve parameterization to relate the basin-averaged snow water equivalence to snow-covered area. A secondary purpose is to present a method for deriving the depletion curve from snowpack measurements at peak accumulation. This is part of an ongoing effort to extend physically based modelling methodology to larger scales where it is impractical to apply a point model over a grid of model elements small enough to ignore subgrid variability.

METHODS

The basic approach of this study compared the outputs of the lumped model to outputs from a distributed snowmelt model and a series of distributed snow water equivalence observations. The lumped model treats the study area (26 ha) as a single model element with subgrid variability parameterized through a depletion curve. The distributed model was applied at a 30 m grid scale ignoring only subgrid variability smaller than this scale and amounts to a numerical integration of the spatially distributed processes that are parameterized in the lumped model. The depletion curve parameterization used for the lumped simulation was derived from observations of the snow water equivalence pattern near the time of peak accumulation. This depletion curve was compared to that derived from the series of distributed observations and from the output of the distributed model.



Figure 1. Map of northwestern United States showing approximate region of study watershed

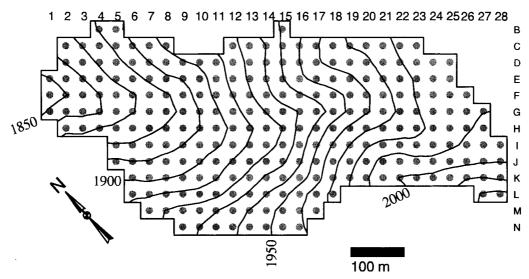


Figure 2. Map of Upper Sheep Creek with snow survey grid. Contour interval is 10 m

Study area and observations

Snow survey and climatological data from the Upper Sheep Creek subbasin of the Reynolds Creek Experimental Watershed in southwestern Idaho (Figure 1) form the observational basis of this study. The Upper Sheep Creek watershed has an area of 26 ha and ranges between 1840 and 2040 m elevation (Figure 2). Low sagebrush (*Artemisia arbuscula* Nutt.) communities cover the northeast portion of the basin, and big sagebrush (*Artemisia tridentata* Nutt.) communities cover much of the southwestern half of the basin. Aspen (*Populus tremuloides* Michx.) grow in a narrow strip along the northeast-facing slope where snow drifts typically form (Figure 3). Severe winter weather and winds prevent the aspen from growing to heights greater

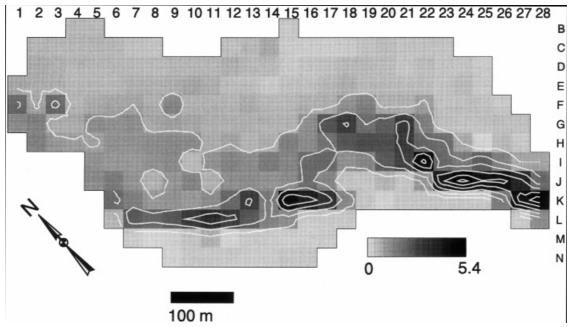


Figure 3. Map of drift factors calibrated at Upper Sheep Creek based on 1993 observations. Darker areas are areas of greater snow accumulation and greater drift factor values. Contour interval is 0.8

than 4–7 m. Average annual precipitation is 508 mm, and the first-order stream exiting the basin is ephemeral.

Upper Sheep Creek has been the site of many previous hydrologic investigations (Stephenson and Freeze, 1974; Cooley, 1988; Duffy et al., 1991; Flerchinger et al., 1992, 1994; Jackson, 1994; Seyfried and Wilcox, 1995; Tarboton et al., 1995; Luce et al., 1997, 1998; among others). Runoff generation has been the focus of much of the work, and all of the studies have noted the importance of the wind-induced snowdrift in the southwest portion of the basin to the basin hydrology. Previous work (Luce et al., 1997, 1998) has shown that snow drifting is the primary determinant of spatial variability of snow in this watershed, more important than topographically induced variations of radiation. Previous work measuring snow drifting (Cooley, 1988) and distributed snowmelt modeling (Jackson, 1994; Tarboton et al., 1995; Luce et al., 1997, 1998) has provided both foundation and incentive for development of a lumped snowmelt model of the basin that parameterizes the subgrid variability due to snow drifting and spatially variable radiation processes.

The data used in this paper comprise measurements of snow water equivalence taken on nine dates in 1993 with a snow tube and scale. A systematic grid sampling strategy was used throughout the watershed (Figure 2). The grid spacing was 30.48 m (100 ft), and the long axis was oriented 48 degrees west of north. Precipitation, temperature, relative humidity, and incoming solar radiation were measured for water year 1993 at a weather station near loction J 10. Wind speed was measured at D 3.

Distributed point model

The distributed model is a cell-by-cell execution of the Utah Energy Balance (UEB) snowpack energy and mass balance model (Tarboton *et al.*, 1995; Tarboton and Luce, 1997). In order to run the model in a distributed fashion, climatic inputs (radiation and precipitation) were calculated individually for each cell based on measurements from the weather station, topography, and a calibrated drift factor.

UEB is an energy and mass balance model with a vertically lumped representation of the snowpack. A schematic is shown in Figure 4a. Two primary state variables are maintained in the model, snow water

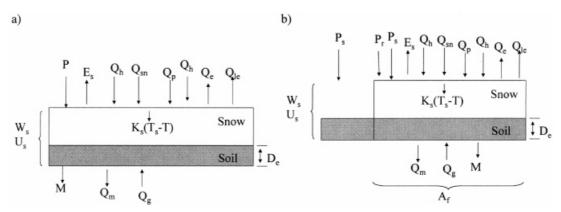


Figure 4. Schematic diagrams of a) Utah Energy Balance point scale snowmelt model and b) the lumped snowmelt model

equivalence, W [m], and internal energy of the snowpack and top 40 cm of soil, U [kJ m⁻²]. U is zero when the snowpack is at 0 °C and contains no liquid water. These two state variables are updated according to

$$dU/dt = Q_{sn} + Q_{li} - Q_{le} + Q_p + Q_g + Q_h + Q_e - Q_m$$
 (1)

$$dW/dt = P_r + P_s - M_r - E (2)$$

where Q_{sn} is net solar radiation; Q_{li} is incoming longwave radiation; Q_{le} is outgoing longwave radiation; Q_p is advected heat from precipitation; Q_g is ground heat flux; Q_h is the sensible heat flux; Q_e is the latent heat flux; Q_m is heat advected with melt water; P_r is the rate of precipitation as rain; P_s is the rate of precipitation as snow; M_r is the melt rate; and E is the sublimation rate. The model is driven by inputs of precipitation, air temperature, humidity, wind speed and incoming solar radiation. Snow surface temperature, a key variable in calculating latent and sensible heat fluxes and outgoing longwave radiation, is calculated from the energy balance at the surface of the snowpack, where incoming and outgoing fluxes must match. These simulations were run on a six-hour time step.

The effect of plant canopy on snowmelt is parameterized by decreasing the albedo of the snow surface as the snow depth decreases below the canopy height. This parameterization is most appropriate for short vegetation, such as sagebrush. Because the aspens are free of leaves until the soil warms slightly, errors introduced by not considering the taller canopy are minimal.

The distributed model uses a drift multiplier to estimate enhancement of local incoming snow at each cell through wind transport. The fraction of precipitation falling as rain or snow is a function of temperature. The fraction of the gauge catch falling as snow is multiplied by the drift multiplier to estimate grid cell precipitation. The drift multiplier was calibrated at each grid cell to minimize the mean square error of the point model relative to observations on 10 February, 3 March, and 23 March, 1993. Values of the multiplier over the basin are shown in Figure 3 and ranged from 0.16 to 5.36, with an average of 0.928.

Temporal variations in solar radiation were estimated based on an average atmospheric transmission factor calculated from pyranometer data at the weather station. Local horizons, slope, and azimuth were used to find local sunrise and sunset times and to integrate solar radiation received on the slope of each grid cell during each time step. The calculated atmospheric transmission factor characterized cloudiness for incoming longwave radiation calculations.

Lumped model with depletion curve parameterization

Figure 4b depicts schematically the lumped model with subgrid parameterization using depletion curves. This is a modification of the UEB point-model (Figure 4a) described above. The snow-covered area fraction, A_f , is introduced as a new state variable, and the basin or element average snow water equivalence,

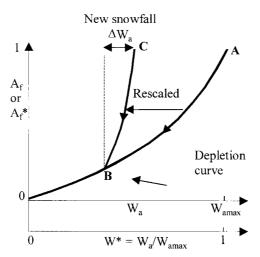


Figure 5. Schematic of depletion curve in lumped snowmelt model

 $W_a = W_s A_f$, is used as the mass state variable. The point snowmelt model is driven by basin averaged climate inputs to calculate fluxes to and from this fractional area. Because there is only one meteorological station at Upper Sheep Creek, basin average inputs were calculated from a single meteorological station and topographic information. A_f is adjusted after each time step, based on changes in W_a . During accumulation A_f increases to full cover quickly with initial snowfall, and stays at full cover until melt begins. During melt, as W_a decreases, A_f is decreased following a depletion curve (Figure 5), $A_f(W_a)$, starting from a point of maximum accumulation, A towards B.

When there is new snowfall part of the way along, for example at point B, W_a is incremented by the new snowfall water equivalence ΔW (taken over the whole area) and A_f goes to one (point C in Figure 5). The new snowfall (covering the whole element) will be subjected to the same processes that led to spatial variability in the old snow, and the new snow will melt first. Therefore, we assume the system returns along a rescaled depletion curve to the point of original departure, B. In this fashion multiple accumulation and ablation periods can be accommodated. In principle there could be multiple rescalings and multiple points B, reminiscent of hysteresis loops in soil wetting and drying characteristic curves. However, our code kept track of only one departure (point B) at any given time.

Snow accumulation may vary between years, theoretically requiring a different depletion curve for each year dependent on the peak accumulation of the year. The spatial pattern, however, is relatively consistent. Therefore, we used a single dimensionless depletion curve, scaled by the maximum snow water equivalence (W_{amax}) since W_a was last 0 (generally the beginning of the snow season). This provides scaling of the depletion curve, letting the onset of melt be determined naturally from the modelling of physical processes, rather than using parameters determining the 'beginning' of the melt season. It allows for melt episodes during the accumulation season and accumulation episodes during the melt season. The following equation gives a particular depletion curve, $A_f(W_a)$, in terms of the dimensionless depletion curve.

$$A_f(W_a) = A_f^*(W_a/W_{amax}) \tag{3}$$

Snowfall inputs to the lumped model are adjusted by an element (basin) average drift factor to account for the fact that even at the larger lumped model element scale, drifting and differences between the basin average precipitation and gauge precipitation may affect the net snow accumulation. In the results reported here the basin average drift factor, 0.928, was used.

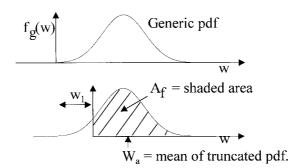


Figure 6. Schematic of generic snow water equivalence probability distribution

Depletion curves

The depletion curve represents the functional decrease of snow-covered area fraction, A_f , with decreasing basin-average snow water equivalence, W_a , through the melt season. This can be viewed as a parameterization of the distribution of snow over the basin. Note that this definition of a depletion curve differs somewhat with the classical definition of A_f as a function of melt, so requires some description on how such curves may be estimated.

Spatial heterogeneity in snowpack water equivalence is linked to spatial variability in topography and vegetation, which control relative accumulation and melt. Topography controls relative accumulation through elevational temperature effects (precipitation as rain or snow) and drifting and controls melt through elevational temperature effects and exposure to sunlight (Dozier, 1979; Dozier and Frew, 1990). Vegetation controls accumulation through effects on drifting and interception and controls melt through effects on solar radiation, wind, and temperature. The primary drivers in variability change with scale (Seyfried and Wilcox, 1995). Luce *et al.* (1997, 1998) found that the primary control on the spatial distribution of snow water equivalence in Upper Sheep Creek was drifting. In larger basins, variations in wind, temperature, or solar exposure could be important sources of variability in melt. Drifting exerts its influence during the accumulation season. This suggests that the depletion curve for Upper Sheep Creek would be related to the distribution of snow water equivalence during the peak accumulation.

To formally develop this relationship, assume a generic probability distribution (pdf) for snow water equivalence, $f_g(w)$, that retains a consistent shape through the melt season. The implication is spatially uniform melt. This pdf gives the probability for point snow water equivalence areally sampled, offset by an additive constant. As the snow accumulates and ablates this function shifts to the right or left. This procedure is shown in Figure 6, and is conceptually similar to a procedure suggested in Dunne and Leopold (1978) but generalized to non-Gaussian pdfs. The positioning of the generic pdf is controlled by the parameter w_1 , which represents the amount of melt that has occurred. The tail to the left of the y-axis represents snow free area, for any particular melt depth, w_1 . The snow-covered area fraction in terms of this pdf is defined as:

$$A_f(w_1) = \int_0^\infty f_g(w + w_1) dw = \int_{w_1}^\infty f_g(w) dw = 1 - F_g(w_1)$$
 (4)

where $F_g(w_1)$ is the cumulative density function evaluated at w_1 . For any arbitrary w_1 , $A_f(w_1)$ is the fraction of the basin with snow water equivalence at peak accumulation greater than w_1 . Practically, the function, $A_f(w)$, may be numerically evaluated directly from a sample of snow water equivalence values across the area of interest. (Note: This function, $A_f(w)$ is not the same as the depletion curve, $A_f(W_a)$, the difference being indicated through a lower case dummy argument, w or w_1 .)

The probability distribution of snow water equivalence for any particular w_1 has a nugget at zero because a negative snow water equivalence has no physical interpretation. This nugget can be represented mathematically with a Dirac delta function, so that the finite probability of the areally sampled snow water

equivalence being zero is $1-A_f(w_1)$. The part of the pdf to the right of the axis represents the snow water equivalence pdf for non-zero snow water equivalence (all of the snow-covered points in the areal sampling). Consequently, the basin-average snow water equivalence is defined (from the usual definition of a mean) as:

$$W_a(w_1) = \int_{w_1}^{\infty} (w - w_1) f_g(w) dw = \int_{w_1}^{\infty} (w) f_g(w) dw - w_1 A_f(w_1)$$
 (5)

Now recognize from (4) that

$$f_g(w) = -\frac{dA_f}{dw} \tag{6}$$

Therefore,

$$W_a(w_1) = -\int_{w_1}^{\infty} (w) \frac{dA_f}{dw} dw - w_1 A_f(w_1)$$
 (7)

Integrating by parts

$$W_a(w_1) = -[wA_f(w)]_{w_1}^{\infty} + \int_{w_1}^{\infty} A_f(w)dw - w_1 A_f(w_1)$$
(8)

Because $A_t(w) = 0$ for w greater than the maximum point snow water equivalence in the basin,

$$\lim_{w \to \infty} w A_f(w) = 0 \tag{9}$$

and the first and third terms of equation (8) cancel leaving us with

$$W_a(w_1) = \int_{w_1}^{\infty} A_f(w)dw \tag{10}$$

Equation (10) may be thought of as a layer-cake integration (i.e. a layer-by-layer integration of the areal extent of each layer) of the amount of snow in the basin after melt of depth w_1 . This form is useful, because $A_f(w)$ can be obtained easily from data. Numerical integration of $A_f(w)$ can be used to obtain $W_a(w)$. W_{amax} is $W_a(w=0)$. The depletion curve, $A_f^*(W_a/W_{amax})$, may be approximated by calculating $A_f(w)$ and $W_a(w)/W_{amax}$ for several values of w.

The pdf of snow water equivalence values sampled at peak snow accumulation define the pdf for all $w_1 \ge 0$. Using the 254 sampled values of snow water equivalence from Upper Sheep Creek on 3 March, 1993, $A_f(w_1)$ was calculated using equation (4) for w_1 between 0 and the maximum observed snow water equivalence in steps of 0.05 m. $W_a(w_1)$ was calculated for the same w_1 values using equation (10). A_f is plotted against W_a/W_{amax} in Figure 7. A three-part curve was used to numerically encode this function.

$$A_{f} = \begin{cases} 0.18\sqrt{W_{a}/W_{amax}} & \text{if } 0 \leq W_{a}/W_{amax} \leq 0.13 \\ 0.42\sqrt{W_{a}/W_{amax}} - 0.11 & \text{if } 0.13 \leq W_{a}/W_{amax} \leq 0.34 \\ (W_{a}/W_{amax})^{1.5} & \text{if } 0.34 \leq W_{a}/W_{amax} \leq 1.0 \end{cases}$$
(11)

 $A_f(W_a)$ was also found from the series of nine measurements and from the distributed model run for comparison to the curve estimated from the peak accumulation pdf.

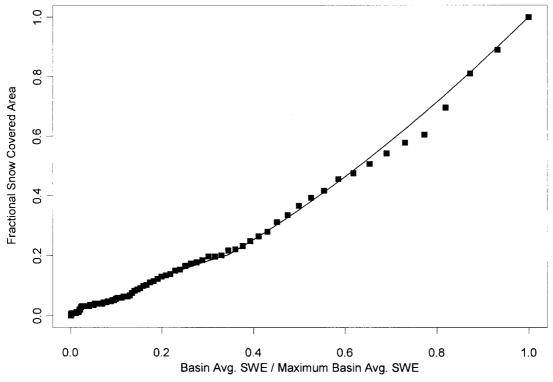


Figure 7. Depletion curve derived from pdf of 3 March 1993 snow water equivalence measurements and fitted curve

RESULTS AND DISCUSSION

A comparison between snow water equivalence predicted by the lumped and distributed models and measured in the snow survey is presented in Figure 8. The lumped model matched the distributed model very well, but both models overestimated the peak accumulation and showed a slightly early melt compared to observations.

The cell-by-cell calibration of the drift factor done using the 10 February, 3 March, and 23 March observations gives some insight into the source of the error for the two models. Figure 9 shows an example of the fit for two adjacent cells. One curve is a better fit to the data than the other and are typical of calibrations obtained at other cells. Both modelled curves have a similar shape, dictated by the model physics and driving climatic inputs. The differences between the model curves are based on one cell receiving greater modelled snow precipitation than the other, as determined by the value of the drift multiplier, the only parameter adjusted in the calibration. At some cells, the fit to the calibration period was good (e.g. L14) and at others, it was poor (e.g. K14). In almost all cells with a poor fit, the pattern was similar to that at K14 (i.e. overprediction of the peak accumulation). Both modelled curves predict early melt. The sum of many cells with this pattern of overprediction and underprediction is an identical pattern of overprediction and underprediction of the average (Figure 8).

In this study we used the Utah Energy Balance Model with the intent of having only the drift factor as an adjustable parameter. All other parameters were set based on literature or calibration to a few sites (Tarboton and Luce, 1997). From this basis, it could be said that the remaining differences between the observation and point model estimates indicate problems with the point model. It is possible that these errors could be rectified by making changes to the point model, but the emphasis of this paper is not incrementally

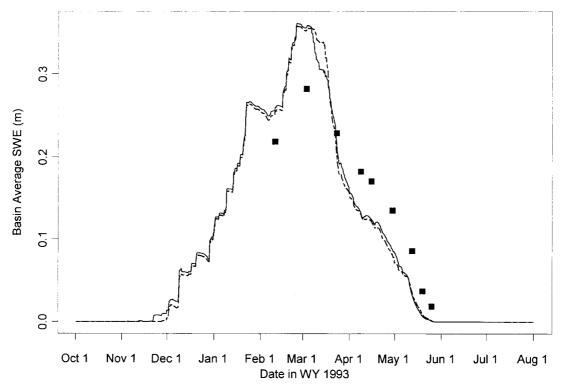


Figure 8. Modelled and observed basin averaged snow water equivalence for water year 1993. Solid line is the distributed model; dashed line is the lumped model. Points are observations

improving point models, rather it is the development of the distribution function approach which could be used with any point model (e.g. Anderson, 1976; Jordan, 1991).

With one adjustable parameter in the point model, there are theoretically 255 adjustable parameters for the basin, corresponding to each grid cell. However, our experience shows that the fundamental shape of the snow water equivalence graph over time is affected little by the drift factor. This means that for the aggregated distributed simulation there is really only one adjustable parameter, the average drift factor. Simulating over 255 cells provides a pdf of snow water equivalence over the basin, and distributed solar inputs. Luce *et al.* (1997, 1998) showed that the distributed solar information is of lesser value for this basin. Thus the fundamental information used by the distributed model is the same information used by the lumped model, a mean 'drift factor' and a pdf of relative snow accumulation. When seen in this light, the close agreement between the two models is not surprising.

Beven (1996) suggests that distributed models have too many degrees of freedom to be properly calibrated. Indeed, it may be possible that the 255 values of drift factor could have been manipulated together to provide a much better fit of the basin averaged snow water equivalence. However, when the distributed model is constrained to match the values at each cell, the aggregated distributed model results are the same as a lumped model using the probability density function information. This supports the idea that processes that can be modelled in a distributed fashion with independence from cell to cell may also be efficiently modelled using a lumped model that relates a probability density function of important site characteristics to important lumped state variables.

A comparison of the dimensionless depletion curves derived from the pdf of peak snowpack accumulation, the distributed model run, and the nine observations is shown in Figure 10. This figure shows that the depletion curve derived from the pdf of snow water equivalence at the date of maximum accumulation

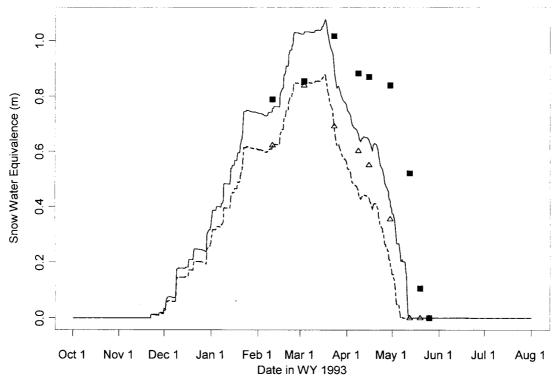


Figure 9. Plots of observed and modelled snow water equivalence at cells K14 (solid line and solid squares) and L14 (dashed line and open triangles). Model was calibrated by minimizing mean square error for the first three measurements of the year. The drift factor estimated for K14 is 2.55 and for L14 is 2.09. The solid line is representative of locations where poor calibrations were obtained and the dashed line is representative of locations where good calibrations were obtained

(3 March) is a good approximation to the observed and distributed model estimates of the actual depletion curve.

This finding improves the utility of the depletion curve concept because detailed observations of snow water equivalence over a basin at multiple times are unusual. Such observations would be necessary to either directly estimate the depletion curve or calibrate a distributed model. One may protest that gridded observations of snow water equivalence over a basin during peak accumulation are also rare. Tools have been developed (Elder *et al.*, 1989, 1991, 1995, 1998; Elder, 1995; Rosenthal and Dozier, 1996) to use remote sensing and modelling to estimate the distribution of peak snowpacks. These tools and data are comparatively inexpensive and provide a practical means to generate a depletion curve for the lumped model.

CONCLUSIONS

Through the use of an areal depletion curve it is possible to obtain lumped snowmelt model simulations that agree well with distributed model results and observed data. We have also presented a new method for the derivation of areal depletion curves from the distribution of peak snow water equivalence, and shown that the areal depletion curve obtained using this method compares well with the actual and modelled (using a detailed distributed model) areal depletion of snow. The finding suggests that the lumped model formulation applied here is a good substitute for the distributed model when detailed spatial patterns are not required. The distributed model required 255 simulations using the UEB model for each time step where the lumped model required only one, demonstrating considerable savings in computational effort. Effort in determining distributed parameters is likewise reduced.

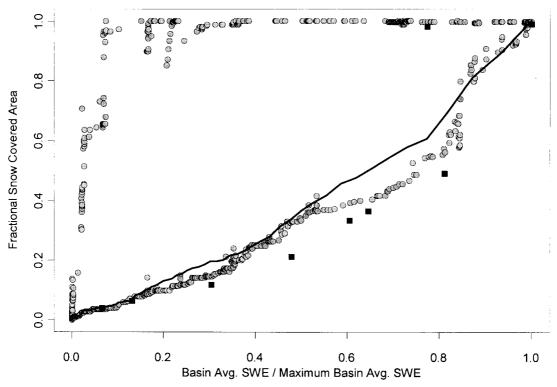


Figure 10. Comparison of depletion curves derived from 1) pdf of 3 March 1993 snow survey (line), 2) distributed model output (grey circles), and 3) snow surveys taken on nine dates (solid squares)

The reasoning behind the model should work with any point energy balance model. From the point by point calibration work, it was clear that the Utah Energy Balance model (Tarboton and Luce, 1997) did not always match the point scale data well. It is possible that both the distributed results and basin average results could be improved with a more detailed energy balance model.

Comparison of the depletion curve derived from the probability density function of peak snow accumulation to the observed depletion curve and that produced by the distributed model are encouraging. This finding combined with tools to quantify the distribution of snow over basins based on topography or remote sensing gives the lumped modelling approach presented here potential practical utility. This finding may also be useful for lumped empirical models that use the depletion curve concept.

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