STREAMFLOW QUALITY AND QUANTITY RELATIONSHIPS ON A FOREST CATCHMENT IN ALBERTA, CANADA

SURFACE AND SUBSURFACE HYDROLOGY

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Abstract

The composition of main constituents in the natural waters of remote catchments is usually difficult to determine. A data base, however, is required to assess impact of intended land use practices on upland areas which are the main source of prime quality water. Relationships between water quality and the readily available data on streamflow can provide the required baselines for comparative purposes.

Dilution effects in the stream waters of Marmot Experimental Watershed were examined to investigate water quality and quantity relationships. A number of mathematical models incorporating linear and nonlinear functional forms were hypothesized. The model parameters were determined from 31 samples collected and analyzed during 1971 and 1972. Calcium, Mg, Na, K, HCO₃, SO₄, Cl, and SiO₂ were the main constituents of these waters. Of the 28 postulated models, the regression models incorporating current flow, specific conductance, and a lagged variable of streamflow gave the best fit for most of the analyzed constituents. Other models based on the relationships of the above mentioned constituents to the components of streamflow (base flow and storm flow) are presently being developed and tested.

The coefficients for all models were derived from the collected data by the method of least squares. The models' prediction capabilities were tested by comparison with actual quality data acquired subsequently. The models are reassessed on this basis and ranked according to the magnitude of the sum of the squared deviations of observed values from those predicted by each model.

Introduction

The study of the composition and concentration of dissolved chemical constituents in natural waters is important in assessing the environmental impacts of land use practices. The laboratory analyses required for such determinations are costly An obvious variable is streamflow. Johnson et al. (1969), 'inder and Jones (1969), Hall (1970, 1971), Pionke, Nicks nd Schoof (1972), and Steele (1973, 1976) have shown how the 'elationship between stream discharge and water chemistry can 'e used as a predictive tool. The present study explores the 'redictive feasibility of using the relationships of streamlow and its components with the inorganic constituents in he streams originating from upland mountain forest watersheds.

Water is the chief agent releasing and transporting nutrients in an ecosystem (Hewlett and Nutter, 1969). Relationships of chemical constituents with streamflow are thus important for an understanding of the ecosystems in relation to forestry practices causing changes in nutrient exports.

As Hem (1970) states, the streams having the most consistent relationship between water discharge and dissolvedsolids concentrations ought to be the streams that receive a arge part of their mineral load from a relatively constant source upstream. The stream used in this study fulfills this riterion, and thus further provides a basis for testing this sypothesis.

<u>Study area</u>. The study was conducted in Marmot Creek experimental watershed, situated on the eastern slopes of the locky Mountains, about 80 km west of Calgary. It ranges in elevation from about 1500 m to 2750 m (mean elevation 2113 m) and represents spruce-fir [*Picea engelmanni* Parry, *P. glauca* (Moench) Voss, and *Abies lasiocarpa* (Hook.) Nutt.] vegetation eypical of the Saskatchewan River headwaters. The watershed was selected in 1962 as the first major research effort under the Alberta Watershed Research Program. In 1974 a treatment was applied on the Cabin subbasin to determine the effects of commercial logging on water yield, quality, and regime. The creatment consisted of clear-cutting six blocks totalling 40 ha.

The geology, soils, and vegetation of the area are described by Stevensen (1967), Beke (1969), and Kirby and Dgilvie (1969), respectively. Jeffrey (1965), Golding (1970), Singh and Kalra (1972), Singh (1976), and Telang et al. (1976), have described the hydrological and water quality aspects of the research program in the watershed.

Methods

<u>Collection of data</u>. The water quality samples collected from the permanent gaging site were analyzed at the Water Quality Laboratory of Environment Canada, Calgary. Stream discharge at the time of sampling was also determined from the hydrograph. Samples were collected at least monthly when the streamflow was nearly constant, as in the winter months, and more frequently in other months when it showed greater variation. Only samples collected from the Main Marmot Creek prior to commercial logging were used in this study.

<u>Modelling data</u>. The 31 samples collected during the 2-year period (1971-1972) were used for model building. These samples provided concentration data (mg/ℓ) for Ca, Mg, Na, K, HCO₃, SO₄, SiO₂, C1 and total dissolved solids which were treated as the variables to be predicted. The streamflow data (ℓ /sec) and specific conductance (μ S/cm) were the predictor variables.

Test data. The data collected during 1973 were used for testing or validating the models. The same input data were used for all models. As no data were available for SO_4 , the models for this constituent could not be tested for the present. Further tests and updating of the models are planned for future years when more data become available.

<u>Modelling procedures</u>. Stream discharge at the time of sampling (X_1) was used to find linear and nonlinear relationships with the concentration of each constituent. Another variable, specific conductance (X_2) , was included later to improve the goodness of fit. In order to incorporate the lagged effect, one more variable for the mean daily streamflow on the previous day (X_3) was also included as a predictor variable to further improve the fit.

All models were fitted to the data by the method of least squares. In addition to the estimation of model parameters, the correlation coefficient and the standard error of estimate were also computed.

For validation, the estimated values from each model were compared with the actual data. Each residual was listed and used for determining sum of residuals, sum of the squared deviations, mean square error, and standard error of estimate.

A computer program (Bathlahmy, 1972) was used on the 1973 data for determining the groundwater, interflow, and rapid flow components of stream discharge. A model based on three predictor variables (base flow, storm flow, and specific conductance) was derived for each modelled constituent. The model parameters, the correlation coefficient, and the standard error of estimate were also computed for this model by the method of least squares.

Results

The summary of modelling data is provided in Table 1. Table 2 gives a summary of the test data for comparative purposes.

Twenty-eight models were tried in all (Table 3). Of these, 12 had only one predictor variable, 12 had two predictor variables, and 4 had three predictor variables included.

Although outputs of model parameters and related statistics were obtained for all models, only results from I, XIII and XXV are listed in Tables 4, 5, and 6 and presented here. These tables summarize comparative statistics for the individual constituents Ca, Mg, Na, K, HCO_3 , Cl, and SiO₂ and also for the total dissolved solids.

Table 7 gives the related information on model parameters and other statistics when two of the three predictor variables of model XXV are replaced by the variables containing information on current base flow (or groundwater flow) and storm flow.

Discussion

The dependence of the concentration of chemical constituents of stream waters on discharge is evident from the generally highly significant relationships listed in Table 4. Total dissolved solids and Ca^{2+} show the most significant relationship, whereas Cl^- relationship is the least, and statistically nonsignificant. The negative sign of the correlation coefficient shows dilution effects, i.e., the concentration of constituents become lower if streamflow increases.

The slope (b) representing change in concentration with unit change in streamflow is the highest for total dissolved solids and the lowest for K. From Table 4, the individual constituents can be ranked in order of the absolute value of b as $HCO_5^- > Ca^{2+} > Mg^{2+} > SiO_2 > Na^+ > K^+ > C1^-$.

The relationships improve considerably when specific conductance is included as an additional variable in the prediction model. The use of specific conductance for estimating the concentration of total dissolved solutes has been tested earlier (Singh and Kalra, 1975). In this paper its use has been extended to other dissolved constituents. The results (Table 5) show improvements in all the tested constituents.

Table 1. Summary of Modelling Data, Marmot Creek

	-		0		
Variable	Mean	Standard Deviation	1 Minimum	Maximum	Coefficient of Variation
Ca ²⁺ (mg.	(2) 44.3	2 1.0.4	25.2	56.6	23.61
Mg ²⁺	. 11.4	3.1	6.8	17.2	25.87
Nat	1.	1 0.6	0.4	2.5	51.90
κ+	. 0.:	5 0.1	0.3	0.7	21.09
HCO3	173.	5 40.7	101.7	232.8	23.44
CL ⁻		2 0.1	0.1	0.3	41.45
SiO ₂	. 4.	5 0.3	2.9	5.5	18.48
Total dissolved solids	164.	4 42.1	83.0	226.5	25.58
Current streamfl (1/	ow 279. sec)	7 388.2	11.6	1441.3	138.79
Specific conduct	ance 302. /cm)	9 66.8	182.0	406.0	22.06
Streamflow previ day (2/	ous sec) 284.	0 390.7	11.6	1452.6	137.54

Table 2. Summary of Test Data, Marmot Creek

Variable	Mean	Standard Deviation	Minimum	Maximum	Coefficient of Variation
Ca ²⁺ (mg/l)	41.8	10.1	28.2	62.6	24.09
Mg ²⁺ "	10.5	2.6	3.4	16.7	24.72
. _{Na} + "	1.1	0.4	0.6	2.0	38.01
к + "	0.5	0.1	0.3	0.8	19.25
HCO3 "	161.9	34.0	115.0	230.0	21.01
c1 ⁻ "	0.4	0.2	0.1	1.0	58.01
SiO ₂ "	4.7	0.6	3.6	6.0	12.18
Total dissolved solids "	152.1	33.9	107.0	218.0	22.29
Current streamflow (1/sec)	369.9	289.3	15.3	996.7	78.20
Specific conductance (µS/cm)	282.6	62.6	204.0	452.0	22.15
Streamflow previous day (L/sec)	354.4	261.6	15.6	985.4	73.82

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Table 4. Model Parameters and Related Statistics for Estimating Concentration (mg/ℓ) of Dissolved Constituents (Y) Using Current Streamflow $(X_1, \ell/sec)$ as Predictor Variable (Model: Y = a + bX₁)

	Model Par	ameters		• S	Sum of		
Constituent (mg/l)	а	Ъ.	τ*	Model	Predicted	prediction residuals	
Ca ²⁺	50.6317	-0.0229	-0.85	5.56	6.86	-10.23	
Mg ²⁺	13.4855	-0.0057	-0.72	2.18	2.01	-27.70	
Na ⁺	1.3937	-0.0016	-0.69	0.42	0.42	8.03	
к+	0.4913	-0.0001	-0.23	0.10	0.13	2.41	
нсо,	197.565	-0.0858	~0.82	23.74	22.08	-138.92	
c1	0.1864	-0.00003	-0.09	0.08	0.35	7.09	
SiO ₂	5.0612	-0.0027	-0.77	0.54	0.90	19.77	
otal dissolved solids	190.179	-0.0920	-0.85	22.58	20.41	-122.03	

* r is correlation coefficient and $S_{\rm E}$ is standard error of estimate.

Table 5. Model Parameters and Related Statistics for Estimating Concentration (mg/ℓ) of Dissolved Constituents (Y) Using Current Streamflow $(X_1, \ell/sec)$ and Specific Conductance $(X_2, \mu S/cm)$ as Predictor Variables (Model: Y = a + bX₁ + cX₂)

	Model Parameters]	s _E *	Sum of
Constituent (mg/l)	a	ь	с R*		Model data	Predicted data	prediction residuals
Ca ²⁺	3.7706	-0.0035	0.1368	0.984	1.94	3.46	20.09
Mg ²⁺	-3.4576	0.0013	b.0494	0.94	1.10	2.20	-15,59
Na ⁺	-1.5724	0.0002	0.0084	0.85	0.31	0.28	6.20
к+	0.5393	-0.0001	-0.0001	0.24	0.10	0.13	2.39
HCOJ	-5.2898	-0.0016	0.5920	0.985	7.33	8.23	5.72
C1 ⁻	0.1035	0.00002	0.0002	0.14	0.08	0.37	6.91
S102	1.9047	-0.0008	0.0089	0.85	0.46	0.84	17.26
Total dissolved solids	4.8911	-0.0151	0.5407	0.977	9.27	7.72	2.14

* R is multiple correlation coefficient and $S_{\rm E}^{-}$ is standard error of estimate.

Table 3. Mathematical Models for Estimating Concentration (mg/ℓ) of Dissolved Constituents (Y); X₁ is Current Streamflow (ℓ /sec), X₂ is Specific Conductance (μ S/cm), and X₃ is Mean Streamflow (ℓ /sec) on Previous Day

$Y = a + bX_1$ $Y = a + bX_2$ $Y = a + b (ln X_1)$ $Y = a + b (ln X_2)$ $Y = a + b (ln X_3)$ $Y = a + bX_1$ $Y = a + bX_2$ $Y = a + bX_3$ $Y = a + b (ln X_1)$ $Y = a + b (ln X_2)$ $Y = a + b (ln X_3)$
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$Y = a + bX_1 + cX_3$
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$Y = a + b (ln X_1) + c (ln X_2)$
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$X = a + b (ln X_2) + c (ln X_3)$
$Y = a + bX_1 + cX_2$
$Y = a + bX_1 + cX_3$
$Y = a + bX_2 + cX_3$
$Y = a + b (\ln X_1) + c (\ln X_2)$
$Y = a + b (ln X_1) + c (ln X_3)$
$Y = a + b (ln X_2) + c (ln X_3)$
$Y = a + bX_1 + cX_2 + dX_3$
$Y = a + b (ln X_1) + c (ln X_2) + d (ln X_3)$
$Y = a + bX_1 + cX_2 + dX_3$
$Y = a + b'(\ln X_1) + c (\ln X_2) + d (\ln X_3)$

Table 6. Model Parameters and Related Statistics for Estimating Concentration (mg/l) of Dissolved Constituents (Y) Using Current Streamflow (X₁,l/sec), Specific Conductance (X₂, μ S/cm), and Mean Streamflow on the Previous Day (X₃, l/sec) as Predictor Variables (Model: Y = a + bX₁ + cX₂ + dX₃)

Constituent (mg/l)		Model pa	· ·	s _e *		
	а	b	R*	Ē		
Ca ²⁺	-6.0787	-0.00003	0.00014	0.1739	0.95	2.93
Mg ²⁺	3.5253	0.00001	-0.00009	0.0271	0.78	1.54
Na ⁺	-0.9534	-0.000002	-0.000004	0.0071	0.93	0.13
к+	0.1487	-0.0000007	-0.000002	0.0014	0.53	0.10
HCO3	4.7719	-0.00007	-0.00002	0.5680	0.986	4.67
C1	0.2865	-0.000006	0.000002	0.0013	0.68	0.19
S102	3.8966	-0.000007	-0.000003	0.0039	0.62	0.46
tal dissolved solids	-0.6559	-0.00015	0.00007	0.5635	0.990	4.19

-R is multiple correlation coefficient and S_F is standard error of estimate.

Table 7. Model Parameters and Related Statistics for Estimating Dissolved Constituents Using Components of Hydrograph. X_1 is Base Flow (ℓ /sec), X_2 is Storm Flow (ℓ /sec), and X_3 is Specific Conductance (μ S/cm) (Model: Y = a + bX₁ + cX₂ + dX₃)

	Model parameters					s _e *		Sum of
Constituent (mg/ ʃ)	a	b	с	d	R*	Model data	Pre- dicted data	prediction residuals
Ca ²⁴	3.4044	-0.0090	0.1378	0.0056	0.984	1.97	3.71	23.62
Mg ²⁺	-3.3459	0.0030	0.0491	-0.0017	0.94	1.12	2.29	-16.67
Na ⁺	-1.7233	-0.0009	0.0088	0.0012	0.85	0.31	0.32	6.68
к+	0.5626	0.00004	-0.0002	-0.0002	0.25	0.10	0.14	2.31
HCO3	-5.0626	0.0018	0.5914	-0.0035	0.985	7.46	8.41	4.56
C1 ⁻	0.1904	0.0005	0.00004	-0.0005	0.26	0.08	0.37	6.77
\$10 ₂	1.9188	-0.0007	0.0089	-0.0001	0.85	0.47	0.85	17.22
otal dissolv- ed solids	4.5247	-0.0206	0.5417	0.0056	0.977	9.44	8.01	5.67

-R is multiple correlation coefficient and $S_{\rm E}$ is standard error of estimate.

The inclusion of mean streamflow on the previous day as yet another variable produced a slight improvement in the fit, as indicated by the R and S_E values in Table 6. These improvements occurred primarily in K^+ and $C1^-$, as shown by the R values and the sum of prediction residuals for these constituents.

The models incorporating logarithmic transformation showed slightly higher R values in some cases, but were not preferred over those presented here. The slightly higher R values may be due to the slightly lesser deviations in the transformed data, rather than actual data (Brownlee, 1953). It was therefore decided to use models incorporating untransformed data only.

The use of the components of streamflow in deriving model equations for the 1973 data shows the best overall fit for all constituents. The R values for the modelled constituents range from 0.53 to 0.990 (Table 7). The R values for the model XXV in Table 6 had a range of 0.25 to 0.985.

The models based on the components of streamflow and specific conductance can therefore provide the most accurate estimates. Further work along these lines is in progress and consists of testing nonlinear predictive models for constituents which have multiple correlation coefficients lower than 0.90.

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