QUANTIFYING THE ERRORS INVOLVED IN TRANSFERRING SUSPENDED SEDIMENT RATING EQUATIONS ACROSS ABLATION SEASONS

by C.R. Fenn


ABSTRACT

Bias-corrected suspended sediment rating equations are developed for the data from six separate ablation seasons in the Glacier de Tsidjore Nouve basin. These equations differ significantly from one another. Suspended loads as predicted by equations used beyond their year of origin fall in the range of 34–278% of the true load. Mean absolute percentage errors (MAPE) based on forecasts computed from each year’s model applied to the other 5 years of data range from 33 to 81%, with a mean of 52%. The equivalent MAPE for a lumped multi-year model is 38%, for an autocorrelation-correcting generalized least-squares (GLS) regression model, is 15%, and for a transfer function (TF) developed from a single 23rd order, period, only 5%. Simple structure-transfer functions appear to be more robust to temporal transfer than the usual type of rating equation, and offer the possibility of greatly improving upon the poor performance of OLS rating equations in providing estimates of suspended loads for periods beyond frames of origin.

INTRODUCTION

Rating equations are unquestionably the most widely used means of estimating unknown suspended sediment concentrations (C) and loads (L) in glacier-fed and normal rivers alike. Their popularity lies in the apparent simplicity of their derivation and application. C is held to be a function of stream discharge (Q), and the relationship between measured values of C and Q is formulated as a regression model. The resulting equation is combined with measured values of Q to estimate unmeasured concentrations (Ĉ) directly and unmeasured loads (L̂) as a time-referenced ĈQ̂ product. A model which requires the use of only Q data to provide estimates of C and L is clearly highly suitable for practical purposes. On the other hand, such a model excludes reference to variations in sediment delivery (notably in relation to drainage shift, glacier motion, rainfall, and mass-movement events), and is thus an oversimplified representation of the process system. Whether the simple rating model remains, nevertheless, a valuable practical tool hinges upon the quality of the estimates it delivers.

The quality of rating equation estimates depends in part upon the extent to which the deterministic effect modelled by the fitted function exerts control upon variability in C (relative to stochastic effects and omitted deterministic effects), in part upon how well the data used to derive the rating equation fit the assumptions of the model applied, and in part upon how well conditions in the forecast period (the horizon) match those in the original derivation period (the origin). This paper extends previous evaluations of the quality of estimates from suspended sediment rating equations and transfer functions for the basin of Glacier de Tsidjore Nouve, Switzerland (Gurnell and Fenn, 1984; Fenn and others, 1985), by examining the stability of rating relationships derived from the same basin for six different ablation seasons. The general question addressed here is that of temporal specificity: how good are estimates of C and L when simple rating equations are used beyond their frame of origin?

THE DATA BASE

Table I details the sampling base of the six data sets examined herein. Suspended sediment concentrations were measured in units of mg l−1 (1 mg l−1 = 1 g m−3). Streamflow discharges are expressed in units of m3 s−1 (1 m3 s−1 = 1 × 10−9 m3 s−1). Scatter plots of C against Q for each year are shown in Figure 1. The data differ substantially from year to year. Full details of the site, and of the field and laboratory methods employed have been given in Fenn and others (1985).

DERIVATION OF RATING EQUATIONS: TECHNICAL BACKGROUND

Rating equations are typically derived via the ordinary least-squares (OLS) regression procedure. The assumptions of the technique, and the consequences of violating those assumptions, are by now well known (e.g. Ferguson, 1977). Whilst those relating to the response between the dependent and the independent variables can be satisfied by careful experimental design, by partitioning and/or lagging cases as required and by adopting the correct functional form of model, those relating to the properties of the residuals are more difficult to satisfy.

Untransformed C–Q rating models are prone to yield residuals which are not normally distributed, are not homoscedastic and, in the case of glacier basins, are not serially uncorrelated. The log–log transform is frequently adopted as a joint solution to the first two of these problems. Since estimates of C, as opposed to log C are required, the log–log model is re-expressed via the usual back–transformation procedure as a multiplicative relationship between C and Q. The back-transformation introduces an
underestimation bias into predictions of C, however, since it yields the geometric as opposed to the arithmetic mean of the conditional distribution of C at any given value of Q (Ferguson, 1986). A correction factor \( (B_f) \) proportional to the standard error of estimate \( (s) \) of the log-log model reduces this bias (Miller, 1984; Jansson, 1985), \( (B_f = \exp(\frac{s^2}{2}) \) for natural logs, \( \exp(2.65s^2) \) for common logs). The adjustment amounts to an upward, parallel shift of the rating curve on the log-log plot, and a higher, steeper curve on the arithmetic plot (Ferguson, 1986). The problem of true autocorrelation in the residual series (as distinct from quasi-autocorrelation resulting from the mis-specification of the model, the omission of important variables, and the presence of lags and/or changes in response) can be treated by the use of a generalized least-squares (GLS) model weighted by the serial correlation coefficient of the residual series. By turning the autocorrelation structure to advantage, a transfer function (TF) can be developed as an alternative to a regression equation, compared to an OLS regression equation. It is thus appropriate to evaluate the simple rating model in terms of its ability to deliver satisfactory estimates, as well as in terms of its statistical validity.

**OLS RATING MODELS FOR THE 6 YEAR GLACIER DE TSIDJIORE NOUVE DATA**

Inspection of the C-Q scatter plots (Fig. 1) indicates the need to transform each and every data set in order to stabilize the variance. The Box–Cox transformation search procedure (Box and Cox, 1964) indicated the suitability of the log-log transform in all cases. The transformed scatter plots (Fig. 2) show the improvement which is achieved. The results from OLS regression applied to each data set (and to the combined multi-year data set) are given in Table II. Bias-correction factors calculated for each data set are also shown, together with the corrected coefficient in the multiplicative model. The Durbin Watson d statistic indicates that all models yield positively autocorrelated residuals, that is \( d < 1.5, P < 0.01 \). Detailed investigation of the 1978 data set, involving lagging the series to the best match position, splitting the data into sections of constant response, testing the effects of air temperature and rainfall, and deriving discriminant ratings for rising and falling stage and ablation periods has established that the autocorrelation is of the true rather than the quasi type, and is likely to be an inherent property of this type of data (Fenn and others, 1985). We should therefore beware of the attendant effects on the coefficients and statistics of the models. The coefficients \( a \) and \( b \) will be statistically unbiased, but their standard errors are likely to be underestimated, as is the variance of the residual series and the \( R^2 \) statistic correspondingly inflated.

**TESTING THE TRANSFERABILITY OF RATING EQUATIONS**

Bearing in mind the reservations noted above, we can assess the degree to which a given rating model is likely to be temporally transferable by examining the stability of the coefficients of individual rating equations. An examination...
of the 95% confidence bounds of the regression coefficients that is, \( a = \pm 1.96SE_a \) and \( b = \pm 1.96SE_b \) indicates a minimal degree of overlap across the 6 year data set (Table II). The Chow test of equality of coefficients (Chow, 1960) on the 6 year set results in rejection of an hypothesis of no difference in coefficients between years \( (F^* = 195.4, P < 0.001) \). Allowing for the fact that standard errors \( SE_{a,b} \) involved in the first test, and the error sums of squares (ESS) involved in the Chow test may both be artificially low, it would still appear reasonable to accept that the coefficients of the rating model of 1 year cannot be taken to be applicable in any other year with any degree of confidence. Inspection of the positions and slopes of the curves for each of the 6 years on the rating plot (Fig. 3) provides an unambiguous confirmation of this conclusion.

A similar picture emerges with respect to the stability of coefficients of rating models developed for partial periods of a single ablation season. The Chow test was applied in a comparison of coefficients of the whole 1978 data set against the coefficients of models for the June and July data separately \( (F^* = 5197.5) \), models for the pre-first flush and post-first flush periods identified in Fenn and others (1985) \( (F^* = 3604.4) \), and models for each of the nine sub-periods of alternately rising and falling ablation which constituted the 1978 season \( (F^* = 611.8) \). All \( F^* \) values lie well above the critical value at \( P < 0.001 \).

It would appear reasonable to conclude from the above analyses that the sampling variability of the C-Q relationship is so great, on both year-to-year and within-year time-scales, as to preclude the transferability of a sample-rating equation to any other period save that which mirrors the conditions of the origin to a very high degree. This confirms the conclusions of Østrem and others (1967), Østrem (1975), and Fenn and others (1985). The strict statistical implication is that the data from different years are derived from different populations. A glacio-hydrological translation of this, assuming that each glacier basin does have a population distribution, is that this distribution must be very large, and very heterogeneous, and that observations from a single-year sample only parts of the total population. This being so, there would appear to be some benefit in combining data from different samples as a means of approximating a population relationship for the basin.

**A MULTI-YEAR RATING EQUATION: TOWARDS A GENERAL MODEL?**

Combination of the 6 years of data for the Glacier de Tsidjöre Nouve basin results in a multi-year data base which includes the very different conditions associated with early season snow-melt periods, summer snowfall recession...
estimating suspended sediment load in different years is now assessed in comparison with equivalent estimates obtained from each of the single season OLS models. For further comparisons, estimates are also obtained from a GLS model and from a transfer function (TF). The GLS model was derived from the 1978 data set via the Cochrane–Orcutt method (Cochrane and Orcutt, 1949) as a means of handling the autocorrelation present in the residual series. The transfer function was developed from an arbitrary, as opposed to an optimally selected, 23-d period within the 1981 set via the Box–Jenkins method (Box and Jenkins, 1970). The model was deliberately specified in only one part so it could be used for making practical predictions. The constant was fixed at zero, and an equation containing only one coefficient was selected. The noise term used in the estimation phase was dropped when making predictions, on the grounds that zero is the expected value of the residuals. Gurnell and Fenn (1984) have described the procedures involved in developing the function.

Each equation was used to produce a series of estimated suspended sediment concentrations for the full duration of each of the six field seasons examined. These estimated concentrations (\( \hat{C} \)) were then combined with the discharge data to produce an estimated load (\( \hat{L} \)) for each year, whilst the measured concentrations (\( C \)) were similarly used to produce a true load (\( L \)). The prediction equations used to calculate \( \hat{C} \) are given in Table III. The OLS equations require only a discharge record to yield estimates of \( C \), but the GLS and TF equations also require the provision of a starting value for \( C \) (\( C_0 \)), since both involve a \( C_{t-1} \) term. The starting values used in this analysis were not specially selected, nor were different selections attempted in order to achieve an optimum result; the starting value \( C_0 \) was, in all cases, simply taken to be that value of \( C \) measured immediately before the start of the prediction period. Thereafter, estimated values of \( C \) (\( \hat{C} \)) were used as values of \( C_{t-1} \) to calculate forward. Only a single value of \( C_0 \) thus needs to be measured in order to run the GLS and TF equations. Intermittent measured values of \( C \) can, of course, be fed in as values of \( C_{t-1} \) to reestimate the estimation as required.

Each predicted load is expressed as a percentage of the corresponding measured load in Table III. The figures confirm, on a year-to-year time-scale, that an equation developed from a period of low sediment transport underestimates loads when it is applied to periods of higher sediment transport and vice versa, according to the disparity between conditions in the origin and the horizon (e.g. OLS85 badly underestimates load in all other years; OLS77 badly overestimates load in all other years). The figures in Table III show just how serious the under- or overestimation may be. The MAPE value is a relative measure of the performance of each model as a predictive tool, enabling us to quantify the errors involved in applying a given rating equation beyond its derivation period. It is clear from the MAPE figures given in Table III that transferring OLS rating equations from one year to another leads to absolute errors of estimation in the order of 35–81%, averaging at 52%. The multi-year data model gives errors averaging 38%. The GLS78 model performs substantially better, yielding an average absolute percentage error of 15%. The transfer function, however, performs on a higher level altogether, giving a MAPE value of only 5%. The year-to-year transferability of what is in essence a very simple transfer function is remarkable. It raises the possibility that the structural form of the function has a sound physical basis. The stability of the structure and coefficients of transfer functions developed from different years thus require examination.

CONCLUSIONS

Ordinary rating equations change from year to year and from period to period within a year. Their application in times other than those in which they were developed accordingly leads to substantial error in estimation of suspended sediment concentrations and loads. A rating equation developed from a multi-year data set improved upon the average performance of single-year models, but only to a limited extent. In contrast, GLS and TF models produced significantly better results. This is because these models calculate changes in suspended sediment concentration resulting from change in discharge, and
because they address the autocorrelation structure in the C-Q relationship. Whilst the GLS and TF models are more difficult to develop than an OLS model, they are no more difficult to apply, in that they require only an initial measured value of C in order to run. The initial values used in these tests were not optimized but, since the value used for \( C_0 \) is pivotal, the effects on final estimates of using different values of \( C_0 \) require evaluation. The simple transfer function used here showed a consistent ability to estimate suspended sediment loads in different years with an absolute error margin of only 5%. We can conclude that the simple transfer function has been shown to be substantially more robust to temporal transference than OLS-rating curves. OLS-rating equations provide poor estimates of the suspended sediment load exported from glacierized basins when used beyond their frames of origin. Transfer functions appear to offer much greater promise in estimating suspended sediment transport in glacier melt-water streams.

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Table III. Predicted Suspended Sediment Loads as Percentages of Measured Loads

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<tbody>
<tr>
<td>Length (d):</td>
<td>93</td>
<td>60</td>
<td>96</td>
<td>95</td>
<td>14</td>
<td>21</td>
<td></td>
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<tr>
<td>Measured load (tonnes): 6780</td>
<td>1296</td>
<td>8250</td>
<td>7718</td>
<td>1793</td>
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Prediction equation

OLS77 \( \hat{C}_i^t = 3.4404 Q_i^t \) 0.9232 \( Q_i^t \) 1.2851 (96) 183 108 197 141 278 81.2

OLS78 \( \hat{C}_i^t = 0.1776 Q_i^t \) 1.2323 55 (102) 67 133 94 180 39.4

OLS81 \( \hat{C}_i^t = 0.3662 Q_i^t \) 0.2584 75 140 (90) 175 124 239 60.6

OLS82 \( \hat{C}_i^t = 112.51 Q_i^t \) 0.184 60 125 60 (93) 67 140 35.4

OLS84 \( \hat{C}_i^t = 0.0035 Q_i^t \) 1.8506 48 87 67 148 (101) 185 46.2

OLS85 \( \hat{C}_i^t = 1.1987 Q_i^t \) 0.9232 34 64 38 69 49 (97) 49.2

MULTI \( \hat{C}_i^t = 0.5177 Q_i^t \) 1.1470 64 120 76 145 103 200 38.0

GLS78 equation: \( \hat{C}_i^t = 0.343 \hat{C}_i^{t-1} \) 0.632 \( Q_i^t \) 1.285

TF81 equation: \( \hat{C}_i^t = 1.03 \hat{C}_i^{t+1} \) 0.847 40 67 49 97 164 39.0

*MAPE = Mean Absolute Percentage Error = \( \frac{1}{n} \sum (|L_i - \hat{L}_i|/L_i) \) 100

Figures in brackets indicate a prediction made for the year in which the prediction equation was derived. All MAPE figures are based on years in which the prediction equation is applied beyond its origin. The prediction equations are given in their bias-corrected back-transformed form. Note that the constant appearing in the TF81 equation is simply the bias-correction factor.