REACH-SCALE SEDIMENT TRANSFERS: AN EVALUATION OF TWO MORPHOLOGICAL BUDGETING APPROACHES

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ABSTRACT
This paper compares two approaches used to derive measures of annual sediment transfers within a 1 km long piedmont reach of the gravel-bed River Coquet in Northumberland, northern England. The techniques utilize: (i) channel planform and cross-section surveys based on a theodolite/electronic distance measurement (EDM) survey of 21 monumented channel cross-sections and channel and gravel bar margins; and (ii) theodolite-EDM survey generating a series of $x$, $y$, $z$ coordinates, from which digital elevation models (DEMs) of the reach were constructed. Calculating the difference between DEM surfaces provided a measure of volumetric change between surveys carried out during the spring of 1999 and 2000. The use of kriging in DEM generation and differencing permits computation of estimate variances and confidence intervals for sediment transfer. Error analysis, validating the DEMs using surveyed cross-sections, indicated a mean error between surveyed and DEM-generated cross-sections of around twice the value of the $D_{50}$ of the surface sediment in the reach. Comparison of sediment volumes derived from the two approaches suggests that, compared with the DEM method, monumented cross-sections underestimate the magnitude of volumetric changes that occur within the reach. The cross-section approach relies on a simplistic integration of the volumes, whereas DEM differencing provides an estimate at a resolution under the control of the analyst. Furthermore, the cross-section approach does not permit a reliable estimate of the uncertainty of the volumes calculated. In addition, the DEM methodology based on the morphological unit scale provides an explicit identification of spatial patterns of erosion and deposition, a feature that cross-section-based approaches may fail to include.

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KEY WORDS: gravel-bed river; morphological budget; DEM; kriging; sediment transfer

INTRODUCTION
Quantification of changes in channel morphology, which take place in response to sediment erosion, transfer and deposition (Ashmore and Church, 1998), provides a means by which bedload volumes can be monitored (Davies, 1987). Use of this time–space integrated measure avoids some of the complications caused by spatial and temporal variability of bedload transport (Hubbell, 1987) and provides a means of estimating sediment transport, which also yields information on storage and displacement of bed material, important in assessing channel stability (Ashmore and Church, 1998). Investigation of three-dimensional morphological change in gravel-bed rivers requires repeat survey of bed elevation and planform adjustments. The majority of research in this field has focused on the bar scale (e.g. Neill, 1987; Ferguson and Ashworth, 1992; Ferguson et al., 1992; Lane et al., 1994).

Brewer and Passmore (2002) present budgeting techniques designed to evaluate sediment transfers at intermediate reach scales (1–3 km) which are typical of major instability zones in UK piedmont rivers. This approach also discriminates between morphological units (channel and bar) within the reach, allowing more accurate determination of sediment fluxes and storage in these instability zones. Wathen et al. (1997) note that reach-scale sediment storage is rarely quantified in sediment budget studies, but recognize that it has a significant effect on the precision of morphological methods of bedload estimation at the reach scale.

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To date, application of the ‘morphologic approach’ (sensu Ham and Church, 2000) to reach-scale sediment budgeting in its various forms (e.g. Goff and Ashmore, 1994; Martin and Church, 1995; Paige and Hickin, 2000; Ham and Church, 2000; Brewer and Passmore, 2002; Fuller et al., 2002), has largely relied upon channel cross-sections to provide the morphological information required to calculate volumetric changes. Lane et al. (1994) suggest that this emphasis on the channel cross-section is a legacy of hydraulic geometry, and may represent a weakness in the approach. Extrapolating changes measured along a cross-section to areas of channel and bars which may be tens of metres from a section may be problematic, as this assumes the cross-section is an adequate representation of the units it bisects (Fuller et al., 2002). Furthermore, the cross-section may not adequately reflect downstream change in channel morphology (Lane et al., 1994), and may fail to accommodate the potential effects of downstream sedimentological structures on channel processes (Naden and Brayshaw, 1987; Wittenberg, 2002). This has led Brasington et al. (2000) to note that, ‘the implicitly cross-stream emphasis gives rise to a high degree of uncertainty in reach-scale sediment budgets derived through the interpolation of cross-section data’. Some morphological approaches (e.g. Lane et al., 1994; Eaton and Lapointe, 2001) have moved towards a ‘distributed terrain-sensitive survey’ (Ashmore and Church, 1998) to generate reach-scale digital elevation models (DEMs), while others, such as Westaway et al. (2000) and Lane (2001), have examined cross-sections extracted directly from DEMs. However, few comparisons between DEMs and independently surveyed ground survey data (here cross-sections) have yet been made in the context of morphological budgeting. Using this approach, this paper assesses the reliability of a cross-section-dependent morphological approach to quantifying sediment transfers in the wandering, gravel-bedded River Coquet, Northumberland, UK.

One of the reasons for the traditional reliance on cross-stream approaches to reach-scale sediment budgeting lay in the practical difficulties of topographic data acquisition, storage and analysis at resolutions which adequately reflect channel morphology (Brasington et al., 2000). Cross-section-derived budgets, which can be acquired and processed rapidly (Brewer and Passmore, 2002), have thus formed the traditional ‘backbone’ of this research to date. Recent developments have led to the production of high resolution DEMs of fluvial environments (e.g. Lane et al., 1994; Milne and Sear, 1997; Heritage et al., 1998; Brasington et al., 2000; Eaton and Lapointe, 2001), where acquisition of high-resolution topographic information is central to effective construction of such DEMs (Westaway et al., 2000). However, it is important to note that volumes derived from morphological budgeting represent lower-bound estimates of sediment transfer. Firstly, sediment may move through or within a reach without any surface morphological expression. Secondly, significant negative bias in volumes of scour-fill can be produced by localized scour-fill compensation (Lindsay and Ashmore, 2002).

STUDY SITE

The River Coquet rises in the Cheviot Hills (776 m) in northern England (Figure 1). The focus of this research has been an instability zone located at Holystone (National Grid Reference NY 958027), a piedmont setting at the upland fringe in the catchment, where the Coquet drains an area of c. 255 km² (Figure 1). The Coquet at Holystone has been characterized by a high degree of lateral instability and channel avulsion over the last 150 years (Fuller et al., 2002). The contemporary active channel at Holystone comprises features common to both braided and meandering rivers, being locally divided by expanses of bare gravel, but having well defined pool–riffle units (Fuller et al., 2002); as such it fits Ferguson and Werritty’s (1983) classification as a wandering river. According to Nanson and Croke’s (1992) classification, the 400–500 m wide valley floor at Holystone is a medium-energy non-cohesive, wandering gravel-bed river floodplain. The valley as a whole is similar to other gravel-bed rivers in northern England, in that it displays a characteristic ‘hourglass’ valley morphology, with alternating confined and unconfined sections. The D₅₀ of the surface sediment in the reach is 51 mm (Fuller et al., 2002).

CROSS-SECTION-DERIVED BUDGETING

A Sokkia Set 5F Total Station (electronic distance measurement (EDM) precision ±(5 ± 5 ppm × D)mm, angular resolution 1 s) was used to survey channel planform (channel boundaries, barforms and major chute channels) and monumented cross-sections in March 1999 and 2000. The spacing of the cross-sections along the study
reach was designed to ensure that channel bed elevation was measured at regular intervals through the reach and that each bar in the reach was crossed by at least one cross-section. This provided the means of assessing the vertical changes, produced by sediment gain or loss, which occurred across the morphological units in the reach. The locations of the cross-sections dividing the site into subreaches are shown overlaid on the 1999 channel planform (Figure 1). Cross-section data for each survey were incorporated into a geographical information system (GIS) using ArcInfo™ to facilitate the validation described later in this paper.

Sediment budgets

Sediment budgets were calculated using the morphological budget described in detail by Brewer and Passmore (2002) and Fuller et al. (2002) (Figure 2). One important difference between this and previous application of this methodology is in the treatment of bank erosion. Both Brewer and Passmore (2002) and Fuller et al. (2002) treat bank erosion as a sediment gain, contributing material from the floodplain. However, as DEM sediment volumes calculated here are based on the subtraction of interpolated surfaces (with any bank erosion creating a void between surfaces), bank erosion will be treated as a sediment loss in this application of the morphological budget described by Brewer and Passmore (2002) and Fuller et al. (2002). This approach integrates planform and cross-section data at the resolution of the discrete morphological unit within each subreach. Within ArcInfo™, vertical changes in area along each cross-section were calculated and standardized to net gain/loss values per square metre. These values were then multiplied by the corresponding planform area values to give a net
Figure 2. Derivation of sediment budgets using channel cross-sections and outline planform (after Brewer and Passmore, 2002, reproduced by permission of the Geological Society Publishing House).

Stage 1: Calculation of volumetric change for each morphological unit

For morphological units traversed by:

\[ \delta V_d = \frac{\delta A_{dw}}{L_{dw}} \times A_d \]

\[ \delta V_a = \left[ \frac{\delta A_{aw}}{L_{aw}} + \frac{\delta A_{av}}{L_{av}} \right] \times \frac{A_a}{2} \]

Where

- \( \delta V_d \) = volumetric change for morphological unit ‘d’ (m³)
- \( \delta A_{dw} \) = change in cross-sectional area of morphological unit ‘d’ (m²)
- \( L_{dw} \) = length of morphological unit ‘d’ across profile ‘W’ (m)
- \( A_d \) = planform area of morphological unit ‘d’ at end of the budget period (m²)
- \( \delta V_a \) = volumetric change for morphological unit ‘a’ (m³)
- \( \delta A_{aw} \) = change in cross-sectional area of morphological unit ‘a’ across profile ‘W’ (m²)
- \( \delta A_{av} \) = change in cross-sectional area of morphological unit ‘a’ across profile ‘V’ (m²)
- \( L_{aw} \) = length of morphological unit ‘a’ across profile ‘W’ (m)
- \( L_{av} \) = length of morphological unit ‘a’ across profile ‘V’ (m)
- \( A_a \) = planform area of morphological unit ‘a’ at end of the budget period (m²)

Stage 2: Calculation of volumetric change for each sub-reach

E.g. for sub-reach ‘VW’: \( \delta V_{VW} = \Sigma (\delta V_a + \delta V_b + \ldots + \delta V_g) \)

Figure 2. Derivation of sediment budgets using channel cross-sections and outline planform (after Brewer and Passmore, 2002, reproduced by permission of the Geological Society Publishing House).

Gain/loss value (cubic metres) for each morphological unit within the subreach. Where morphological units are bounded by two cross-sections (e.g. large bar and main channel units), the planform area was halved and multiplied by the upstream and downstream cross-section data respectively. The results of this budgeting are shown alongside those derived from the DEM budget approach described below.
DEM GENERATION AND VOLUME CALCULATION

Coincident with the planform mapping exercise described above, additional points were surveyed from a common local datum throughout the active channel to produce a series of elevation measurements along the reach. Sampling of the topographic surface using these points was conditioned by breaks in slope within and between morphological units throughout the reach. In 1999, 2661 irregularly located points were surveyed in the active channel area (50 509 m²), producing an overall point density in the reach of 0·05 points m⁻² (Figure 3). In 2000, 2985 points were surveyed within the same area, producing a point density of 0·06 points m⁻² (Figure 3). Reach-scale averages include areas of higher sampling density (in excess of 1 point m⁻²) in areas of rapid topographic change. These areas include features such as bar edges, bend apices and riffles. Both surveys took approximately eight man-days for the cross-sections (described above), and a further eight for the additional points for DEM construction.

Creating DEMs using kriging

A DEM is a (usually) rectangular array of regularly spaced estimates of the elevation of land surface above a given datum. The locations at which the elevations are supplied are the mesh points of the array. In a computer representation the DEM is represented as a lattice or grid of (usually) square pixels or rasters with the elevation attribute being that at the centre of each grid cell. There is a range of different methods for interpolating a DEM.
from irregularly spaced data, but where error estimates are required kriging is an option. Wackernagel (1995, p. 89) points to three advantages of spatial interpolation using kriging: (i) the process begins with an analysis of spatial structure of the data which is then used in the estimation procedure; (ii) ordinary kriging is an exact interpolator; and (iii) kriging provides an estimate of the estimation error. Kriged estimates of the interpolated surface between sampling points were made at those locations which represent the centre of each grid cell; corresponding variances were also computed at the same locations. These variance estimates can be used to create a confidence interval around the elevation estimates (which are themselves means). They can also be used to provide an estimate of the propagated uncertainty associated with GIS operations (such as addition or differencing) and this will be discussed below.

Each survey point consists of a triple \( \{x_i, y_i, z_i\} \) in which \( x_\alpha = \{x_\alpha, y_\alpha\} \) corresponds to a measurement of location and \( Z(x_\alpha) = z_\alpha \) corresponds to some measure of the elevation at that location. Elevation from these measurements, at any location in the study area, can be estimated using Equation 1:

\[
Z^*(x_0) = \sum_{\alpha=1}^{n} w_\alpha Z(x_\alpha)
\]

where \( x_0 \) is the location at which the estimate \( Z^* \) is required. The kriged estimate is effectively a locally weighted mean. This form of kriging is known as Ordinary Kriging.

The weights \( w_\alpha \) are required to sum to unity to ensure that the results are unbiased. The weights are computed from the semivariogram function \( \gamma(h) \) which returns the value of the variance between locations at distance \( h \) apart. The semivariogram is zero when \( h \) is zero and rises to a maximum (the sill) at a distance known as the range. The assumption is that a second-order stationary process is being modelled, otherwise the value of \( \gamma(h) \) would vary in different parts of the study reach. The variance of the elevation prediction, \( Z^*(x_0) \) at the location \( x_0 \) is calculated from Equation 2:

\[
\sigma^2(x_0) = \sum_{\alpha=1}^{n} w_\alpha \gamma(x_0 - x_\alpha) + m
\]

where \( m \) is a Lagrange multiplier to ensure unbiasedness and \( x_0 - x_\alpha \) is the distance between the location at which the variance is desired and that of the \( \alpha \)th sample point.

The elevation estimates and their associated variances may be generated from the weighted mean of all the data (the unique neighbourhood) or a local sample (moving neighbourhoods). The variances are likely to be higher with moving neighbourhoods because the sample sizes are smaller. The size of this sample may be in terms of a fixed number of nearest neighbours or a fixed radius. In the former case the method can be considered as being adaptive to the local spatial configuration of the data, while with a fixed radius the approach is justified if the sampling density of the measurements is not too variable across the study area.

An experimental semivariogram can be estimated using Equation 3 (Bailey and Gatrell, 1995, p. 164):

\[
\hat{\gamma}(h) = \frac{1}{2n(h)} \sum_{\delta x, \delta y = h} (z_i - z_j)^2
\]

in which the squared differences of all point pairs which are a distance \( h \) apart are summed and divided by twice the number \( n(h) \) of such points. The 2 in the denominator ensures that this is a semivariance rather than a variance estimate.

The value of \( h \) is increased (from zero to about half the maximum extent of the study area) and the resulting values of the semivariogram can be plotted against \( h \). The semivariogram for the 2000 data is shown in Figure 4 and reveals that there is spatial trend in the dataset – a curve which continues to rise and never reaches a sill is indicative of non-stationary data.

If the variogram is not to produce different values for \( h \) in different parts of the study area, then the process being modelled must be isotropic. If the covariance depends on direction then an anisotropic process will have a different variogram in say the north–south direction from the east–west direction. Bailey and Gatrell (1995,
p. 180) indicate two varieties of anisotropy: geometric, in which the range changes with direction; and zonal, in which the sill is dependent on direction. In the case of the relatively straight (and narrow) river reach under study, anisotropy need not be considered problematic.

**Drift**

Whilst kriging assumes a second-order stationary process, measurements of gravel-bed river morphology are rarely stationary in this sense, as the data will exhibit a slight downward trend between upstream and downstream measurements within a reach. The existence of drift means that the value of the semivariance depends on where it has been estimated in the reach. This invalidates the assumption of stationarity and alternative approaches to dealing with this should be considered. The drift may either be modelled with a linear model and Ordinary Kriging applied to the residuals from the trend, or Universal Kriging may be used. If the former option is chosen, the linear trend has then to be added back into the estimated DEM to produce the final surface (Bailey and Gatrell, 1995). A suitable trend model is expressed in Equation 4:

\[ z_i^* = \alpha + \beta x_i + \gamma y_i + \epsilon_i \]  

where \( x_i \) and \( y_i \) are the coordinates of the \( i \)th sample point whose elevation is \( z_i \), \( \epsilon_i \) is a random error term, and \( \alpha \), \( \beta \), and \( \gamma \) are coefficients to be estimated.

However, modelling of a spatial process in this fashion can be problematic and should *sensu strictu* be undertaken using a generalized least-squares estimate of the parameters \( \alpha \), \( \beta \) and \( \gamma \), although ordinary regression may suffice. In this case the trend has been modelled with ordinary least-square regression to give an indication of the nature of the trend. The fitted values are shown in Table I. The residuals for each sample point form the spatially detrended dataset.

Universal Kriging assumes that the \( z \) values are a combination of a deterministic component (drift), and a random residual function which is itself assumed to exhibit second-order stationarity. The drift is often modelled as either a linear or quadratic function of location. The spatial covariance structure may then be modelled with a suitable semivariogram. Universal Kriging splits the random function into two components: the first is a
deterministic component which represents the drift and the second is a random component which represents the residual function. Kriging generally gives more precise predictions than those based on trend surface models because the Universal Kriging weights are optimal (Cressie, 1990: 164). The ArcInfo™ Universal Kriging option used here uses a linear trend model with a linear semivariogram model.

As a further check on whether the use of Universal Kriging was appropriate, the semivariogram for the detrended data was estimated to determine whether it informally suggested stationarity in the variance estimate: this semivariogram is shown in Figure 5. This semivariogram corresponds in shape to the ‘classic’ semivariogram for a second-order stationary process which can be modelled (e.g. Wackernagel, 1995, p. 43). This semivariogram exhibits a small nugget effect of about 0·06 m, and a sill at about 0·17 m at a range of about 50 m. The nugget effect corresponds to measurement errors at small ranges, and a semivariance of 0·06 m corresponds to a standard deviation of 0·035 m. This compares with the estimate of measurement error of 0·026 m by Brasington et al. (2000).

Grid size

An important consideration in DEM generation and differencing is the grid size of the computed DEMs – DEMs with finer grids take longer to compute than those with coarser grids, but volume estimates under finer grids may be more accurate than those made under coarser grids, as coarse grid intervals may lead to the loss of potentially important breaks in slope within the channel environment (Brasington et al., 2000). However, Brasington and Richards (1998) suggest that the use of very fine grids may incorporate ‘spurious artefacts’

<table>
<thead>
<tr>
<th>Survey date</th>
<th>Intercept (α)</th>
<th>Easting (β)</th>
<th>Northing (γ)</th>
<th>r²</th>
</tr>
</thead>
<tbody>
<tr>
<td>1999</td>
<td>129·03</td>
<td>−0·004256</td>
<td>−0·004574</td>
<td>0·9084</td>
</tr>
<tr>
<td>2000</td>
<td>128·91</td>
<td>−0·004226</td>
<td>−0·004624</td>
<td>0·9057</td>
</tr>
</tbody>
</table>

Figure 5. Experimental semivariogram for the 2000 surveyed data after linear trend removal
within the DEM, which are not representative of the real elevation surface. Florinsky (2002) suggests that fine grain DEMs may include interpolation errors due to the Gibbs phenomenon, and these errors are a source of unnecessary noise in subsequent surface analysis. To assess the most appropriate grid cell size for this dataset, Universal Kriging was used to create four DEMs with a grid size of 5 m, 1 m, 0.5 m, and 0.25 m. The root mean square error results of these DEMs (Table II) suggest that there is a dependency on grid cell size, and that there is some difference between a 5 m grid and the other grids but little to choose between 1 m, 0.5 m, and 0.25 m grids. This indicates that no extra benefit in integrating under the surface is gained by using a grid smaller than 1 m. Given the costs of creating the grids, the 1 m grid represents an appropriate trade-off between fitness for purpose and cost of production. Consideration of the computer times required is illuminating in terms of the cost of creating the DEMs (Table III). The cost of creating a 5 m or 1 m DEM is small, given the area of the reach being modelled; decreasing the grid cell size by a factor of 16 increased the run time by a factor of about 17.

**DEM validation**

In order to independently assess the quality of DEMs, which are controlled by the precision, reliability and accuracy of individual points, and their spatial coverage (Lane, 1998), a comparison of DEM-extracted information with independently acquired ground measurements is required (Westaway *et al.*, 2000). In the first instance, independently acquired information here is provided by a series of channel cross-section surveys, surveyed separately (but within a week) from the topographic survey used to provide the source data for the DEM creation. No geomorphologically significant flows occurred during this time period. The accurate location of the channel cross-sections within the high-resolution survey is known, as the pegs used to monument each cross-section were surveyed using the same equipment and local datum. The process of validation was to compare the elevation estimate from the DEM with the surveyed elevations along each cross-section. The resulting mean deviation is shown in Table IV together with an estimate of the confidence interval (CI).

The results in Table IV indicate that the mean difference between the surveyed cross-sections and those extracted from the DEM lies around the $D_{sd}$ of the surface sediment, measured at 83 mm, which itself has a 12 per cent error (Fuller *et al.*, 2002). However, a close correspondence between the surveyed and DEM-derived cross-sections is not necessarily to be expected. Both surveys (cross-sections and $x,y,z$ DEM points) will be subject to errors associated with surface roughness (e.g. placement of the detail pole used in survey will vary between the top and bottom of clasts). Thapa and Bossler (1992) classify errors as gross, systematic and random, and whilst field procedures for this study were designed to minimize the occurrence of gross errors or blunders,
Table IV. Differences between surveyed cross-sectional elevations and DEM estimates

<table>
<thead>
<tr>
<th>Survey data</th>
<th>Mean difference (m)</th>
<th>Standard error</th>
<th>95% CI (m)</th>
</tr>
</thead>
<tbody>
<tr>
<td>2000</td>
<td>0.4054</td>
<td>0.0065</td>
<td>0.0278, 0.0533</td>
</tr>
<tr>
<td>1999</td>
<td>-0.1071</td>
<td>0.0105</td>
<td>-0.1277, -0.0865</td>
</tr>
</tbody>
</table>

The data used to create the DEMs are subject to measurement (random) and interpolation (systematic) error. The data used to create the cross-sections are subject to measurement error as well. Furthermore, it may be unreasonable to assume that the errors are normally distributed and independent (López, 2002). Thus an analytic derivation of the effects of the compounded error is difficult – it may be that the only means of assessing this is through Monte Carlo simulation. However, the generation of 99 random DEMs to create 99 test statistics to compare with the means in Table IV is computationally intensive, provides a single whole-map statistic which yields little useful information for the budget estimates, and is beyond the scope of this paper.

**Error estimation and propagation**

What is the accuracy of a digital elevation model? Can the uncertainty associated with an operation on two digital elevation models be quantified? Error estimates are sometimes encountered in the form of a statement to the effect that the root mean square error is \( p \) or that \( n \) per cent of the estimates are within \( m \) per cent of their true value. These are whole map statistics (Openshaw, 1996) and may not be particularly helpful – the errors in interpolation are likely to be lower nearer the sample points and higher away from them.

Assuming that the measurement errors are independent and normally distributed with a mean of zero and that the variance can be estimated, it might be supposed that the errors from a GIS operation such as DEM differencing would be normally distributed with a mean of zero and standard deviation given by Equation 5:

\[
\sqrt{\sigma_1^2 + \sigma_2^2}
\]  

(5)

This has been used to derive a ‘minimum level of detection’ between two surfaces (Brasington et al., 2000). The problem with this is that it assumes that both surfaces are independent – this is an unlikely assumption in the case of DEMs of the same river bed. To estimate the propagated error the covariance between the surfaces should be considered. If the correlation between the two DEMs, \( r_{12} \), is available, an estimate of the propagated error is given by Equation 6 (Burrough and McDonnell, 1998, p. 247):

\[
\sqrt{\sigma_1^2 + \sigma_2^2 + 2\sigma_1\sigma_2 r_{12}}
\]  

(6)

In this study the cross-correlation between the 1999 and 2000 surface was estimated as 0.98. As grid cell variance estimates for both surfaces are available, the propagated variance for the difference DEM can be estimated to provide 95 per cent confidence surfaces.

**Volume estimation**

The volume of scour and fill, and the net change, can be estimated by simple DEM differencing. Kriging provides an estimate of the variance in each cell, and these variance estimates may be propagated for the differenced DEM. A tentative estimate of the error associated with the estimation of scour, fill, and net change may be made using Equation 6. A lower bound on the estimate can be made by summing the individual variances contributing to the areas of scour or fill, and the overall total variance for the grid can be used to estimate the variance for the net change. It is likely that this will be an underestimate of the true variance since this ignores the covariance between the cells in the difference DEM. Computation of the covariance is lengthy – the grid has 768 rows and 571 columns, and the covariance between every pair of cells requires separate computation. ArcInfo™ does not provide access to the covariance information. The results are shown in Table V where the estimates together with their tentative 95 per cent confidence intervals are given. For fill the
Table V. Net volumes between 1999 and 2000 DEMs compared with gains and losses derived from a cross-section-based morphological budget for the same period; 95 per cent confidence interval computed from kriged variance estimates

<table>
<thead>
<tr>
<th></th>
<th>DEM volume (m³)</th>
<th>Confidence interval</th>
<th>Cross-section volume (m³)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Gain</td>
<td>1539</td>
<td>1388–1689</td>
<td>1227</td>
</tr>
<tr>
<td>Loss</td>
<td>9423</td>
<td>9129–9716</td>
<td>2177</td>
</tr>
<tr>
<td>Net change</td>
<td>−7884</td>
<td>−8213 to −7554</td>
<td>−950</td>
</tr>
</tbody>
</table>

Figure 6. DEMs of the Holystone reach in 1999 and 2000

size of the confidence interval is about 19·6 per cent of the size of the estimated sediment transfer. For scour, the equivalent proportion is 6·3 per cent, and for the net gain it is 8·4 per cent. Further technical research is required to set these error estimates in context.

COMPARISON OF MORPHOLOGICAL APPROACHES
Subtraction of the 1999 DEM surface from the 2000 surface (Figure 6) produced not only volumetric estimates of gain, loss and net sediment transfer (Table V), but also a DEM of difference between the surfaces, allowing clear identification of the most significant changes in channel morphology (i.e. net erosion and deposition) within
Figure 7. DEM difference surface. The interval of 0.1 m (effectively blanking elevation changes smaller than this) reflects a figure twice the $D_{50}$ in the reach. Any differences smaller than 0.1 m could reflect errors induced by surface roughness alone. Areas of net sediment gain are shown in lighter shading and enclosed within a boundary. 2000 unit morphology is superimposed.

the reach for the period 1999–2000 (Figure 7). To provide a more detailed comparison between the cross-section and DEM budgeting approaches requires division of the reach DEM of difference into subreaches exactly equivalent to the cross-section-defined subreaches, followed by a DEM subtraction (1999–2000) for each discrete subreach. Volumetric changes were therefore calculated for identical subreaches to those defined by the cross-sections (cf. Figures 1 and 7). Sediment transfers calculated using the cross-section-dependent approach are shown alongside those from the DEM subtraction approach for each subreach at Holystone (Figure 8). This shows apparent (and consistent) underestimation of the magnitude of sediment losses using cross-sections. Given the compatibility of the two methods of topographic representation (surveyed cross-section and DEM: Table IV),
this is likely to be due to significant morphological changes taking place between the cross-sections. Losses are particularly concentrated within the channel (Figure 7). Sediment gains appear to be highly localized, and may be identified both within the channel and at the margin of bars (Figure 7). While the pattern of change demonstrated by the difference DEM suggests a degradational regime with significant channel scour, the cross-section morphological budget suggests a state much closer to equilibrium (Table V and Figure 8). However, DEM differencing employing low precision data models may be unable to represent shallow surface changes reliably (Brasington et al., 2000). If this is the case, there are important implications as, in gravel/cobble-bed rivers, channel scour typically occurs in deep, but comparatively localized areas (e.g. bank erosion, chute dissection), while deposition may occur in a spatially extensive, but shallower pattern. An exception to this is the deposition of gravel/cobble lobes, which are a common feature of this particular reach. However, where deposition is in the form of shallow sheets, it may go unnoticed, especially while deep channel or bar erosion in excess of the minimum level of detection of the DEM may dominate the pattern of apparent channel change. Nevertheless, the magnitude of the discrepancy between these DEM and cross-section-based budgets (Figure 8, Table V) is likely to be due to localized channel scour and deposition occurring between channel cross-sections within this specific reach.

CONCLUSIONS

This paper (a) attempts to quantify the consistency of the DEMs of the reach, (b) assesses the validity of a cross-section budgeting method and (c) provides a tentative estimate of the uncertainty associated with budgeting using DEM differencing. The DEM methodology presented here, which uses a sampling strategy based on the morphological units, provides a rigorous characterization of spatial patterns of erosion and deposition, which cross-section-based approaches at the site have failed to identify. In particular, the DEM approach to morphologic budgeting has permitted identification of long-stream three-dimensional channel (and barform) development for this reach. The use of a DEM-based approach to monitoring the three-dimensional nature of river channel form has a major advantage over the cross-section approach in that data collection is topographically defined, resulting in better surface representation (Lane, 1998). Furthermore, change in location of scour/
deposition from the originally identified ‘sensitive’ areas, which directed cross-section location, may account for some errors, as whilst cross-sections were carefully selected at the outset of the investigation in 1997, these positions remain fixed, irrespective of later channel development and shifts in areas of activity. The consequence of this dynamism was such that by 2000, some cross-sections were no longer effectively identifying geomorphological activity in the Coquet channel. This does not suggest that cross-section-derived budgets as a whole have a tendency to underestimate losses, but rather reflects local processes operating within the Holystone reach during the survey period, and may well be a function of cross-section spacing at this site.

Lane et al. (1994) suggest that much more detailed information on river bed topography is needed to accommodate the increasing recognition of spatially distributed form–process feedback in fluvial environments and to provide a more rigorous linkage between channel topography and sediment transport processes. The use of DEMs in a 1 km reach of the River Coquet at Holystone provides a much improved representation of channel morphology, and in the context of morphological budgeting, a more reliable means of linking changes in channel morphology with sediment transport processes. Furthermore, this paper demonstrates that DEMs of the channel topography may be a reliable representation of morphological units (within the limitations imposed by surface sediment characteristics), and can be used not only to estimate volumes of sediment transfer up to the 1 km reach scale, but also to quantify the uncertainty associated with these estimates. Uncertainty can be viewed as one component of DEM reliability and this paper provides some initial steps towards dealing with this in the context of morphological budgeting. The results here point to the need for a more explicit consideration of error propagation in the production of morphological sediment budgets for gravel-bed rivers.

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