

Process feedbacks that control transport capacity at formative flows in laterally-constrained gravel-bed rivers: a laboratory study

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Abstract

In gravel-bed rivers, deterministic approaches to predicting bedload transport use the mean bed shear stress (termed one-dimensional or ‘1D’ equations) or integrate across the frequency distribution of shear stress (2D equations). At low flows, incorporating a range of shear stress values increases prediction accuracy, but at relatively high flows the 1D and 2D approaches are similarly accurate. We contribute to an understanding of the stage-dependent relationship between morphology and bedload transport, and specifically why the mean shear stress characterises transport capacity at formative discharges. We performed physical modelling using a generic Froude-scaled model of a steep laterally-constrained gravel-bed river and captured digital elevation models to perform 2D hydraulic modelling. Both 1D and 2D Meyer-Peter Müller equations were highly accurate across two distinct channel morphologies. In alternate bar channels, transport capacity was controlled by negative feedbacks between flow depth and local bed slope that resulted in a relatively homogeneous distribution of bed shear stress. In plane-bed channels, which lacked the degrees-of-freedom available for large-scale morphologic adjustment, transport capacity was controlled by a spatially variable migrating surface texture. The contrasting spatial patterns of morphology, hydraulics, and surface texture between the two channel morphologies highlight the potential for the same correlation between mean shear stress and transport capacity to emerge through different mechanisms. We suggest that nonlinear feedbacks explain why simple bedload transport equations can be highly effective above a certain flow stage across a range of channel morphologies, and further work should examine whether lateral adjustment confounds this result.

Process feedbacks that control transport capacity at formative flows in laterally-constrained gravel-bed rivers: a laboratory study

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4 **Key words**

- 5 • Both 1D and 2D Meyer-Peter Müller equations were highly accurate across two distinct
6 experimental channel morphologies at formative discharges
- 7 • The effectiveness of 1D equations at high flows was explained by nonlinear feedbacks
8 between morphology, hydraulics, and sediment transport
- 9 • Specifically, in alternate bar channels transport capacity was controlled by negative feed-
10 backs between flow depth and local bed slope

Abstract

In gravel-bed rivers, deterministic approaches to predicting bedload transport use the mean bed shear stress (termed one-dimensional or '1D' equations) or integrate across the frequency distribution of shear stress (2D equations). At low flows, incorporating a range of shear stress values increases prediction accuracy, but at relatively high flows the 1D and 2D approaches are similarly accurate. We contribute to an understanding of the stage-dependent relationship between morphology and bedload transport, and specifically why the mean shear stress characterises transport capacity at formative discharges. We performed physical modelling using a generic Froude-scaled model of a steep laterally-constrained gravel-bed river and captured digital elevation models to perform 2D hydraulic modelling. Both 1D and 2D Meyer-Peter Müller equations were highly accurate across two distinct channel morphologies. In alternate bar channels, transport capacity was controlled by negative feedbacks between flow depth and local bed slope that resulted in a relatively homogeneous distribution of bed shear stress. In plane-bed channels, which lacked the degrees-of-freedom available for large-scale morphologic adjustment, transport capacity was controlled by a spatially variable migrating surface texture. The contrasting spatial patterns of morphology, hydraulics, and surface texture between the two channel morphologies highlight the potential for the same correlation between mean shear stress and transport capacity to emerge through different mechanisms. We suggest that nonlinear feedbacks explain why simple bedload transport equations can be highly effective above a certain flow stage across a range of channel morphologies, and further work should examine whether lateral adjustment confounds this result.

Key words: bedload transport, transport capacity, morphodynamics, channel morphology, wavelet transform, gravel-bed rivers

1 Introduction

In alluvial systems, there are strong feedbacks between the morphology of a river and the transport of bedload material (Bridge and Jarvis, 1982; Dietrich and Smith, 1983; Church, 2010). Bedload transport and deposition are key processes by which morphology emerges (Church and Ferguson, 2015), but by concentrating flow in preferential paths, the morphology may determine both the spatial distribution and rate of bedload transport (Ferguson, 2003). For a given flow rate, the rate of bedload transport averaged over time may be termed the system's 'transport capacity.'

Equations that aim to predict bedload transport have been developed using both physical models and field data. Most of these are one-dimensional or '1D' as they index the forces that drive bedload entrainment using the reach-averaged shear stress acting on the bed (Shields, 1936; Gomez and Church, 1989). Researchers have sought to improve upon the 1D approach by accounting for the frequency distribution of shear stress (a 2D approach). By incorporating the range of shear stress values, 2D approaches predict bedload transport more accurately at low flows (Paola and Seal, 1995; Paola, 1996; Nicholas, 2000; Ferguson, 2003; Bertoldi et al., 2009; Monsalve et al., 2020). However, under relatively high flow conditions, 1D and 2D approaches are similarly accurate, yielding similar estimates of transport capacity (e.g., Monsalve et al., 2020).

The contrasting performance of these approaches across different discharge stages has been explained by the role of channel morphology (e.g., bars and pools) that controls the spatial distribution of transport. Under low flow conditions, where the threshold of motion may occur at shear stresses around or above the mean, the 1D approach considerably underestimates transport capacity or incorrectly predicts zero transport because active bedload is concentrated in narrow pathways (Davoren and Mosley, 1986; Carson and Griffiths, 1987).

In this investigation, we aim to identify the feedbacks between bedload transport, channel morphology, and hydraulics that explain why transport capacity is well characterised by the mean shear stress at formative discharges. We performed physical modelling using a generic Froude-scaled model of a steep laterally-constrained gravel-bed river and captured digital elevation models to perform 2D hydraulic modelling. The results highlight process interactions that modulate the spatial and temporal distribution of bedload, which have implications for predicting sediment transport and channel behaviour under flood conditions.

Table 1: Summary of unit discharges (Q/W) used in each phase (P) of experimental Runs a-c.

	unit discharge q [L/s/m]			
	P1	P2	P3	P4
Run a	5.00			
Run b	3.33			
Run c	2.22	3.33	5.00	7.50

66 2 Methodology

67 Experiments were performed in the Adjustable-Boundary Experimental System (A-BES) at the
68 University of British Columbia (Figure 1), a portion of which have been reported by Adams and
69 Zampiron (2020). The A-BES comprises a 1.5 m wide by 12.2 m long tilting stream table, where
70 the experiments were run as generic Froude-scaled models based on 2003 field measurements
71 from Fishtrap Creek in British Columbia, Canada. The channel had a gradient of 0.02 m/m,
72 average bankfull width of 10 m, formative discharge of approximately 7,500 L/s, and bulk D_{50}
73 of 55 mm. With a length scale ratio of 1:25, the A-BES is scaled to within around 30 percent
74 of the prototype, with an initial width of 0.30 m, formative discharge Q of approximately 1.5 L/s,
75 and D_{50} of 1.6 mm ($D_{84} = 3.2$ mm, $D_{90} = 3.9$ mm). The sediment mixture comprised natural
76 clasts with a density of around 2,500 kg/m³.



Figure 1: Adjustable-Boundary Experimental System (A-BES) at the University of British Columbia, featuring cameras (top-right) and bank control system at a width of 30 cm.

77 The experiments utilised interlocking landscaping bricks to constrict the channel to various
78 widths W between approximately 0.08–0.60 m. The narrowest setting was selected based on

Table 2: Summary of experiments conducted in the A-BES. Length refers to the median length of digital elevation models (DEMs), which generally varies by ± 0.1 m, and does not include approximately 20–30 cm of bed at the upstream end. DEM count excludes screeded bed. Experiments 1 and 2 are published in Adams and Zampiron (2020).

Exp	W [m]	L [m]	Q [L/s]	Duration [hrs]	DEMs
Exp2a	0.08	8.7	0.40	16	24
Exp2b	0.08	8.6	0.27	16	24
Exp1a	0.30	10.8	1.50	16	24
Exp1b	0.30	10.7	1.00	16	24
Exp1c	0.30	10.8	0.66, 1.00, 1.50, 2.25	8, 4, 4, 4	20, 16, 16, 16
Exp3a	0.45	10.8	2.25	16	24
Exp3b	0.45	10.8	1.50	16	24
Exp3c	0.45	10.7	1.00, 1.50, 2.25, 3.37	8, 4, 4, 4	20, 16, 16, 16
Exp4a	0.60	10.8	2.00	16	24
Exp4b	0.60	10.8	3.00	16	24
Exp4c	0.60	10.7	1.33, 2.00, 3.00, 4.50	8, 4, 4, 4	20, 16, 16, 16

79 preliminary experiments where the channel was narrowed until bar formation was suppressed
80 entirely. In addition to the various channel widths, four different unit discharges ($q = Q/W$)
81 were used across the experiments (i.e., discharge was scaled by width) that increased by a
82 factor of 1.5 (Table 1). We conceptualise all four unit discharges as formative as they are capa-
83 ble of forming alternate-bar morphologies in the wider channels. Two constant-discharge runs
84 used the middle two discharges, and one multi-discharge run consisted of the four discharges
85 in increasing order. A full list of experiments is provided in Table 2.

86 At the beginning of each experiment the bulk mixture was mixed by hand to minimise lateral
87 and downstream sorting, and then the in-channel area was screeded to the height of weirs at
88 the upstream and downstream end using a tool that rolled along the brick surface. The flow
89 was run at a relatively low rate (at which there was little-to-no movement of sediment) until the
90 bed was fully saturated, and was then rapidly increased to the target flow. There was no initial
91 feed of sediment.

92 Three different types of data were collected throughout each experiment; surface photos,
93 stream gauge measurements, and sediment output. A rolling camera rig positioned atop the
94 A-BES consisted of five Canon EOS Rebel T6i DSLRs with EF-S 18–55 mm lenses positioned
95 at varying oblique angles in the cross-stream direction to maximise coverage of the bed, and
96 five LED lights. Photos were taken in RAW format at 0.2 m downstream intervals, providing a
97 stereographic overlap of over two-thirds. Ten water stage gauges comprised of a measuring
98 tape (with 2 mm intervals) on flat boards were located along the inner edge of the bricks every
99 1 m (but every 0.8 m for the 0.08 m experiments due to the slightly shorter length). To minimise
100 edge effects, gauges were not placed within 0.60 m of either the inlet or the outlet. Also, the

101 gauges were read at an almost horizontal angle, which in conjunction with the dyed blue water,
102 minimised systematic bias towards higher readings due to surface tension effects. Based on
103 the measurement precision of the stream gauge readings, errors of 6—11 percent could be
104 expected for mean hydraulic depths (relative errors are variable due to different depths).

105 The data collection procedure was designed to maximise measurement accuracy as much as
106 was reasonably possible. Given that stream gauge data would later be paired with topographic
107 data, the timing of gauge readings needed to closely coincide with surface photography (i.e.,
108 so there was as little morphologic change as possible between the time gauges were read and
109 the time the bed was captured). Every time photos were taken the bed was drained, as the
110 surface water would distort the photos. These constraints necessitated a procedure in which
111 manual stream gauge readings (to the nearest 1 mm) were taken 30–40 seconds before the
112 bed being rapidly drained (around the minimum time it would take to obtain the readings), after
113 which the bed was photographed and gradually re-saturated before resuming the experiment
114 (approximately 10 minutes).

115 Each experimental phase was divided into a series of segments between which the data col-
116 lection procedure would occur. The procedure occurred in 5, 10, 15, 30, 60, and 120 minute
117 segments with four repeats of each (i.e., 4 x 5 min, 4 x 10 min), which was designed to reflect
118 the relatively rapid rate of morphologic change at the beginning of each phase. For exam-
119 ple, in wider channels, alternate bars developed within an hour, and there was relatively little
120 morphologic change in the following hours (Adams and Zampiron, 2020).

121 Throughout the experiments, sediment falling over the downstream weir was collected in a
122 mesh bucket, drained of excess water, weighed damp to the nearest 0.2 kg, placed on the con-
123 veyor belt at the upstream end, and gradually recirculated at the same rate it was output (gas
124 opposed to a 'slug' injection). Based on a range of samples collected across the experiments,
125 we determined the weight proportion of water to be approximately 5.8 percent and applied this
126 correction factor to obtain approximate dry weights. Zero sediment was fed into the system
127 during the first 5-minute phase. The experiments are best described as pseudo-recirculating
128 as sediment was fed at the end of each segment, and every 15 minutes regardless of whether
129 the bed was drained.

130 **2.1 Data processing**

131 Using the images, point clouds were produced using structure-from-motion photogrammetry in
132 Agisoft MetaShape Professional 1.6.2 at the highest resolution, yielding an average point spac-

133 ing of around 0.25 mm. Twelve spatially-referenced control points (and additional unreferenced
134 ones) were distributed throughout the A-BES, which placed photogrammetric reconstructions
135 within a local coordinates system and aided in the photo-alignment process. Using inverse dis-
136 tance weighting, the point clouds were converted to digital elevation models (DEMs) at 1 mm
137 horizontal resolution. Despite the use of control points, the DEMs contained a slight arch effect
138 (an artefact of the processing) whereby the middle of the model (in a downstream direction)
139 was bowed upwards. This effect was first quantified by applying a quadratic function along the
140 length of the bricks, which represent an approximately linear reference elevation (brick eleva-
141 tions vary by ± 4 mm). The arch was then removed by determining correction values along the
142 length of the DEM using the residuals, which were then applied across the width of the model.

143 For each DEM, ten wetted cross-sections were reconstructed using the water surface elevation
144 data, which were then used to estimate reach-averaged hydraulics. For more detailed spatial
145 analysis, the flow conditions (water depth, shear stress) were reconstructed using a 2D numer-
146 ical flow model (Nays2DH) to the final DEM of each discharge phase. To minimise rounding er-
147 rors associated with the relatively shallow depths in the stream table and the grid size, the DEM
148 size and discharge were adjusted to the prototype scale (i.e., using a length scale of 25) for the
149 flow modelling. The estimated water depths, shear stresses and velocities from Nays2DH were
150 then back-transformed to the model scale (Table 3). We removed cells with relatively shallow
151 flow to eliminate non-active channel areas, defined as depths less than $2D_{84}$. The mean-
152 normalised frequency distributions of shear stress were fit with a Gamma distribution, where
153 the goodness-of-fit was statistically significant ($P < 0.05$) based on both Kolmogorov-Smirnov
154 and Anderson-Darling tests. Shape α and rate β parameters for the Gamma distributions were
155 highly correlated and only the former is presented here (Table 4).

156 To account for spatial variability of surface texture, without specific measurements of flow re-
157 sistance, flow modelling was conducted twice for a given surface. First, we used a constant
158 Manning's n value of 0.045, and second, a spatially variable value that was back-calculated
159 using the flow resistance law presented by Ferguson (2007). A flow duration of 100 seconds
160 was sufficient to establish convergence. The results of the flow model were quantitatively val-
161 idated by comparing measured reach-averaged hydraulic depths ($h = A/w$, where A is flow
162 cross-sectional area and w is wetted width) to modelled ones (Figure 2). Most estimates fall
163 within 10–15 percent of the line of equality, although the flow model estimates a narrower range
164 of mean flow depths across the experiments. The flow model is likely a more accurate estimate
165 of flow depths compared to the stream gauge measurements as they are easily biased towards
166 either large or small values due to the relatively small sample size.

Table 3: Summary of mean hydraulics from Nays2DH models for the final DEM of each discharge phase. Flow depth is d , and Reynolds number $Re = Ud/v$, where v is the kinematic viscosity.

Exp	w [m]	d [m]	U [m/s]	Fr [-]	$\bar{\tau}$ [Pa]	Re [-]
Exp2a	0.07	0.016	0.36	0.92	2.47	4431
Exp2b	0.07	0.013	0.30	0.85	2.18	3026
Exp1a	0.26	0.015	0.36	0.93	2.68	4155
Exp1b	0.22	0.013	0.31	0.85	2.39	3166
Exp1c(1)	0.18	0.012	0.28	0.81	2.07	2556
Exp1c(2)	0.22	0.013	0.30	0.81	2.26	2997
Exp1c(3)	0.26	0.016	0.35	0.88	2.85	4167
Exp1c(4)	0.28	0.018	0.44	1.03	3.35	6127
Exp3a	0.48	0.015	0.34	0.87	2.69	3873
Exp3b	0.34	0.014	0.33	0.87	2.45	3613
Exp3c(1)	0.28	0.013	0.30	0.83	2.18	3027
Exp3c(2)	0.37	0.013	0.31	0.83	2.35	3190
Exp3c(3)	0.45	0.015	0.35	0.89	2.69	4090
Exp3c(4)	0.52	0.017	0.41	0.97	3.22	5400
Exp4a	0.56	0.015	0.35	0.89	2.73	3975
Exp4b	0.46	0.013	0.31	0.82	2.28	3188
Exp4c(1)	0.36	0.013	0.31	0.84	2.11	3129
Exp4c(2)	0.46	0.014	0.32	0.85	2.33	3391
Exp4c(3)	0.54	0.015	0.37	0.93	2.79	4377
Exp4c(4)	0.67	0.017	0.41	0.97	3.09	5413

2.2 1D and 2D sediment transport equations

We compared the observed mean transport rate over the last three hours of each experiment (twelve measurements at intervals of 15 min) to predictions from two sediment transport equations (1D and 2D). This approach assumes that the spatial distribution of shear stress remained relatively similar over the averaging period, which is supported by the lack of morphologic change and relatively stable mean hydraulics as estimated by gauge measurements. We used the Meyer-Peter and Müller (1948) equation (MPM)

$$q_s = k(\bar{\tau} - \tau_c)^{1.6} \quad (1)$$

where q_s is width-averaged sediment transport, and Wong and Parker (2006) estimate that $k = 4.94$. The parameter $\bar{\tau}$ is the mean shear stress, and the critical shear stress value is defined as $\tau_c = \theta_c g(\rho_s - \rho)D$, where θ_c is the dimensionless critical shear stress, g is gravity, ρ is the density of water, ρ_s is the density of sediment, and D is the grain size (median size D_{50})

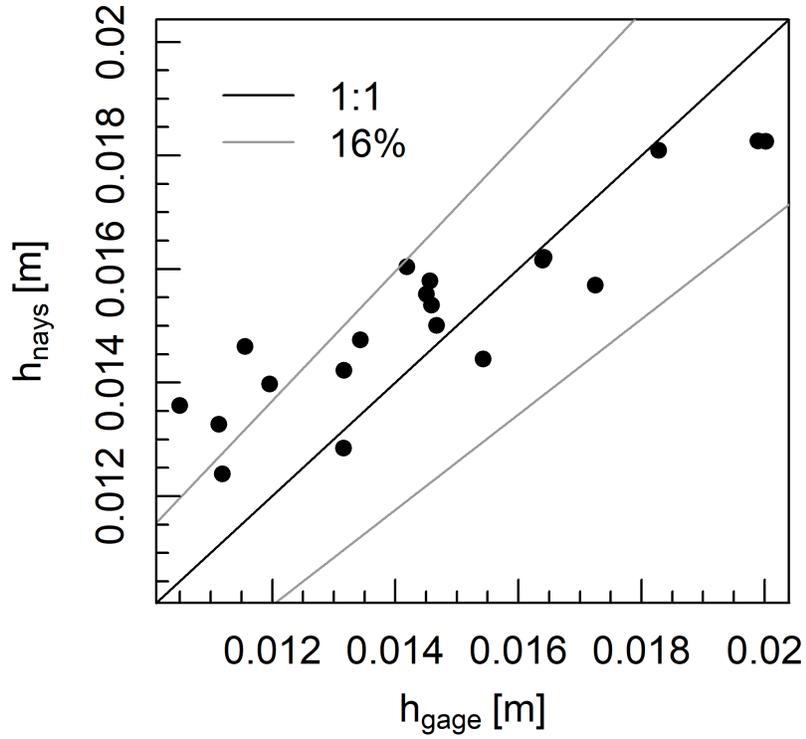


Figure 2: Measured versus modelled mean hydraulic depth h at the end of each experimental phase, featuring 16 percent bounds.

178 is used here). This 1D equation may be divided into flows greater than or less than the critical
 179 shear stress and expressed as

$$\begin{aligned}
 q_{s(x)} &= k(\tau_{(x)} - \tau_c)^{1.6} && \text{(for } \tau_{(x)} > \tau_c) \\
 &= 0 && \text{(for } \tau_{(x)} \leq \tau_c)
 \end{aligned}
 \tag{2}$$

180 and a 2D approach is derived by integrating across a known frequency distribution of shear
 181 stress

$$q_s = k \int (\tau_{(x)} - \tau_c)^{1.6} dx
 \tag{3}$$

182 These 1D and 2D equations provide width-averaged estimates of sediment transport using
 183 reach-averaged shear stress (Equation 1) and the entire frequency distribution of shear stress
 184 ($q_{b,\bar{\tau}}$ and $q_{b,f(\tau)}$, respectively). In the 1D approach, we optimised the values of θ_c and the coeffi-
 185 cient k based on a non-linear least-squares approach. This provided values of 0.053 and 3.4,
 186 respectively, which were then used for the 2D equation. A summary of experimental results is
 187 presented in Table 4.

Table 4: Summary of experimental results with sediment transport emphasis. Units: τ [Pa], q_b [kg/m/s]. For each experiment, all parameters represent mean values derived from the final DEM or associated Nays2DH model, except for q_b which is averaged over three hours.

Exp	w/d	$\bar{\tau}$	q_b	$q_{b,\bar{\tau}}$	$q_{b,f(\tau)}$	α
Exp2a	4.60	2.47	1.06	1.35	2.20	5.39
Exp2b	5.40	2.18	0.16	0.63	1.21	8.59
Exp1a	17.5	2.68	1.75	1.98	3.36	3.87
Exp1b	16.3	2.39	0.97	1.13	2.52	3.25
Exp1c(1)	15.3	2.07	0.34	0.43	1.50	3.13
Exp1c(2)	16.8	2.26	0.86	0.81	2.20	2.46
Exp1c(3)	16.6	2.85	1.81	2.56	4.09	3.72
Exp1c(4)	15.4	3.35	3.68	4.54	5.86	5.99
Exp3a	32.2	2.69	2.48	1.99	3.61	2.85
Exp3b	23.9	2.45	1.10	1.29	2.61	3.16
Exp3c(1)	21.6	2.18	0.53	0.64	1.69	3.31
Exp3c(2)	27.7	2.35	1.21	1.04	2.40	3.10
Exp3c(3)	29.7	2.69	2.01	2.01	3.71	2.60
Exp3c(4)	30.2	3.22	4.34	3.96	5.86	3.36
Exp4a	37.8	2.73	2.42	2.15	3.87	2.97
Exp4b	34.0	2.28	1.14	0.86	2.28	2.18
Exp4c(1)	27.4	2.11	0.49	0.50	1.60	2.49
Exp4c(2)	33.2	2.33	1.22	0.97	2.38	3.01
Exp4c(3)	35.0	2.79	2.52	2.35	3.83	2.90
Exp4c(4)	38.5	3.09	4.41	3.44	4.62	5.96

188 2.3 Analysing longitudinal scaling patterns

189 To explain the spatial patterns of entraining forces we compare the spatial patterns of local bed
 190 slope and flow depth that give rise to shear stress. To quantify and compare these patterns
 191 we use wavelet transform, which decomposes signals into oscillations occurring at different
 192 wavelengths (Torrence and Compo, 1998). Specifically, we use the maximal overlap discrete
 193 wavelet transform (MODWT) which has been used in a similar application (Adams and Zam-
 194 piron, 2020). Using the results from the hydraulic model, we located the primary flow path
 195 by identifying the highest shear stress at each cross-section, which was then smoothed in a

196 downstream direction by removing spatial outliers and applying a moving average (an example
197 is presented in Figure 4). This longitudinal transect, representing the primary flow path, was
198 then used to extract downstream profiles of local shear stress, bed slope, and flow depth.

199 **3 Results**

200 The experiments spanned a range of mean bed shear stresses and width-depth ratios, which
201 are presented in Figure 3. The middle two phases of the multiple-discharge experiments (cir-
202 cled), and the two constant discharge experiments which share the same imposed unit dis-
203 charge and maximum width, developed similar width-depth ratios and mean shear stress val-
204 ues. For the same unit discharge, the mean shear stress of the 0.08 m experiments was lower
205 compared to the wider experiments due to the sidewall effect whereby there was energy loss
206 to the lateral boundaries.

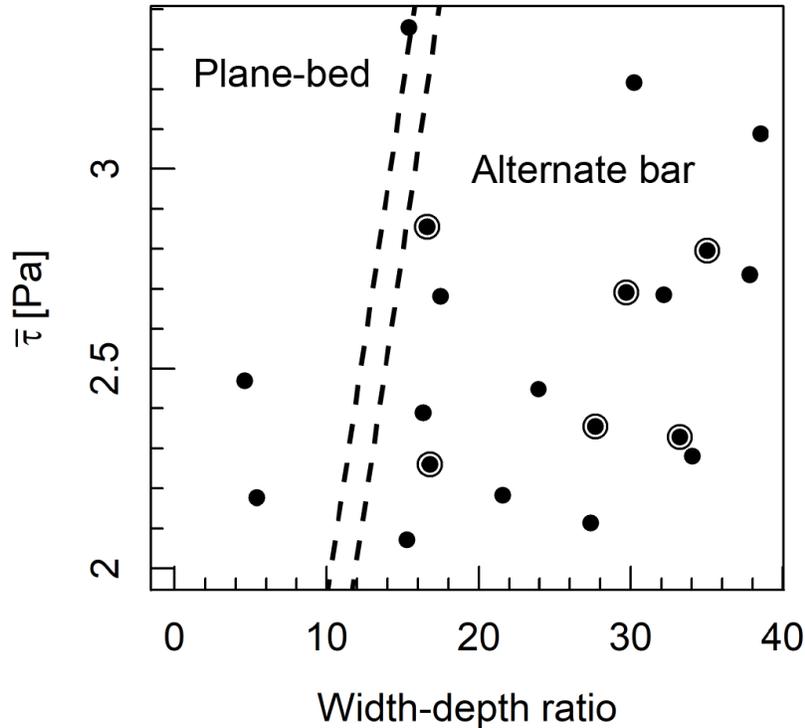


Figure 3: Mean bed shear stress and width-depth ratio at the conclusion of each experimental phase, estimated using the flow models. Circled points represent experimental phases 2 and 3 of the multiple discharge experiments which have the same imposed unit discharge and maximum width as the constant discharge experiments (Table 1). Dashed lines indicate approximate transition zone between plane-bed and alternate bars based on our experimental results, as well as width-depth ratio and excess shear stress thresholds summarised by (Rhoads and Welford, 1991).

207 With increasing width-depth ratio there was a transition from a plane-bed to an alternate bar
 208 morphology, and an example is presented in Figure 4 for the latter. In contrast, channel mor-
 209 phology was less sensitive to shear stress (or unit discharge) for a given channel width. As
 210 discharge was increased, there was an increase in bar wavelength and at the highest flow, the
 211 morphology was more topographically subdued. For simplicity, we classify the experiments into
 212 two groups based on morphology: 1) plane-bed channels comprising both 0.08 m wide chan-
 213 nels and the highest discharge 0.30 m wide channel, and 2) alternate bar channels (Figure
 214 3).

215 We observed different spatial patterns of surface texture across the experiments. Plane-bed

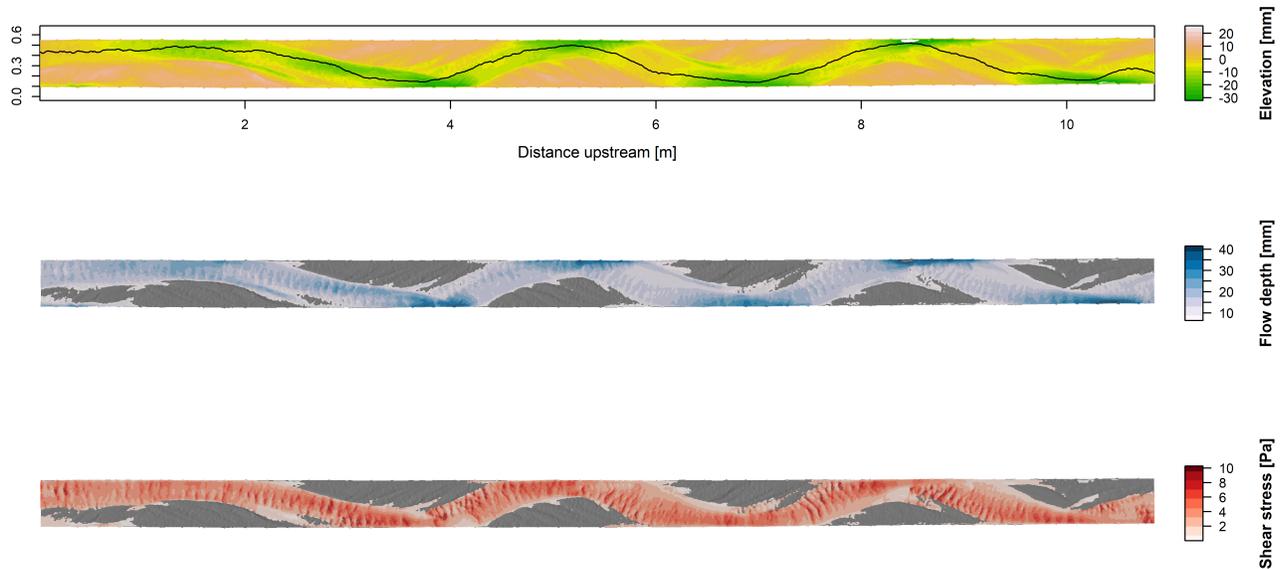


Figure 4: Channel area at the conclusion of Experiment 3b ($W = 0.45$ m, $\bar{\tau} = 2.41$ Pa) displaying characteristics (top to bottom): a) elevation, b) flow depth, and c) shear stress. Cells where $d < 2D_{84}$ have been removed from the hydraulic model. Thalweg transect is displayed as a black line.

216 channels were characterised by only longitudinal sorting patterns (alternating coarse and fine
 217 patches), whereas in the alternate bar channels fine patches were located around the channel
 218 thalweg and coarse patches were concentrated away from the primary flow path.

219 3.1 Longitudinal scaling patterns

220 In this section, we apply a wavelet transform to describe the longitudinal scaling patterns of
 221 flow depth, local bed slope, and shear stress along the primary flow path. The results of this
 222 analysis are presented in two ways: 1) the variance of each wavelength, and 2) the cumulative
 223 distribution of variance showing the relative contribution of each scale to the total.

224 At wavelengths shorter than 0.1 m, plane-bed and alternate bar morphologies are indistinguish-
 225 able based on the variance in flow depth, local bed slope, and shear stress. Differences emerge
 226 at longer wavelengths, where in the plane-bed channels there is considerably less variance in
 227 flow depth and local bed slope, compared to the more topographically variable alternate bar
 228 channels. These contrasting scaling patterns are evident in both the absolute and cumulative
 229 representations, which show two groupings of channels. In the longer wavelength scaling pat-
 230 terns of shear stresses the binary grouping gives way to a graded distribution, in which there is
 231 less variation in narrow channels compared to wider ones.

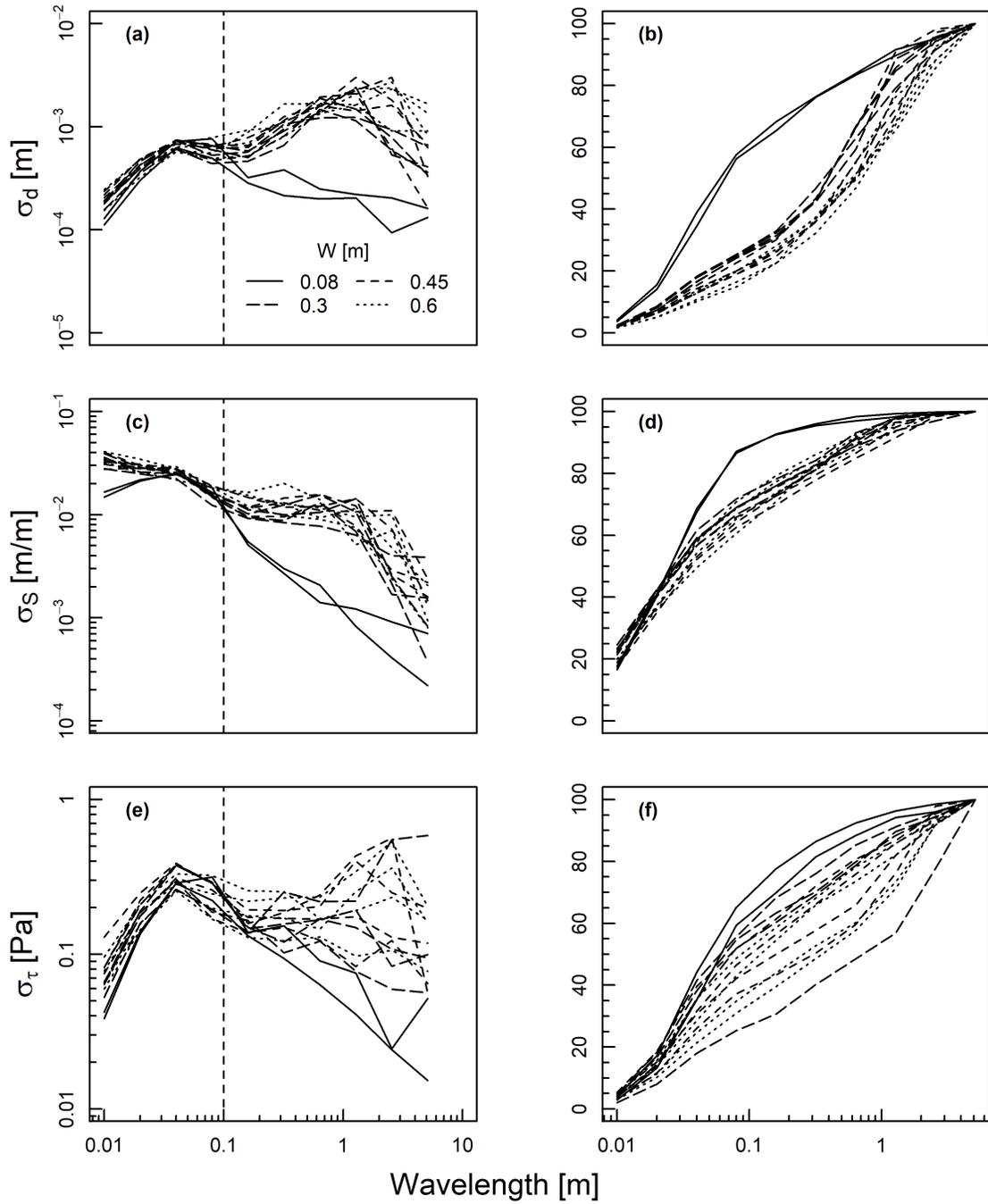


Figure 5: Scaling patterns of standard deviation in local flow depth, bed slope, and shear stress for each unique experimental phase, presented as both absolute (a, c, e) and cumulative values (b, d, f), respectively. For brevity, phases 2 and 3 of the multiple discharge experiments are not shown (circled points in Figure 3), so only experiments with unique unit discharge and maximum width combinations are presented. The line style refers to the channel width, and the vertical dashed line refers to the scale below which all channel follow similar scaling patterns.

232 3.2 Spatial concentration of flow

233 The two morphologic types are well distinguished by their mean-normalised frequency distribu-
234 tions of flow depths (Figure 6a). Plane bed channels have peaked distributions of depths with
235 relatively short tails, whereas alternate bar channels have broader distributions corresponding
236 to areas of relatively shallow and deep flow. This grouping is also evident in the normalised
237 frequency distributions of shear stress, the difference between them is lesser (Figure 6b). The
238 two plane-bed channels, as well as the channel with subdued bars, have Gamma distributions
239 with larger shape and rate parameters (Table 4), which is consistent with their distributions
240 being less positively skewed and more peaked, respectively. There is no systematic variation
241 between the shape of the distribution and the mean shear stress.

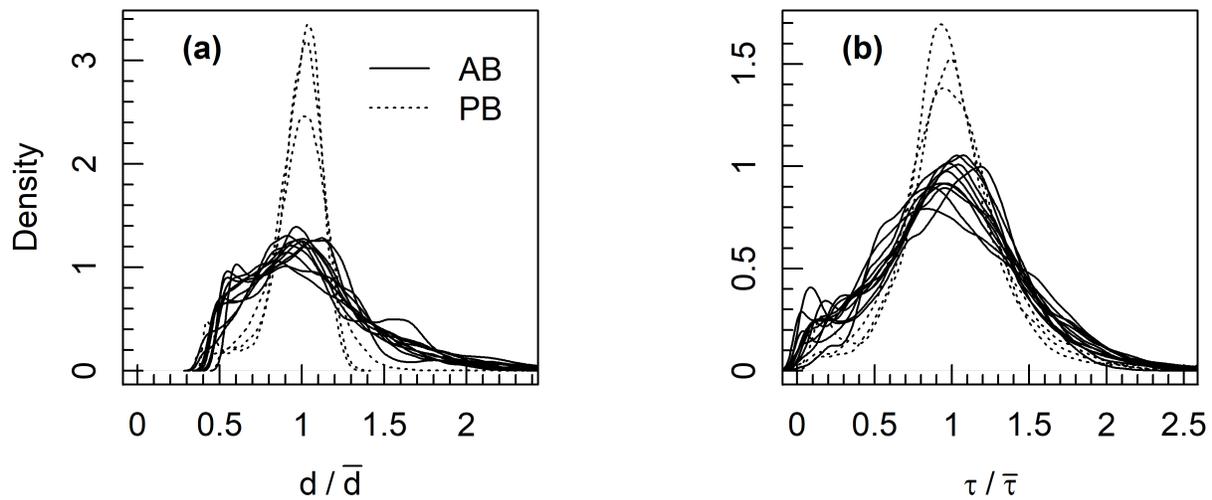


Figure 6: Frequency distribution of mean-normalised flow depth and shear stress at the end of each unique experimental phase (see Figure 5 for note on excluded data). AB and PB correspond to alternate bar and plane-bed morphology, respectively. Note the absence of shallow depths which have been removed during data processing to eliminate non-active channel areas.

242 3.3 Sediment transport

243 All experiments reached a steady-state sediment transport rate whereby individual measure-
244 ments oscillated around a mean value, although temporal patterns of transport varied between

245 channel morphologies. We present a comparison between two channels with contrasting
 246 sources and magnitudes of fluctuations, but similar transport capacities when averaged over
 247 time (Figure 7). In the plane-bed example, the presence of coarse and fine patches of sediment
 248 produced a spatially variable transport rate, and their downstream migration (i.e., position rela-
 249 tive to the outlet) resulted in oscillating volumes of sediment output, ranging from approximately
 250 0.0–3.0 kg/m/min under steady-state conditions. In alternate bar example, bedload transport
 251 had less variation through time, ranging from approximately 0.75–1.75 kg/m/min, which was
 252 likely associated with morphologic activity at the bar-scale.

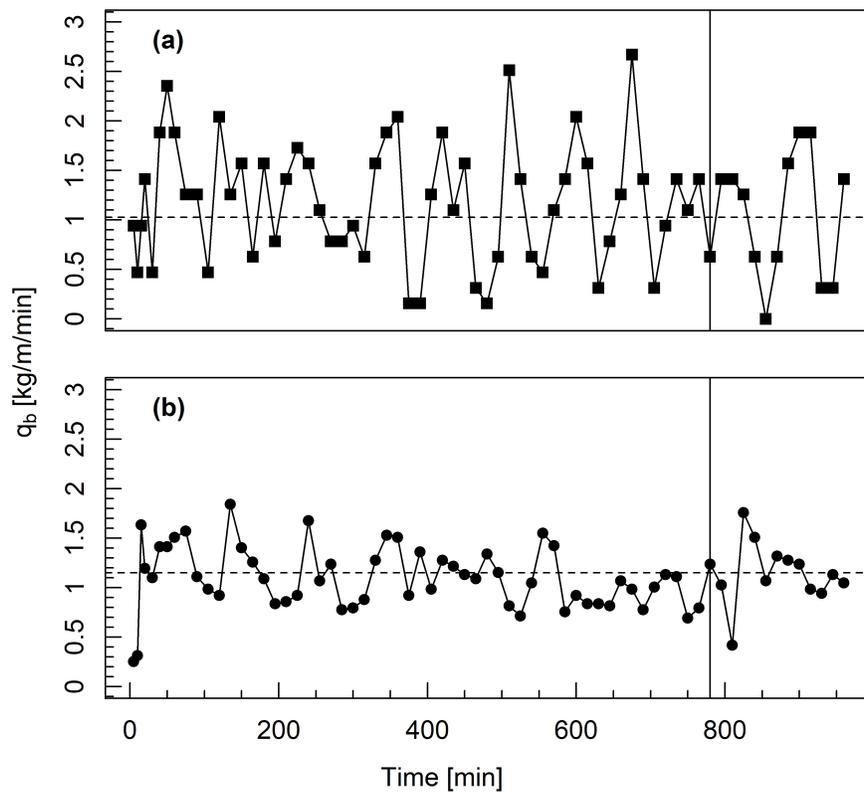


Figure 7: Width-averaged bedload transport over time in two experiments with different widths but similar reach-averaged shear stress and transport capacity: a) Experiment 2a ($W = 0.08$ m, $\bar{\tau} = 2.45$ Pa), and b) Experiment 4b ($W = 0.60$ m, $\bar{\tau} = 2.22$ Pa). The beginning of the time window over which bedload transport is averaged is indicated by the solid vertical line, and mean transport over this period is indicated by a horizontal dashed line.

253 We compare observed time-averaged sediment transport to predictions made by two versions
 254 of the MPM equation, which represent 1D (Equation 1) and 2D (Equation 3) approaches (Fig-
 255 ure 8). The 1D approach was calibrated using the coefficient k and the critical value τ_c , and
 256 therefore predictions are located along the line of equality. Using the same coefficients the 2D

257 predictions lie approximately parallel to but at an intercept above the line of equality. However,
 258 both methods predict sediment transport almost equally well ($R^2 = 0.89$ & 0.90 , $RMSE = 0.38$
 259 & 0.42 , respectively), yielding strong correlations between observed and predicted q_b .

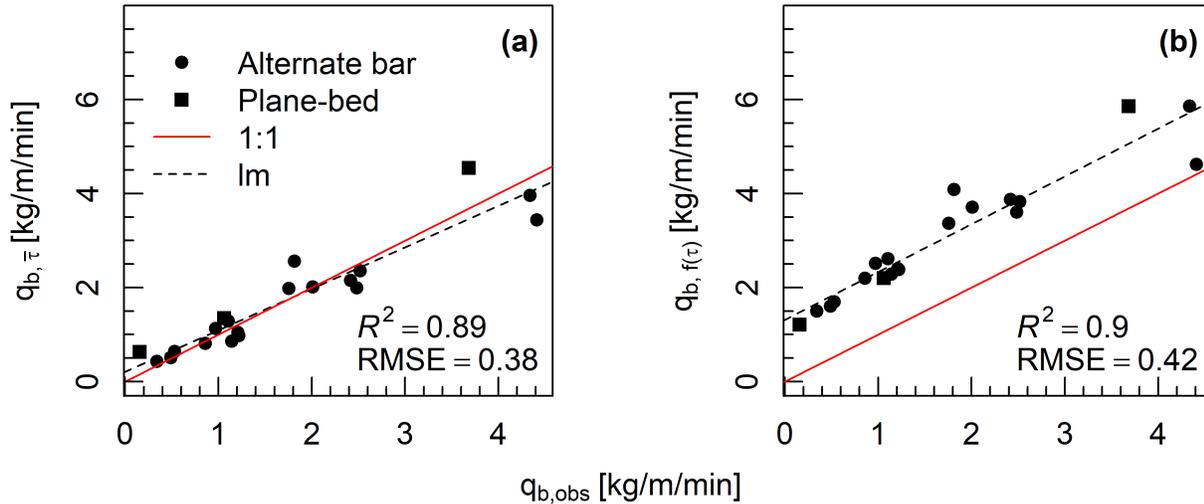


Figure 8: Observed vs predicted sediment transport in each experiment, where the time- and width-averaged bedload transport rate q_b is predicted using two MPM equations; a) Equation 1 (1D) and Equation 3 (2D). The dashed black line is least-squares best fit, the red solid line corresponds to a 1 : 1 relation between observed and predicted volumes, and point type refers to morphology.

260 4 Discussion

261 4.1 Width-depth ratio and channel character

262 The suite of experiments, comprising width-depth ratios from 5–45, and mean bed shear stress
 263 values approximately 2–3.3 Pa, produced two primary channel morphologies. With increas-
 264 ing width-depth ratio, there was a transition from plane-bed to alternate bar. With increasing
 265 mean shear stress, there was an increase in bar wavelength (and decrease in amplitude) such
 266 that channels with the highest unit discharge had a more subdued morphology. The devel-
 267 opment of alternate bars in wider channels with lower excess shear stresses is supported by
 268 several investigations (Fujita, 1989; Garcia Lugo et al., 2015; Carbonari et al., 2020; Rhoads
 269 and Welford, 1991). Notably, our experiments conform to the threshold $w/d \approx 10$ (Chang et al.,

270 1971; Ikeda, 1984).

271 The spatial patterns of morphology and surface texture were coupled. Plane-bed channels
272 exhibited only longitudinal variation in surface texture, with alternating coarse, intermediate,
273 and fine patches, similar to Iseya and Ikeda's (1987) observations of congested, transitional,
274 and smooth states. The spatial distribution of surface texture in alternate bar morphologies was
275 characterised by a repeated pattern of relatively coarse bars and a fine thalweg. This is the
276 reverse of patterns typically observed in sinuous channels, where bars form by the deposition
277 of fine sediment as flow separation arises at bends. This process cannot occur as the fixed
278 straight channel prevents meandering, and under these conditions, bars can only develop via
279 the deposition of large grains. Subsequently, the experiments presented here best replicate
280 processes in relatively straight or confined reaches in nature.

281 The mean-normalised frequency distributions of shear stress followed Gamma distributions,
282 whose shape varied across the two morphologies. The Gamma distribution shape has been ob-
283 served in meandering (Segura and Pitlick, 2015; Monsalve et al., 2020) and braided (Nicholas,
284 2003) channels, although, in contrast to the analysis herein, these results were obtained by
285 modelling a range of flows over the same bathymetry (i.e., the channel boundary could not
286 adjust). In the alternate bar channels, the frequency distribution of shear stress was more
287 positively skewed and less peaked with heavier tails, indicated by lower α and β parameters.
288 Previous studies have observed both positive (Segura and Pitlick, 2015) and negative (Mon-
289 salve et al., 2020) correlations between flow strength and α , whereas there was no correlation
290 in these experiments. These conflicting results may indicate that the relationship between the
291 distribution of τ and flow strength is more dependent on the specific shape of the channel.

292 **4.2 Decoupling of flow depth and shear stress**

293 Alternate bar and plane-bed morphologies were strongly differentiated by their frequency dis-
294 tributions of flow depth (Figure 6). The former was characterised by broad distributions as the
295 flow was associated with pools and riffles, comprising areas of relatively deep and shallow flow,
296 respectively. Although the two morphologies could be distinguished based on the frequency
297 distribution of shear stress, the difference was more subtle as shear stress was relatively spa-
298 tially homogeneous in the alternate bar morphology, which is evident in the hydraulic maps
299 (Figure 4).

300 Using the wavelet transform it is possible to describe how the differences between shear stress
301 and flow depth manifest at different spatial scales. Across the suite of experiments, scaling

302 patterns of local bed slope, flow depth, and shear stress were almost identical at wavelengths
303 less than 0.10 m (Figure 5). Differences between experiments emerged on larger scales, where
304 scaling patterns of local bed slope and flow depth were grouped by morphology. However, the
305 scaling patterns of bed shear stress were less distinct between the two morphologies and,
306 rather, there was a gradient of variance at longer wavelengths between them. Thus, both
307 frequency distributions and scaling patterns highlight a decoupling of flow depth and shear
308 stress between the two morphologies at relatively large spatial scales. This result challenges
309 the commonly held assumption that local deviations from $\tau \propto d$ tend to cancel out so that their
310 frequency distributions are similar (Nicholas, 2000; Ferguson, 2003; Bertoldi et al., 2009).

311 There were similarities in the scaling patterns of local bed slope, flow depth, and shear stress
312 across all experiments (Figure 5a,c). There were almost identical patterns of variance at shorter
313 wavelengths and large differences at longer wavelengths, and the wavelengths that separated
314 these two scaling regions was always located around 0.10 m. The consistency of this scaling
315 boundary across the three parameters is explained by their co-dependency, although its exist-
316 tence may be ascribed to more fundamental processes. The presence of two distinct scaling
317 regions highlights a process decoupling whereby grain-scale processes appear only indirectly
318 related to ones operating at the channel scale. This characteristic of natural self-organised
319 systems has been discussed in the context of the emergence perspective in geomorphology
320 (Werner, 2003; Murray, 2007). The theory describes an indirect relationship between processes
321 operating at different spatio-temporal scales, where the behaviour of the emergent aspects
322 of the system (the morphology) is decoupled from the behaviour of system constituents (the
323 grains). Thus, our experiments provide evidence that supports a hierarchical view of processes
324 and forms in geomorphic systems.

325 **4.3 Feedbacks that control transport capacity**

326 Under the relatively high discharges modelled in our experiments, transport capacity was well
327 predicted by both 1D and 2D Meyer-Peter Müller equations based on the strength of the cor-
328 relation. This result is consistent with previous investigations concerned with the performance
329 of bedload transport equations across various stages (Bertoldi et al., 2009; Monsalve et al.,
330 2020), and suggests that under formative discharges where most geomorphic change occurs
331 1D approaches may be sufficient to provide accurate estimates of sediment transport, provided
332 that the data is averaged over a long enough period (Recking et al., 2012). Several studies
333 have observed that at low flows morphology (via the spatial concentration of shear stress) acts
334 to increase transport capacity, and the effectiveness of the 1D approach herein supports the
335 notion that the strength of this effect is inversely proportional to the flow stage (Paola and Seal,

336 1995; Paola, 1996; Nicholas, 2000; Ferguson, 2003).

337 The specific processes that underlie the stage-dependent relationship between morphology
338 and bedload transport are not yet clear, although our experiments shed light on some important
339 morphodynamics. In the alternate bar channels, transport capacity (via the spatial distribution
340 of shear stress) was likely controlled by a negative feedback between flow depth and local bed
341 slope. The primary flow path encountered alternating positive and negative slopes (associated
342 with pool heads and tails), and their correlation at approximately the scale corresponding to
343 pool spacing is evident in the scaling patterns of flow depth and local bed slope (Figure 5). The
344 interaction between these two parameters acts to reduce shear stress where flow is deepest,
345 and constrains the maximum shear stress across the active channel. This explains both the
346 relatively homogeneous spatial distribution of shear stress compared to flow depth (Figure 4),
347 and the convergence of frequency distributions of shear stress for both alternate bar and plane-
348 bed channels (Figure 6).

349 The negative feedback between flow depth and local slope indicates that at high flow stages
350 channels may tend to expend excess flow energy and shear stress via the development of
351 morphology. At lower flows, bedload transport is slaved to morphology because the system
352 lacks surplus energy to instrument the morphologic change necessary for such a negative
353 feedback.

354 Moreover, it is interesting that the plane-bed channels (especially the 0.08 m width) had a
355 similar transport capacity to the alternate bar channels, despite lacking the same degrees-of-
356 freedom available for morphologic adjustment. Instead, in the narrowest channels, bedload
357 transport arose via the longitudinal distribution of surface texture that controlled the threshold
358 for entrainment. The initiation of transport coincided with the concentration of fine sediment
359 into relatively homogeneous downstream migrating patches, which also caused considerably
360 larger fluctuations in output compared to the alternate bar channels (Figure 7). Integrated
361 across space and time, these migrating patches gave rise to a similar transport capacity to the
362 alternate bar for the same mean shear stress. The contrasting spatial patterns of morphology,
363 hydraulics, and surface texture across the experiments highlights the potential for transport
364 capacity to be equifinal under formative discharges.

365 Both the experiments and bedload transport equations presented herein have limitations that
366 provide opportunities for further research. By constraining channel pattern and preventing
367 lateral adjustment, the fixed-bank experiments also restrict channel processes. More mobile-
368 bank experiments are required to understand how feedbacks between channel processes give
369 rise to the system's transport capacity. By using only the frequency distribution, so-called
370 2D equations remove the spatial dimension of transport and are only quasi-2D. Processes

371 of sediment entrainment, transport, and disentrainment are affected by local conditions that
372 control the trajectories of grains downstream, and further research must aim to account for this.

373 **5 Conclusion**

374 We compared the performance of 1D and 2D bedload transport equations under formative dis-
375 charge conditions. These flows are particularly relevant for river management as they encom-
376 pass large volumes of transport and geomorphic change. Physical models with varying channel
377 width-depth ratios and discharges (but identical reach-averaged gradients and bulk grain sizes)
378 produced a range of emergent morphologies and steady-state transport capacities under re-
379 circulating conditions. Transport capacity was well predicted by both 1D and 2D Meyer-Peter
380 Müller equations, which is consistent with previous studies indicating that 1D equations may
381 be effective under high discharge conditions when measurements are appropriately temporally
382 averaged.

383 Our experiments contribute to an understanding of the stage-dependent relationship between
384 morphology and bedload transport, and specifically why the mean shear stress characterises
385 transport capacity so effectively at formative discharges. In channels capable of building bars,
386 the relationship may be explained by a negative feedback between flow depth and local bed
387 slope. This feedback results in a spatial distribution of shear stress that is relatively homoge-
388 neous compared to flow depth, and restricts the maximum shear stress available across the
389 active channel. Despite lacking the same degrees-of-freedom available for morphologic ad-
390 justment, narrower plane-bed channels maintained the same relationship between transport
391 capacity and mean shear stress via a spatially variable migrating surface texture. The contrast-
392 ing spatial patterns of morphology, hydraulics, and surface texture between the two channel
393 morphologies highlight the potential for similar system properties to emerge through entirely
394 different mechanisms.

395 The effectiveness of 1D bedload transport equations in channels where there are feedbacks
396 between channel morphology, sediment transport, and hydraulics highlights the potential for
397 non-linear dynamics to drive linear behaviour. This aspect of fluvial systems explains why sim-
398 ple deterministic equations can be highly effective under certain conditions. Further work is
399 required to examine how additional processes such as lateral adjustment affect morphodynam-
400 ics.

401 **Open Research**

402 Raw hydraulic and sediment transport data from Tables 3-4 and Figures 2-3 & 6-8 are available
403 at Zenodo [DOI: 10.5281/zenodo.5750653] with an Open license (Adams, 2021).

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