



RiverBedDynamics v1.0: A Landlab component for computing two-dimensional sediment transport and river bed evolution

Angel D. Monsalve¹, Samuel R. Anderson², Nicole M. Gasparini², Elowyn M. Yager¹

5 ¹Center for Ecohydraulics Research, Department of Civil and Environmental Engineering, University of Idaho, Boise, ID, USA

²Department of Earth and Environmental Sciences, Tulane University, New Orleans, LA, USA

Correspondence to: Angel Monsalve (amonsalve@uidaho.edu)

10 **Abstract.** Computational landscape evolution models (LEMs) typically comprise at least two coupled components: a flow hydraulics solver that routes water across a landscape and a fluvial geomorphological model that modifies terrain properties, primarily bed surface elevation. LEMs used in long-term simulations over large watersheds, including some available in the Landlab library, often assume that only erosive processes occur in rivers and that terrain elevation increases solely due to tectonic uplift. Consequently, these models cannot capture the dynamics of gravel-bedded rivers, lacking the capacity to

15 include sediment mixtures, simulate sediment deposition, and track textural changes in substrate stratigraphy that result from varying flow characteristics. To address this limitation, we developed, implemented, and tested RiverBedDynamics, a new Landlab component that simulates the evolution of bed surface elevation and grain size distribution in two-dimensional grids based on the Exner equation for sediment mass balance. By dynamically coupling RiverBedDynamics with Landlab's hydrodynamic flow solver, OverlandFlow, we created a new LEM capable of simulating the dynamics of local shear

20 stresses, bed load transport rates, and grain size distributions. Comparisons of our LEM results with analytical and previously reported solutions demonstrate its ability to accurately predict time-varying local changes in bed surface elevation, including erosion and deposition, as well as grain size distribution. Furthermore, application of our LEM to a synthetic watershed illustrates how spatially variable rainfall intensity leads to varying discharge patterns, which in turn drive changes in bed elevation and grain size distribution across the domain. This approach provides a more comprehensive

25 representation of the complex interactions between flow dynamics and sediment transport in gravel-bedded rivers, enhancing our ability to model landscape evolution across diverse geomorphic settings.



30 1 Introduction

Landscape Evolution Models (LEMs) are fundamental tools for geomorphologists allowing researchers to understand landscape morphology produced under different climatic and tectonic circumstances, and in some cases, inform management decisions (Coulthard, 2001). These models simulate the long-term development of landscapes by incorporating multiple geomorphic processes, providing insights into how terrains change over time under various environmental conditions.

35 Currently available LEMs differ in the number of physical processes considered, the way they route water and sediment across a landscape, and how the domain and its features are represented when solving the governing equations (Coulthard, 2001; Temme et al., 2017; Tucker & Hancock, 2010). Most LEMs, designed to simulate long time periods, incorporate significant simplifications to make computations feasible. These simplifications often include steady-state flow assumptions, omission of grain size variations, and in many cases, exclusion of sediment deposition processes. For instance, one of the
40 earlier LEMS, GOLEM, (Tucker & Slingerland, 1994) assumed a steady single flow direction with water discharge defined as the product of drainage area and rainfall rate. This approach is still common in LEMs because it greatly simplifies calculations (e.g., (Braun & Willett, 2013; Campforts et al., 2017; Goren et al., 2014; Mitchell & Forte, 2023; Tucker et al., 2001). In contrast, CAESAR (Coulthard et al., 2002) uses a more complex approach, employing a routing scanning algorithm that allows for multiple flow directions and incorporates flow variability, enabling the simulation of non-steady
45 state flow conditions and flood wave routing.

In terms of representing drainage networks and channels within a catchment, LEMs employ various strategies to address the challenge of scale. This is crucial because the model grid resolution significantly impacts how different landscape features—such as channels, floodplains, and hillslopes—are represented and captured by the model, and consequently, how governing
50 equations are solved. For example, the CHILD model (Tucker et al., 2001) uses an adaptive triangulated irregular mesh to accurately capture transitions between different landscape elements, particularly the boundaries between channels and floodplains. This approach allows for a more detailed representation of channel networks, a characteristic that is not achievable when using uniform rectangular grid elements. In contrast, CAESAR employs a finer grid resolution near and within channels, especially at their boundaries to capture the different elements. This method effectively concentrates
55 computational resources where they are most needed for precise flow and sediment transport calculations. The choice of model complexity and resolution has significant implications for the timescales over which LEMs can operate effectively. Simpler models with more assumptions can simulate geomorphological changes over millennia to millions of years with relatively short computation times, but at the cost of reduced accuracy and precision. On the other hand, more detailed models like SedFoam (Cheng et al., 2017) for instance, can predict small-scale phenomena such as sand concentrations in
60 the water column at a submillimeter scale, but are computationally expensive and typically limited to shorter timescales and smaller spatial extents.



While many LEMs have contributed significantly to our understanding of landscape evolution, they often rely on simplifications that limit their applicability to certain geomorphic contexts. A common assumption is that erosive river processes and tectonic uplift are the primary factors shaping a landscape over long time scales (Campforts et al., 2017; Forte et al., 2016; Langston & Tucker, 2018; Li et al., 2018; Whipple et al., 2017). This approach has been particularly successful in modeling bedrock channels, where the rate of sediment removal is limited by the detachment of material from the bed (detachment-limited conditions). However, this simplification is not adequate when applied to gravel-bedded rivers, where the rate of sediment transport is limited by the flow's capacity to move sediment (transport-limited conditions) (e.g., Attal et al., 2011; Gasparini et al., 2004; Whipple & Tucker, 2002). The evolution of alluvial channel geometries in gravel-bed rivers depends on both erosion and deposition, processes intricately linked to grain size distributions and their evolution over time. Very few LEMs include explicit treatment of grain sizes, yet this factor is critical for accurately simulating sediment transport and channel morphodynamics. The ability to link grain size evolution with hydrograph variations is key to modeling realistic shear stress values and the overall behavior of the river system.

Incorporating deposition alongside erosion in LEMs as a mass conservation problem introduces additional complexity to model development. This approach requires conducting a mass balance at individual cells or control volumes, significantly increasing the computational demands of the model. Thus, LEMs that have adapted a mass balance approach have generally relied on simplified hydrology (e.g., Attal et al., 2011; Gasparini et al., 2004; Whipple & Tucker, 2002). These studies prioritize large-scale trends over long time scales, rather than exploring detailed bed evolution. However, accurately modeling erosion and deposition of different grain sizes remains crucial for understanding gravel-bed river dynamics. This requires a more accurate representation of flow velocity and depth, as these variables are used for calculating sediment transport capacity and determining whether deposition or erosion occurs at any given location.

The need for new models to accurately model grain size distributions presents a significant opportunity for advancing our understanding of landscape evolution, particularly in the context of gravel-bed rivers. Gravel-bed rivers are of paramount importance in geomorphology due to their role in sediment routing, their sensitivity to changes in sediment supply and flow regimes, and their close coupling with hillslope processes. Developing a model that can simulate the evolution of gravel-bed streams in a continuum framework, while also linking this evolution to broader landscape processes, would represent a major step forward in the field.

While the limitations of existing LEMs are evident, recent advancements in modeling frameworks offer new opportunities to address these challenges. One such framework is Landlab, a Python-based platform designed for creating, assembling, and running 2D landscape evolution models (Barnhart, Hutton, Tucker, Gasparini, et al., 2020; Hobbey et al., 2017). Landlab's modular structure allows researchers to combine various components, each representing different geomorphic processes, to create customized LEMs tailored to specific research questions. Recent efforts within the Landlab framework have made



significant strides in improving the accuracy of flow routing and erosion modeling. For instance, Adams et al. (2017) coupled the OverlandFlow component with DetachmentLtdErosion to investigate watershed incision patterns. The OverlandFlow component simulates surface water flow across a landscape, while DetachmentLtdErosion models the erosion of bedrock or cohesive sediment. However, while this approach represents an improvement in flow routing accuracy, it cannot be directly used in modeling gravel-bedded rivers. The DetachmentLtdErosion component, based on the stream power law, does not account for the complex dynamics of sediment transport and deposition characteristic of gravel-bed systems. Specifically, it doesn't consider the movement of different grain sizes or the process of sediment deposition, both of which are crucial in gravel-bed river evolution. An alternative approach within Landlab is the NetworkSedimentTransporter component (Pfeiffer et al., 2020). This Lagrangian model predicts changes in bed material grain size and river bed elevation based on bed load estimates within a predefined river network. While it offers improved accuracy for long-term simulations of sediment dynamics, its applicability is somewhat limited. The model focuses solely on the river network, which must be defined beforehand and remains static throughout the simulation. Consequently, it cannot model bed surface changes across the entire watershed, including areas outside channels. This constraint limits the model's ability to capture the broader landscape-river interactions crucial for understanding overall landscape evolution.

These existing components, while valuable, highlight a critical gap in our ability to model gravel-bed rivers within the context of landscape evolution. What's needed is a component that can: i) Accurately represent sediment transport dynamics in gravel-bedded rivers, ii) Account for fractional sediment transport including erosion and deposition processes, iii) Predict bed surface changes across an entire watershed, and iv) Integrate with high-accuracy flow prediction under non-steady and non-uniform conditions.

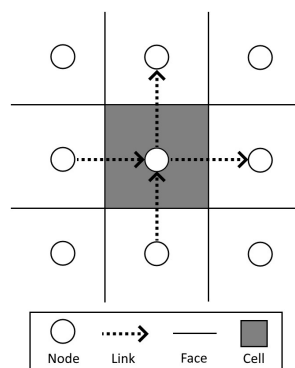
To address this gap, we propose a new Landlab component: RiverBedDynamics. This component is designed to simulate the evolution of gravel-bed streams in a continuum model, allowing for the integration of the channel with other geomorphological processes using the Landlab platform. In our case, by coupling RiverBedDynamics with OverlandFlow, we enable the simulation of non-steady flow conditions, which is crucial for understanding the complex dynamics of gravel-bed rivers. In this article, we introduce RiverBedDynamics and demonstrate its capabilities in modeling gravel-bed river evolution within a landscape context. We show how this new component addresses the limitations of existing models by incorporating grain size evolution, non-steady flow conditions, and watershed-wide predictions of bed surface changes. Through a series of tests and applications, we illustrate how RiverBedDynamics can capture key aspects of landscape evolution, particularly in systems dominated by gravel-bed rivers. All sediment transport predictions are based on the unsteady total shear stress, which accounts for spatial and temporal gradients in flow velocity and local variations in bed elevation and water depth. Evaluations of our component are conducted using test cases with analytical solutions from previously available models. An example in a large watershed is used to explore large scale applications of the component.



130 2 A general overview of the Landlab modeling approach

Landlab, a Python-based interdisciplinary open-source platform, serves as a robust framework for developing computational landscape models addressing earth surface dynamic processes (Barnhart, Hutton, Tucker, M. Gasparini, et al., 2020; Hobley et al., 2017; Tucker et al., 2022). We integrate our component into Landlab, leveraging its seamless support for incorporating new process components. Further, Landlab already contains a simplified hydrodynamic model for computing flow variables, upon which RiverBedDynamics relies (Adams et al., 2017). Numerous studies detail the general structure of Landlab (e.g., Adams et al., 2017; Barnhart et al., 2019, 2020; Hobley et al., 2017; Shobe et al., 2017; Tucker et al., 2022), therefore, we focus here solely on aspects relevant to implementing the RiverBedDynamics component.

At the core of our component lies Landlab's gridding engine, facilitating data manipulation and exchange among various components. This engine operates on a 2D structured grid, enabling numerical operations essential for calculating flow, bed surface changes, and sediment variables during simulations. For instance, the grid contains methods for computing topographic gradients, sediment mass balance, and mapping velocities from grid links to nodes. Currently, our component exclusively operates on raster grids (Figure 1). Within this grid framework, nodes represent discrete (x,y) points, while links denote vectors connecting neighboring nodes with fixed directionality. A cell, bounded by faces, encapsulates the area around a non-perimeter (i.e., interior) node. All cells within our component are rectangular-shaped and maintain uniform dimensions in both the x (Δx) and y (Δy) directions.



150 Figure 1: Elements of a Landlab grid used by our component, illustrating a typical grid segment. Information is stored in nodes and links. For instance, surface bed elevation data is held at the nodes, whereas the gradients of these elevations are stored in the links.



Our component was developed around the `RasterModelGrid` class, chosen for its ability to facilitate the numerical solution of
155 partial differential equations, such as the Exner equation for sediment mass conservation, and spatially variable processes in
a straightforward manner. For instance, let's examine the Meyer-Peter & Müller (1948) equation for bed load transport:

$$q_b^* = \begin{cases} 8(\tau^* - \tau_c^*)^{3/2}, & \tau^* > \tau_c^* \\ 0, & \tau^* \leq \tau_c^* \end{cases} \quad \text{Eq. 1}$$

where $\tau_c^* = 0.047$ is the dimensionless critical shear stress, q_b^* is the dimensionless volumetric bed load transport rate per
unit width, and τ^* is the dimensionless shear stress. Implement Eq. 1 involves calculating τ^* at each node and determining
locations where τ^* exceeds τ_c^* to compute q_b^* . Using the structure and tools in Landlab Eq. 1 can be coded as:

160 `mask = tau_star > tau_star_cr`

`qb_star[mask] = qb_star_coeff * (tau_star[mask] - tau_star_cr[mask]) ** (qb_star_exp)`

here, `tau_star` is extracted from the `grid`, while `tau_star_cr`, `qb_star_coeff`, and `qb_star_exp` are the equation's parameters (
0.047, 8, and 3/2, respectively). The mathematical structure in Landlab closely resembles the form of Eq. 1, enabling the
construction of a bed load transport model applied across the entire domain without the need for iterating over grid indices.

165 Additionally, data exchange between different Landlab components is seamlessly facilitated by the `grid`. For instance,
modifications to the bed surface elevation in `RiverBedDynamics` result in the updated value being immediately available to
all Landlab components via the field `grid["node"]["topographic_elevation"]`.

Boundary conditions in our component are inherited from the Landlab `grid` object, aligning with those specified in the
170 `OverlandFlow` component (Adams et al., 2017). Nodes defined as 'boundary' can be designated as either open, fixed
gradient, or closed. Open boundary nodes allow flux to enter or leave the model domain, acting as flow outlets. Closed
boundary nodes prevent flux from entering or leaving the domain. This classification determines the behavior of surface
water flow at these boundary nodes, with sediment fluxes calculated based on local flow conditions. Links connected to
these boundary nodes are automatically classified as active, inactive, or fixed. Active links, where fluxes are calculated,
175 occur between two core nodes or between a core node and an open boundary node. Inactive links, where no fluxes are
calculated, occur between a closed boundary node and a core node or between any pair of boundary nodes. Fixed links,
which can have assigned values, occur between a fixed gradient node and a core node. The sole exception to these inherited
conditions is at the domain outlet. Here, `RiverBedDynamics` requires specifying either a fixed bed surface elevation or a
zero-gradient condition (refer to section 3.4 for more details).

180 **3 Model description**

Our component was designed to be coupled to a flow solver such that continuous feedback between surface flow and river
bed properties determines the behavior of the system. In our implementation, we have utilized `OverlandFlow` as the flow
solver (Figure 2), but Landlab's "plug-and-play" capabilities allow for compatibility with any flow solver. At each time-step,



the flow governing equations are solved across the entire domain by OverlandFlow, obtaining flow depth, velocity, and discharge. The routines of RiverBedDynamics can be conceptualized in two major parts: i) bed load transport and ii) river bed evolution. In the first part, RiverBedDynamics processes surface flow and bed surface grain size properties variables stored in the grid to calculate local shear stress and bed load transport rate. In the second part, it uses sediment fluxes entering and leaving each cell to compute the mass balance. This process updates the bed surface elevation and bed properties, such as grain size distribution, thereby completing the cycle at each time step.

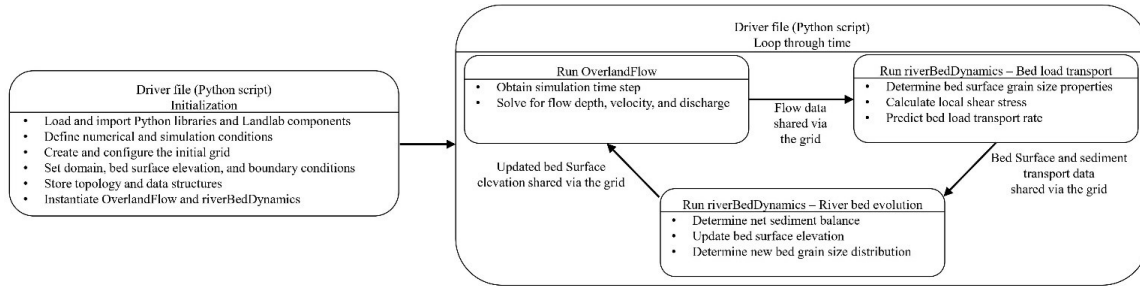


Figure 2: Simplified workflow for the coupled OverlandFlow and RiverBedDynamics routine. The driver file is a procedure script containing the set of instructions to create all the required data and loop through time, dynamically linking and updating surface flow and river sediment variables.

195 3.1 Flow variables and shear stress calculations

During each time step of a simulation, OverlandFlow solves the 2D flow equations across all grid links. This process determines the surface water discharge per unit width (q) and water depth (h). Subsequently, water depth at nodes is calculated based on mass conservation, factoring in all inflow and outflow at a given node. While flow velocity is not directly derived, it can be calculated at links according to $U = q/h$ with velocity components u and v for x and y directions, respectively. Our sediment transport rate calculations are based on the local shear stress considering an unsteady friction slope (Ghimire & Deng, 2011) according to:

$$S_{fx} = -\left(\frac{\partial \eta}{\partial x} + \frac{\partial h}{\partial x} + \frac{u}{g} \frac{\partial u}{\partial x} + \frac{u}{g} \frac{\partial u}{\partial t}\right) \quad \text{and} \quad S_{fy} = -\left(\frac{\partial \eta}{\partial y} + \frac{\partial h}{\partial y} + \frac{v}{g} \frac{\partial v}{\partial y} + \frac{v}{g} \frac{\partial v}{\partial t}\right) \quad \text{Eq. 2}$$

where u and v are the velocity components in the x and y directions, respectively; S_{fx} and S_{fy} are the friction slopes evaluated in the x and y directions, respectively; η is the bed surface elevation; g is the acceleration due to gravity; and t is time.

205

Each individual term in Eq. 2 is calculated directly using built-in methods of the Landlab grids. Topographic gradients ($\partial \eta / \partial x$ and $\partial \eta / \partial y$) are based on the bed elevation slope at the nodes defining a link using the `calc_grad_at_link` method.



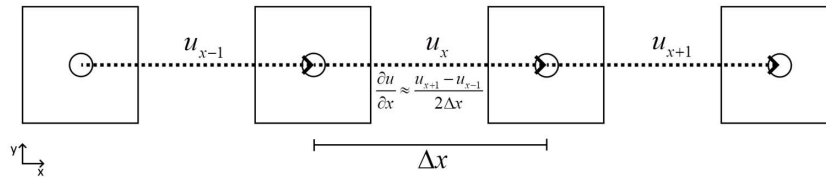
The same approach is used for water depth spatial gradients ($\partial h/\partial x$ and $\partial h/\partial y$). Velocity spatial gradients are approximated using a central difference scheme according to:

$$\frac{\partial u}{\partial x} \approx \frac{u_{x+1} - u_{x-1}}{2\Delta x} \quad \text{and} \quad \frac{\partial v}{\partial y} \approx \frac{v_{y+1} - v_{y-1}}{2\Delta y} \quad \text{Eq. 3}$$

210 where subscripts $x + 1$, $x - 1$, $y + 1$, and $y - 1$ indicate to the right, left, above, and below, respectively, the location of the link considered (Figure 3). Velocity time gradients are approximated using the backward Euler method defined as:

$$\frac{\partial u}{\partial t} \approx \frac{u^t - u^{t-1}}{\Delta t} \quad \text{and} \quad \frac{\partial v}{\partial t} \approx \frac{v^t - v^{t-1}}{\Delta t} \quad \text{Eq. 4}$$

where Δt is the time step and the superscripts t and $t - 1$ indicate the current and previous time steps.



215 Figure 3: A representation of the stencil used to calculate the velocity gradient at links. Cells are separated only to highlight the definition of velocities at links. The gradient $\partial u/\partial x$ at the location of the link with velocity u_x is estimated using a central difference scheme and considers the neighboring links. The same principle applies to calculate $\partial v/\partial y$ but is not represented in this figure.

220 The local shear stress at each link is then calculated according to:

$$\tau_x = \rho gh S_{fx} \quad \text{or} \quad \tau_y = \rho gh S_{fy} \quad \text{Eq. 5}$$

Usually, the shear stress in a river channel is defined using the hydraulic radius instead of water depth to include the effects of channel wall roughness. When modelling hydraulics, this effect is critical if the cross section of a river is well captured by a single cell because a large proportion of flow surface is in contact with the river bed and banks (i.e., bottom and sides of a

225 cross section). When a cross section is defined by more than one cell, most of cell sides are in contact with only water, therefore bank roughness becomes less important, and the hydraulic radius can be simplified as h . It's important to note that not all of the modeled domain represents river channels; a significant portion may simulate overland flow on hillslopes using the same equations. In these overland flow conditions, the concept of hydraulic radius is not applicable due to the absence of channel banks. Using hydraulic radius in such cases would not be representative of the actual flow conditions. This

230 consideration further justifies our decision to use water depth as the default option, as it provides a more versatile approach applicable to both channelized and non-channelized flows across the entire modeled landscape. Given that is impossible to



anticipate if a river section will be well represented by single or multiple cells, we set as default the option of using h in the shear stress definition. Nevertheless, the user can override this and obtain the shear stress as:

$$\tau_x = \rho g R_h S_{fx} \text{ or } \tau_y = \rho g R_h S_{fy} \quad \text{Eq. 6}$$

where $R_h = A_w/P_w$ is the hydraulics radius, $A_w = h\Delta x$ is the wetted area, and $P_w = 2h + \Delta x$ is the wetted perimeter. For north-south links we have $A_w = h\Delta y$ and $P_w = 2h + \Delta y$. To activate this option set `use_hydraulics_radius_in_shear_stress = True` when instantiating the RiverBedDynamics component.

3.3 Bed surface properties and sediment fluxes calculations

Prior to calculating sediment fluxes, RiverBedDynamics determines the bed properties required for the bed load transport equations. During instantiation, bed grain size distributions (GSD) are specified at nodes, which allows them to vary spatially. Grain sizes, defined as percentage passing, can range from fine sand to large boulders. Cohesive sediments are not supported by our component. For any sediment transport model used, it is mandatory to define the grain sizes at 0% and 100% of the distribution. This ensures uniformity in input format across different bed load transport equations. Once the GSD is specified, RiverBedDynamics will calculate the sand fraction (F_{sand}), D_{50} , the geometric mean size (D_{sg}), and the geometric standard deviation (σ_g) at each node. These last two variables are defined following the method of Parker (1990). After bed properties are defined, they are mapped into the links assuming that the connecting nodes have equal weights. The selected bed load equation defines whether these bed surface properties are updated each time step or remain constant throughout the simulation (see below).

Six different sediment transport equations are available in our component. These equations are described in detail in the original articles (Fernandez Luque & Van Beek, 1976; Huang, 2010; Meyer-Peter & Müller, 1948; Parker, 1990; Wilcock & Crowe, 2003; Wong & Parker, 2006) and have been used extensively in sediment transport studies (e.g., Barry et al., 2004; Schneider et al., 2015; Yager et al., 2007). Therefore, only the aspects related to their implementation are described here.

The first group of equations include those by Meyer-Peter & Müller (1948), Fernandez Luque & Van Beek (1976), Wong & Parker (2006), and Huang (2010). We collectively refer to these as 'Meyer-Peter & Müller style equations.' They have the form:

$$q_b^* = \alpha(\tau^* - \tau_c^*)^\beta \quad \text{Eq. 7}$$

where the coefficient α , the exponent β , and τ_c^* are parameters specific to the selected equation. These equations are only valid when $\tau^* > \tau_c^*$ or else $q_b^* = 0$. When selecting these equations, the grain size distribution of the bed remains constant during the entire simulation. The dimensionless shear stress is calculated as $\tau^* = \tau/(\rho R g D_{50})$, where $R = (\rho_s - \rho)/\rho$, and ρ_s and ρ are the densities of the sediment and water, respectively. For simplicity, the x and y subscripts from Eq. 5 are omitted in this and subsequent uses of τ .



The effect of bed slope on critical shear stress (Lamb et al., 2008; Mao et al., 2008; Mueller et al., 2005a; Smith et al., 2023; Yager et al., 2012) in relatively steep slopes (larger than 3%) can be empirically included in Meyer-Peter & Müller style
 265 equations by setting the option `variable_critical_shear_stress = True` during instantiation. When activated, Mueller et al., (2005b) equation is used to calculate τ_c^* :

$$\tau_c^* = 2.18 S_b + 0.021 \quad \text{Eq. 8}$$

where S_b is the topographic gradient defined as $\partial\eta/\partial x$ and $\partial\eta/\partial y$ for the x and y directions, respectively. While this optional setting simplifies the complex processes governing critical shear stress variation with slope, it serves as a practical tool for analyzing the model's response to τ_c^* in terrain slopes exceeding 3%. We acknowledge that this approach is a
 270 generalization and incorporating the full mechanics of these processes is outside the current scope of RiverBedDynamics.

Another option is the surface-based bed load transport equation of Parker (1990) that includes the effects of sediment mixtures in gravel-bedded rivers but does not include sand size material. In this case, if sand is present in the GSD the component will automatically remove it and renormalize the GSD curves to adjust for the change. The shear stress is here
 275 normalized using D_{sg} instead of D_{50} as follows:

$$\tau_{sg}^* = \tau / (\rho R g D_{sg}) \quad \text{Eq. 9}$$

The dimensionless measure of shear stress is:

$$\phi_{sg0} = \frac{\tau_{sg}^*}{\tau_{rsg0}^*} \quad \text{Eq. 10}$$

where $\tau_{rsg0}^* = 0.0386$ is the reference Shields stress. To account for the effects of sediment mixtures a hiding function is used:

$$\phi_i = \omega \phi_{sg0} \left(\frac{D_i}{D_{sg}} \right)^{-0.0951} \quad \text{Eq. 11}$$

where the subscript i denotes the i^{th} grain-size class. The function ω is:

$$\omega = 1 + \frac{\sigma_g}{\sigma_0(\phi_{sg0})} [\omega_0(\phi_{sg0}) - 1] \quad \text{Eq. 12}$$

280 where $\sigma_0(\phi_{sg0})$ and $\omega_0(\phi_{sg0})$ are functions that are calculated automatically within the component. The dimensionless transport rate for each i^{th} size class is defined as:

$$W_i^* = 0.00218G(\phi_i) \quad \text{Eq. 13}$$

and the function $G(\phi_i)$, the normalized dimensionless gravel bedload transport rate, is:



$$G(\phi_i) = \begin{cases} 5474 \left(1 - \frac{0.853}{\phi_i}\right)^{4.5}, & \phi_i > 1.59 \\ \exp[14.2(\phi_i - 1) - 9.28(\phi_i - 1)^2], & 1 \leq \phi_i \leq 1.59 \\ \phi_i^{14.2}, & \phi_i < 1 \end{cases} \quad \text{Eq. 14}$$

To obtain the fraction of bed load in each i^{th} size class (p_i) we used:

$$p_i = \frac{F_i G(\phi_i)}{\sum_{i=1}^N G(\phi_i)} \quad \text{Eq. 15}$$

where F_i is the volume fraction in the bed of the i^{th} grain-size class and N is the number of grain size classes with
 285 characteristic diameters D_i .

The last bed load transport equation included in our component is Wilcock & Crowe (2003). Similar to Parker (1990) this model can handle sediment mixtures. However, in this case the effects of sand content (F_{sand}) are explicitly included in the reference Shields stress, which is defined as:

$$\tau_{rs_{g0}}^* = 0.021 + 0.015 \exp(-20F_{sand}) \quad \text{Eq. 16}$$

290 The dimensionless measure of shear stress is $\phi_{sg0} = \tau_{sg}^* / \tau_{rs_{g0}}^*$ and the hiding function is expressed as:

$$\phi_i = \phi_{sg0} \left(\frac{D_i}{D_{sg}}\right)^{-b} \quad \text{Eq. 17}$$

where the exponent is:

$$b = \frac{0.67}{1 + \exp\left(1.5 - \frac{D_i}{D_{sg}}\right)} \quad \text{Eq. 18}$$

The dimensionless transport rate for each i^{th} size class (W_i^*) is:

$$W_i^* = \begin{cases} 0.002 \phi_i^{7.5}, & \phi_i < 1.35 \\ 14 \left(1 - \frac{0.894}{\phi_i^{1.5}}\right)^{4.5}, & \phi_i \geq 1.35 \end{cases} \quad \text{Eq. 19}$$

To obtain the fraction of bed load in each i^{th} size class (p_i) we used:

$$p_i = \frac{F_i W_i^*}{\sum_{i=1}^N W_i^*} \quad \text{Eq. 20}$$

295 The volumetric bed load transport rate per unit width for each grain size when using Parker (1990) or Wilcock & Crowe (2003) is calculated using:

$$q_{bi} = \frac{(\tau/\rho)^{3/2} F_i W_i^*}{Rg} \quad \text{Eq. 21}$$

Given that we are working in a 2D structured grid we can assign directionality to q_b depending on the link in which it is being calculated, $q_{b,x}$ for east–west and or $q_{b,y}$ for north-south links, by multiplying Eq. 21 by the sign of τ . The total bed load transport rate per unit width q_b is defined as the sum of the bed load transport rates of each grain size q_{bi} .



3.4 Sediment mass conservation and bed properties update

300 Once the sediment fluxes and bed load GSD at each link are calculated, it is possible to conduct a mass balance at nodes and determine changes in bed surface elevation and bed GSD. Surface bed elevation changes are calculated by the *update_bed_elevation* routine within RiverBedDynamics using the Exner equation:

$$(1 - \lambda_p) \frac{\partial \eta}{\partial t} = - \left(\frac{\partial q_{b,x}}{\partial x} + \frac{\partial q_{b,y}}{\partial y} \right) \quad \text{Eq. 22}$$

where λ_p is the bed porosity. The equation states that the change in bed elevation in time within a control volume, a cell in this case, is a function of the sediment fluxes crossing the faces of a cell (Figure 4).

305

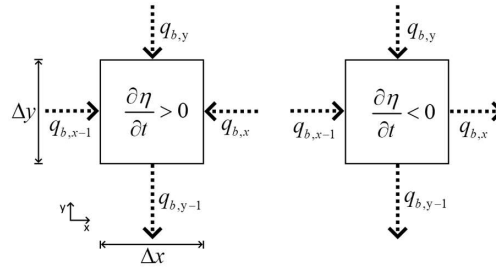


Figure 4: Examples of an increasing (left) and decreasing (right) bed surface elevation in time. Sediment fluxes across cell faces determine the net bed load transport rate within a cell and consequently dictate erosion or deposition of sediment. Sediment transport rate per unit width magnitude is represented by the length of the arrow line. In the left diagram, the sum of the three fluxes entering the cell is larger than those exiting the cells, therefore, there is a net accumulation of sediment, and the bed elevation increases. In the right diagram, fluxes in the x direction are equal in magnitude and cancel each other out, whereas in the y direction the flux leaving is larger than that entering, consequently the bed elevation will decrease.

310

We used an explicit method to approximate the solution of Eq. 22. The gradients in volumetric bed load transport rate per unit width in the x and y directions are:

$$\frac{\partial q_{b,x}}{\partial x} \approx \frac{q_{b,x} - q_{b,x-1}}{\Delta x} = \frac{\Delta q_{b,x}}{\Delta x} \quad \text{and} \quad \frac{\partial q_{b,y}}{\partial y} \approx \frac{q_{b,y} - q_{b,y-1}}{\Delta y} = \frac{\Delta q_{b,y}}{\Delta y} \quad \text{Eq. 23}$$

315 where the locations x , $x - 1$, y , and $y - 1$ are shown in Figure 4. The right-hand side of Eq. 22 can be expressed as:

$$\frac{\partial q_{b,x}}{\partial x} + \frac{\partial q_{b,y}}{\partial y} \approx \frac{\Delta q_{b,x} \Delta y + \Delta q_{b,y} \Delta x}{\Delta x \Delta y} \quad \text{Eq. 24}$$

here, $\Delta q_{b,x} \Delta y$ and $\Delta q_{b,y} \Delta x$ are the volumetric bed load transport rates in each direction, $\Delta q_{b,x} \Delta y + \Delta q_{b,y} \Delta x = \Delta Q_b$ is the net volumetric bed load transport rate, and $\Delta x \Delta y = A_{xy}$ is the area of a cell. Considering these definitions, the explicit solution to Eq. 22 is:

$$\eta^{t+1} = \eta^t - \Delta t \frac{\Delta Q_b}{(1 - \lambda_p) A_{xy}} \quad \text{Eq. 25}$$



320 Boundary conditions for updating the bed surface elevation are only required at the links located immediately upstream of
the watershed outlet. Two options can be specified, zero-gradient, which is the default option, and a fixed-value condition. In
the implementation of the zero-gradient condition within RiverBedDynamics, the model uses user-provided methods for the
watershed outlet boundary condition (e.g., *set_watershed_boundary_condition_outlet_id*,
set_watershed_boundary_condition) to identify all connecting nodes and links upstream of the outlet. Then, at the end of the
325 bed elevation update routine, RiverBedDynamics matches the values of these upstream nodes and links with the outlet
values. When other types of boundary conditions are required, such as an elevation that changes in time following a given
curve, it can be specified by setting individual nodes or links of the grid using Landlab boundary condition handling. Fixed-
value conditions can be applied, not only to the boundaries, but also to internal nodes, such that they can remain unaltered
throughout the whole simulation. This optional capability is accessed by editing the field '*bed_surf_elev_fix_node*'.

330

Sediment mass entering or leaving a cell can not only alter the bed surface elevation but also the bed GSD. We represent the
evolution of the surface and substrate GSD by means of three layers corresponding to the bed load, surface, and substrate
(Figure 5). The bed load layer is the one defined by the bed material being transported close to the river bed and is calculated
according to section 3.3. The surface layer, which is in direct contact with the flow at wet nodes, determines the bed surface
335 elevation, measured from a specific datum point (η). It contains the active layer, characterized by a thickness defined as
 $L_a = 2D_{90}$ (D_{90} is the 90th percentile of the surface GSD). The surface layer can exchange material with the bed load or
substrate layer depending if the bed aggrades or degrades, respectively. To simplify the definition of the surface layer and
facilitate the implementation of its updating algorithm, we adopted the definition of Toro-Escobar et al. (1996), which posits
that the surface layer and the active layer are of equal thickness, L_a (Figure 5). The substrate includes all the material below
340 the surface layer. Its GSD is represented using $F_{s,i}$, analogous to the way F_i defines the surface GSD.

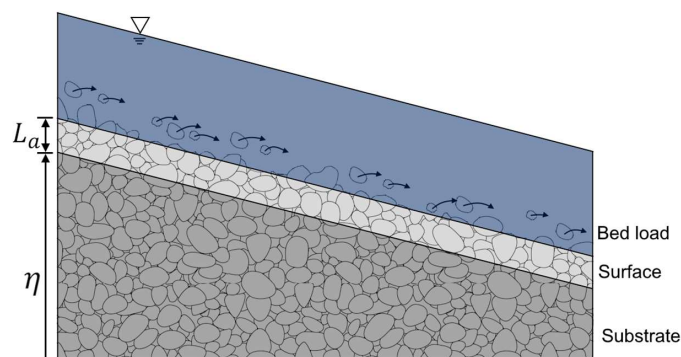


Figure 5: Schematic diagram of the model's three layers, used to represent the evolution of the surface and substrate grain size distribution.



345

To account for the dynamics of active layer grain sizes, we implemented the grain-size-specific form of the Exner equation as described by Parker (1991).

$$(1 - \lambda_p) \left(L_a \frac{\partial F_i}{\partial t} + (F_i - f_{ii}) \frac{\partial L_a}{\partial t} \right) = - \frac{\partial(q_b p_i)}{\partial x} + f_{ii} \frac{\partial q_b}{\partial x} - \frac{\partial(q_b p_i)}{\partial y} + f_{ii} \frac{\partial q_b}{\partial y} \quad \text{Eq. 26}$$

where f_{ii} accounts for the interchange of sediment between the active layer and the substrate interface. This corresponds to the fraction of material in the i^{th} grain size exchanged between these two layers. In our model we used the transfer function

350 of Toro-Escobar et al. (1996):

$$f_{ii} = \begin{cases} F_{s,i}, & \partial\eta/\partial t < 0 \\ 0.7p_i + 0.3F_i, & \partial\eta/\partial t > 0 \end{cases} \quad \text{Eq. 27}$$

This equation states that when the bed degrades the active layer GSD is that of the substrate (Figure 6c). Conversely, when the bed aggrades, a mixture of surface and bedload material transfers to the substrate, thereby creating stratigraphy.

We solved Eq. 26 explicitly approximating the derivatives as:

$$\frac{\partial q_b}{\partial x} \approx \alpha \frac{q_{b,x} - q_{b,x-1}}{\Delta x} + (1 - \alpha) \frac{q_{b,x+1} - q_{b,x}}{\Delta x} \quad \text{Eq. 28}$$

$$\frac{\partial(q_b p_i)}{\partial x} \approx \alpha \frac{q_{b,x} p_{i,x} - q_{b,x-1} p_{i,x-1}}{\Delta x} + (1 - \alpha) \frac{q_{b,x+1} p_{i,x+1} - q_{b,x} p_{i,x}}{\Delta x} \quad \text{Eq. 29}$$

355 the y direction has an equivalent discretization (just replacing x by y). The coefficient α is used to switch from an upwind to central difference scheme. For stability purposes we opted for a default value of 1. The explicit solution to Eq. 26 is:

$$F_i^{t+1} = F_i^t - \frac{1}{L_a} (F_i - f_{ii})(L_a^t - L_a^{t-1}) + \frac{\Delta t}{L_a(1 - \lambda_p)} \left(- \frac{\partial(q_b p_i)}{\partial x} + f_{ii} \frac{\partial q_b}{\partial x} - \frac{\partial(q_b p_i)}{\partial y} + f_{ii} \frac{\partial q_b}{\partial y} \right) \quad \text{Eq. 30}$$

For simplicity we dropped some t superscripts, but all variables are evaluated at the current time step except for L_a^{t-1} .

360 Given that our model can predict temporal changes in bed surface elevation and GSD we implemented stratigraphy tracking capabilities, thus allowing a better representation of processes that are not purely erosional or depositional. Our model stores the current and past GSD and elevation of the surface and substrate across the whole watershed (Figure 6). At the beginning of a simulation the surface and substrate have, by default, the same GSD (Figure 6a). When deposition occurs at a given location, RiverBedDynamics stores the elevation and GSD of the deposited sediment; this data is recorded at regular intervals determined by the variable *num_cycles_to_process_strat*. Once the accumulated deposited material at that location reaches the user-specified vertical thickness (*bed_surf_new_layer_thick*, default value of 1 m, L_s in Figure 6b) it is logged as a new layer of stratigraphy. The recorded GSD for this layer is a time-averaged value derived from all the sediment deposited over the last *bed_surf_new_layer_thick* meters, after which the process begins anew. In scenarios where a new stratigraphic layer is being eroded, the model reads the stored data and adjusts F_s and $F_{s,i}$ based on the elevation of the layer



being scoured (Figure 6d). When the bed surface is eroded below the initial bed surface elevation, the F_s retains its original
 370 state from the beginning of the simulation (Figure 6c).

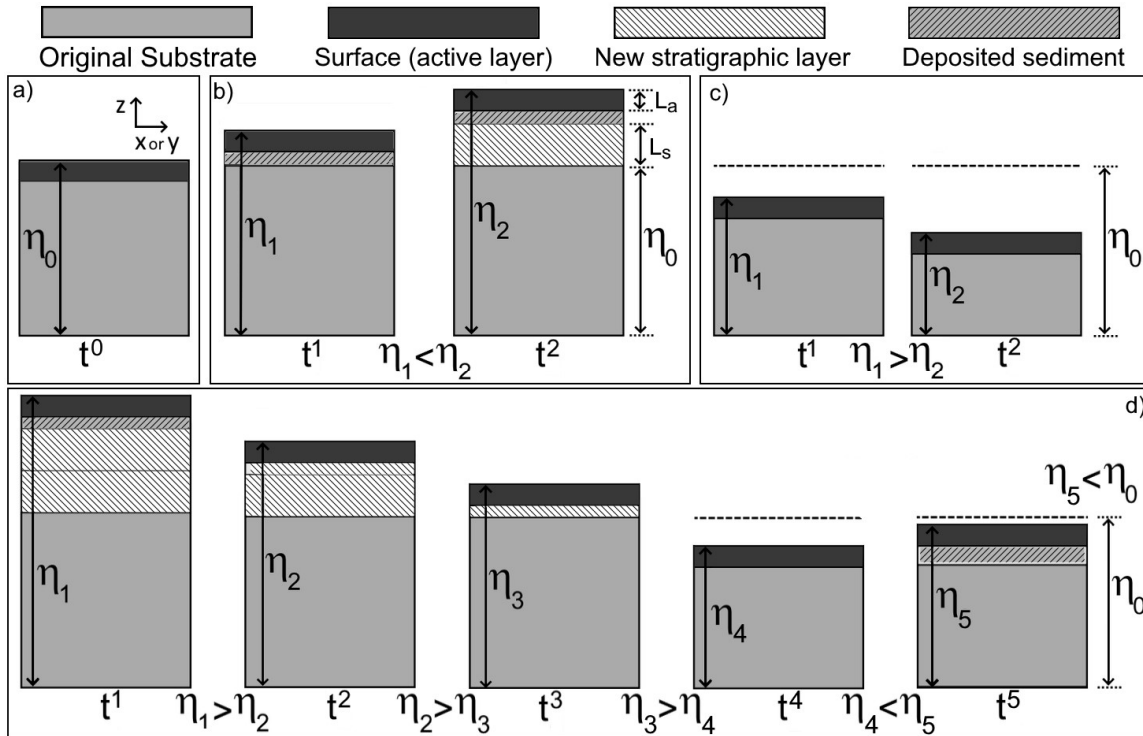


Figure 6: Graphical description of the model's algorithm for updating bed surface and substrate GSD. a) Initial bed surface
 elevation (η_0) at the start of the simulation (t_0). b) A pure depositional process. The bed surface elevation monotonically
 375 increases, new stratigraphic layers form once the deposited layer's thickness reaches L_s . The GSD of the deposited sediment
 is calculated using Eq. 26. c) A pure erosional case. Bed surface elevation monotonically decreases, the surface and substrate
 have the GSD specified at t_0 . d) An alternating erosion/deposition case. The bed is first eroded below the initial elevation (at
 t_4), following the erosion of newly deposited layers (at t_1 and t_2), and then experiences deposition again (at t_5). The local
 minimum bed surface elevation is updated to η_4 and the GSD at η_5 is calculated using Eq. 26.

380 4 Running a model using the Landlab framework

Some general characteristics of the Landlab modeling approach were described in Section 2.0. Therefore, in this section we
 focus only on describing specific details of the variables, default configurations, unit system, and capabilities of our model.
 The component was designed to work exclusively using the International System of Units (SI). If imperial units are required



they must be converted into SI before using them as input. Gravitational acceleration is constant and equal to 9.80665 m/s^2 .

385 During the instantiation of RiverBedDynamics the user can modify and/or define the variables or options listed in Table 1.

Table 1: List variables and options used in RiverBedDynamics during instantiation

Variable or option	Default value	Units	Comment
ρ_s	2,650	kg/m ³	
ρ	1,000	kg/m ³	
<i>bedload_equation</i>	'MPM'	-	See section 3.3
λ_p	0.35	-	
Δt	1	s	
<i>current_t</i>	0	s	Used in case a simulation does not start at time 0 s
α	1	-	
<i>outlet_boundary_condition</i>	<i>zeroGradient</i>	-	Also available: <i>fixedValue</i>
<i>surface_water_velocity_prev_time_link</i>	Same as current time	m/s	Forces gradients in Eq. 4 to be zero.
<i>variable_critical_shear_stress</i>	<i>False</i>	-	
<i>use_hydraulics_radius_in_shear_stress</i>	<i>False</i>	-	
<i>track_stratigraphy</i>	<i>False</i>	-	
<i>num_cycles_to_process_stra</i>	10	-	
<i>bed_surf_new_layer_thick</i>	1	m	

390 When using our component, like all other Landlab simulations, a driver file is required. This file is a procedure script
 containing a set of instructions to import libraries, instantiate classes, load data, run and loop through time, and finalize a
 simulation. Once the elements have been initialized and are ready to loop in time, the two different basic routines that define
 our LEM are executed sequentially, first OverlandFlow then RiverBedDynamics (Figure 2). At every iteration within the
 time loop, OverlandFlow is executed and returns updated flow conditions (e.g., q and h) across the domain and the Δt
 395 required to predict changes in bed surface elevation and GSD (Eq. 25 and Eq. 26). Then, the first part of the
 RiverBedDynamics routine calculates and stores a series of hydraulics and sediment transport variables. When selecting a
 bed load equation the following terminology is used: *MPM* for Meyer-Peter & Müller (1948), *FLvB* for Fernandez Luque &
 Van Beek, (1976), *Parker1990* for Parker (1990), *WilcockAndCrowe* for Wilcock & Crowe (2003), *Huang* for Huang
 (2010), or *WongAndParker* for Wong & Parker (2006). The default option is *MPM*. After all calculations are completed the
 400 second part of the RiverBedDynamics routine starts and uses the calculated bed load transport rates per unit width and bed
 load GSD to modify the bed elevation and GSD according to the equations described in Section 3.4.



The results of the calculations are stored as fields in the grid, but only the current time step is available for reading/writing, except for the velocity at the previous time step and stratigraphy properties. Therefore, when analyzing the changes of a given variable in time, the variable must be stored in a local file in a user-defined format that is specified in the driver file. The format in which RiverBedDynamics stores bed load, surface, and substrate GSD results may be difficult to interpret because it was designed to be easily accessible by the component and not for user-readability. However, a postprocessing function called `format_gsd` is implemented and returns a panda DataFrame that contains the GSD for each node or link, depending on the input, in a user-friendly format.

410 4 Evaluation of RiverBedDynamics

4.1 Equilibrium bed surface slope in uniform flow conditions

To test the ability of our component for predicting changes in the bed surface elevation, we obtained an analytical solution for an idealized channel with uniform flow conditions. In this case, a given bed load transport rate is imposed at the upstream boundary such that the bed surface slope must adjust until the channel reaches a stable condition. We combined Manning's equation to include uniform flow conditions and Eq. 7 to estimate bed load transport rate within the channel. By expanding Eq. 7 we can solve for the bed slope required to transport an imposed bed load rate (q_b in this case):

$$S = \frac{RD_{50}}{h} \left(\left(\frac{q_b}{\alpha \sqrt{RgD_{50}D_{50}}} \right)^{1/\beta} + \tau_c^* \right) \quad \text{Eq. 31}$$

The equilibrium slope (S) is a function of h which in turn depends on the flow discharge (Q) and channel properties, in this case n and channel width (b). Once the equilibrium state has been reached h can be estimated using:

$$h = \left(\left(\frac{Qn}{b} \right) \left(\frac{1}{\sqrt{S}} \right) \right)^{3/5} \quad \text{Eq. 32}$$

which is a form of Manning's flow equation considering a rectangular channel and shallow flows such that $b \gg h$, therefore, $R_h \approx h$. Combining Eq. 31 and Eq. 32 a solution for S can be found.

$$S = \left(\frac{RD_{50}}{\left(\frac{Qn}{b} \right)^{3/5} \left(\left(\frac{q_b}{\alpha \sqrt{RgD_{50}D_{50}}} \right)^{1/\beta} + \tau_c^* \right)} \right)^{10/7} \quad \text{Eq. 33}$$

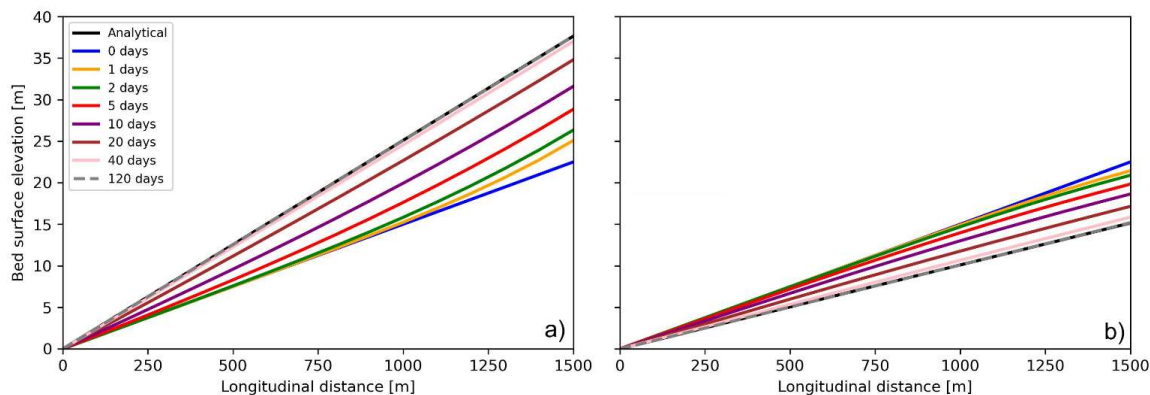
Note that Eq. 33 is valid only under uniform flow conditions and may perform poorly at intermediate bed states (i.e., when the bed is adjusting) because the flow is not uniform locally. This form of the analytical solution is convenient when testing our component because, in terms of hydraulic variables, it only depends on Q , which can be specified as a boundary condition or by using a rainfall intensity that generates the target Q .

425



We conducted two tests to evaluate the response of our component. Both cases started with the same initial bed configuration but differed in the imposed upstream sediment supply rate. In general terms, they consisted of a 1500 m long, straight channel with an initial bed surface slope of 0.015 m/m, a fixed elevation at the outlet at 0 m, and a surface roughness of $n = 0.03874$. Flow discharge was constant $Q = 100 \text{ m}^3/\text{s}$ and was specified by using a rainfall intensity of 0.01 m/s acting over a single cell of 100 m per side ($\Delta x = \Delta y$) located at the upstream boundary. This case represents essentially a one-dimensional flow scenario in which the grid is composed of uniformly sized cells, and the channel has a width of one cell. The digital elevation model (DEM) employed was 19 rows by 3 columns, with 2 of the columns serving as boundaries. The bed surface GSD was uniform with a grain size of 50 mm and the bed load transport equation was that of Meyer-Peter & Müller (1948). In OverlandFlow we specified h_{init} , the initial water depth in all cells, as 1 mm, the time step was limited to a maximum of 5 s, and all other variables were left as their default value. The time limitation was imposed due to the rapid changes in bed elevation; while OverlandFlow alone could accommodate larger time steps based on the water flow Courant number, the coupled system required smaller time steps to maintain stability and prevent simulation crashes. The modeled scenarios were a purely aggradation case in which $q_b = 0.0087 \text{ m}^2/\text{s}$ and a purely degradation case where $q_b = 0.0012 \text{ m}^2/\text{s}$. We ran each case for 120 days of constant, steady flow, and compared the predicted and analytical bed slopes at the end of the simulation. We chose 120 days because at this time, the rate at which the bed elevations were changing were relatively small, $9 \cdot 10^{-5}$ and $-4 \cdot 10^{-4} \text{ m/day}$ in the aggradation and degradation cases, respectively. We considered these rates small enough to be representative of an equilibrium condition.

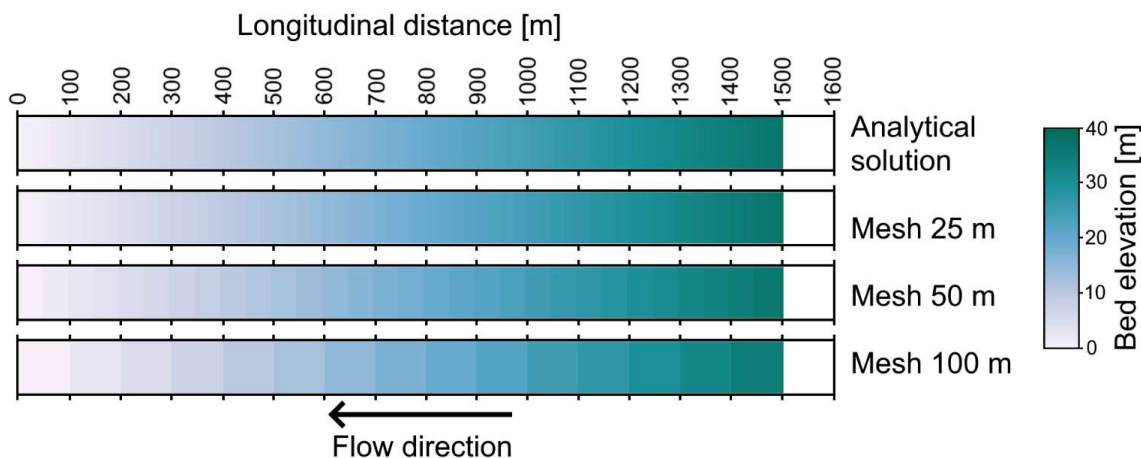
In the aggradation case our LEM predicted an S equal to 0.0251 whereas the analytical solution of Eq. 33 was 0.025 (percentage error of 0.32 %). The degradation case had an S of 0.0101 and 0.010 by the LEM and analytical solution, respectively (1 % error). Locally, the major differences between the LEM-predicted and analytically solved bed elevations were in the upstream region, near where the sediment supply was imposed. During the final time step, the maximum local percent difference in volume of deposited sediment was 0.09% (corresponding to an elevation change of 0.014 m) in the aggradation scenario and 0.68% (corresponding to an elevation change of 0.05 m) in the degradation scenario (Figure 7). The small percentage error and the general trend of the local surface elevation with streamwise distance (Figure 7) suggests that our component can accurately predict changes in bed elevation.



455 Figure 7: Changes in bed surface elevation for a case of a) pure aggradation and b) pure degradation. The analytical solution corresponds to the equilibrium slope given by Eq. 33. The small differences in bed elevation after 40 and 120 days indicate that the systems achieved an equilibrium state.

We analyzed the sensitivity of our results to the mesh size in the pure aggradation case by comparing the bed elevation after 120 days of simulation using meshes with half (50 m) and a quarter (25 m) of the size of the beforementioned case (Figure 8). By the end of each run, the average slope was 0.0251 in all cases. The percentage error in slope compared to the analytical solution was 0.29, 0.31, and 0.32 % for the 25, 50, and 100 m grid resolution, respectively. Other than the mesh size, the configuration was identical in all simulations except for the maximum time step, which was 5 s for the 100 and 50 m cases and 2.5 s for the 25 m run.

460



465 Figure 8: Sensitivity of the predicted bed elevations to the grid resolution after 120 days of simulation. Three different meshes were used and compared to the analytical solution.



4.3 Comparing bed load transport models predictions

We checked the predictions of bed surface elevation and local GSD for all bed load transport models included in our component using a test similar to the one described in the previous section. In this case, we used a 1500 m long, straight channel with an initial bed surface slope of 0.015 m/m, a fixed elevation of 0 m at the channel outlet, a surface roughness of $n = 0.0275$, and a flow discharge $Q = 100 \text{ m}^3/\text{s}$ that was generated by a rainfall intensity of 0.02 m/s acting over two cells of 50 m per side ($\Delta x = \Delta y$) located at the upstream boundary. Similar to the previous case, this configuration models a one-dimensional flow setup, utilizing a grid of uniformly sized cells with a channel width of two cells. The initial bed surface GSD had a D_{50} of 32 mm and D_{sg} of 28.84 mm including grains ranging from 2 to 256 mm (Figure 9 d). The initial water depth (h_{init}) at all cells was 1 mm, and the time step was fixed and equal to 5 s. Similar to the previous test case, this time step constraint was necessary to account for the rapid bed elevation changes. Although OverlandFlow independently allows for larger time steps based on the Courant number, the coupled model demanded shorter intervals to ensure numerical stability. All other variables had their default value. The upstream sediment supply was $q_b = 0.0075 \text{ m}^2/\text{s}$ with the same GSD as the initial bed surface. The total simulation time was 120 days for all the models we ran. We choose this test configuration because our LEM predictions using the bed load equations of Parker (1990) and Wilcock & Crowe (2003) can be verified using the algorithm developed and implemented by Parker (2004) RTE-bookAgDegNormGravMixPW.xls (Sediment transport morphodynamics with applications to rivers and turbidity currents, http://hydrolab.illinois.edu/people/parkerg/morphodynamics_e-book.htm). For the Parker (1990) equations, we considered two scenarios: one where the GSD remains constant, and another where the surface and substrate GSDs are updated in line with Eq. 30. These scenarios are referred to as 'Parker' and 'Parker stratigraphy update' in Figure 9. Additionally, an analytical solution for the equations of Meyer-Peter & Müller (1948), Fernandez Luque & Van Beek (1976), Wong & Parker (2006), and Huang (2010) is available using Eq. 33.

We compared the predicted channel longitudinal profiles between all bed load transport models at different simulation times (Figure 9 a and b). After 10 days, the equations of Parker (1990) and Wilcock & Crowe (2003) predicted a more concave-upward- longitudinal profile and a higher elevation at the upstream boundary compared to the models that do not account for the whole GSD (Figure 9 a). The models of Meyer-Peter & Müller style equations had a more uniform slope along the channel profile. Comparing our LEM predictions with those from the RTE-bookAgDegNormGravMixPW.xls (hereinafter, Parker-ebook), we observed good agreement in the predicted bed elevations along the channel. For Parker (1990) and Wilcock & Crowe (2003), the average errors in elevation were around 0.1%, with a maximum local difference in bed elevation of less than 1.5%. This corresponds to an elevation difference of 0.565 m at the upstream boundary for the Parker (1990) model that uses stratigraphy update. There was no analytical solution after 10 days for any model because the equilibrium condition had not been reached yet.



After 120 days, all models were considered to be in relatively stable conditions, as indicated by the rate of elevation change at the upstream boundary node. The maximum elevation change was 22 mm/day for Wilcock & Crowe (2003), followed by 14 mm/day for Wong & Parker (2006), and less than 10 mm/day for all other models. Considering the increase of 52 m over this period at the upstream end of the model for Wilcock & Crowe (2003), the rate of 22 mm/day can be seen as relatively minor. Although the longitudinal profiles from all models showed a relatively uniform slope (Figure 9 b), local elevations varied. For instance, Wilcock & Crowe, (2003) predicted a final bed slope of 0.0495 m/m, almost twice as steep as the slopes predicted by Meyer-Peter & Müller (1948) at 0.0249 m/m and Fernandez Luque & Van Beek (1976) at 0.0311 m/m. Parker (1990), with stratigraphy updates, predicted an average bed slope of 0.0408 m/m. Similar to the observations after 10 days, the elevation predictions of RiverBedDynamics aligned well with those in Parker-ebook and the analytical solutions. In terms of average percentage error, all predictions were below 1.4%, with a maximum local difference in bed elevation of less than 1.1%. This discrepancy corresponds to an elevation difference of approximately 15 cm. The elevation predicted by the Meyer-Peter & Müller style equations in the LEM closely matched those calculated using the equilibrium slope for the same equations (Eq. 33), with errors below 0.5%.

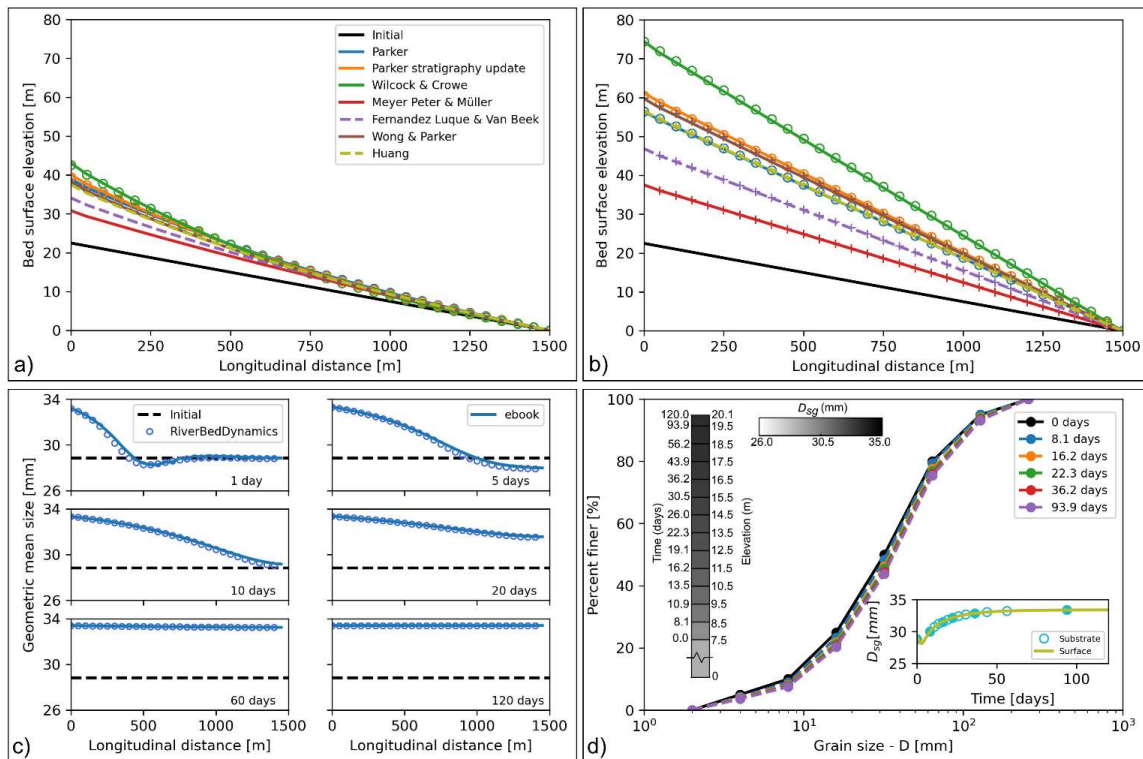
Based on the results of Parker (1990) with stratigraphy updates, we analyzed the local evolution of the surface D_{sg} at different times during the simulation. Initially, the bed at the most upstream node quickly adjusted (Figure 9 c, 1 day panel) with D_{sg} increasing from 28.84 to 33.19 mm. This value remained almost constant until the end of the simulation, with a final D_{sg} of 33.45 mm. In the first 9 days of simulation, the bed also experienced locations of fining, indicated by local D_{sg} values lower than 28.84 mm. However, after 10 days and until the end of the simulation, the bed consistently had a D_{sg} larger than 28.84 mm across all locations. On day 60, D_{sg} was nearly uniform throughout the domain, with 33.4 mm at $x = 0$ and average of 33.3 mm).

To verify the accuracy of our D_{sg} predictions, we compared them with those of the Parker-ebook. Despite small local differences in D_{sg} (maximum of 1.04 mm on day 5), the magnitudes and spatial distribution matched reasonably well (Figure 9 c). The observed differences, though minor, can be attributed to the way flow is calculated. In our LEM, we used the results of OverlandFlow, a 2D flow solver, that accounts for flow unsteadiness while in the Parker-ebook the flow is predicted using simplified relations for hydraulic resistance and the normal flow (local equilibrium) approximation. It's important to note that our goal was not to replicate the Parker-ebook results exactly but to have an approximate comparison to generally validate our findings.

Using this same example, we further explored the stratigraphy tracking capabilities of RiverBedDynamics, focusing on the comparison of surface and substrate GSD, particularly D_{sg} . For simplicity, we selected a single location at $x = 1000$ m and analyzed it through time, thereby not investigating spatial GSD changes in this analysis. In our graphical comparison (Figure



9 d), only the topmost layer of the substrate was considered. With the default setting of *bed_surf_new_layer_thick* at 1 m, the first new layer was created after 8.1 days. This layer had a GSD that was, on average, coarser than the initial GSD (Figure 9 d). Throughout the 120-day simulation, a total of 12 layers were created, with the final one added after 93.9 days. Notably, 535 ten layers were added before 50 days, and seven of these within the first 30 days. This pattern indicates that most substrate GSD updates occurred during the first quarter of the simulation, a period when bed conditions differed significantly from those observed at equilibrium (Figure 9 d subplot).



540 Figure 9: Evolution of bed surface elevation and local grain size distribution (GSD) for all bed load transport models
 implemented in RiverBedDynamics. a) Predicted longitudinal bed surface profiles after 10 days and b) after 120 days of
 simulation. Continuous and dashed lines represent RiverBedDynamics results, circles indicate solutions from the "ebook," and
 crosses denote analytical solutions. c) Changes in space and time of the bed surface geometric mean grain size D_{sg} . Initial
 values are repeated in each panel for reference. The "ebook" line represents the Parker-ebook solution, which is based on
 545 Parker (1990) equations. d) Substrate GSD evolution at $x = 1000$ m for various simulation times. Left of the GSD curve:
 stratigraphic profile showing layer formation times and elevations. Top values indicate final surface elevation (20.1 m) and
 simulation end time (120 days). Right subplot: Surface and substrate D_{sg} temporal changes. The substrate in this context is the
 layer right below the surface as defined in Figure 5 and Toro-Escobar et al. (1996). Circle markers denote substrate GSD
 update times, with filled circles corresponding to times shown in the GSD curves.



550

4.4 Application to a large watershed – Effect of rainfall intensity on morphological changes

Our previous bed evolution tests predominantly focused on flow in a single channel and were restricted to pure erosion or deposition. To expand on this, we conducted a final test of our LEM in a more complex and larger watershed to analyze how flow discharge and bed surface elevation vary at different locations within the domain under different rainfall events. We used a synthetic square watershed similar to that of Adams et al. (2017), covering an area of 36 km² with a resolution of 30 x 30 m per cell. The watershed elevations ranged from 0 m at the basin outlet to 25 m at the highest point (Figure 10 a).

We considered two cases of temporal distributions of rainfall intensities, both having uniform rainfall and the same total volume of water precipitated (24 mm) over all cells. We refer to these cases as i) Steady, where the rainfall intensity was 10 mm/hr lasting for two hours and 24 minutes (8640 s), and ii) Intermittent, where rainfall consisted of four cycles alternating between 60 mm/hr and 0 mm/hr, with each of the two rainfall rates within a cycle lasting for 360 s (Figure 10 b). We quantified changes in flow discharge and bed surface elevation at three locations: Site 1 located at the watershed outlet, Site 2 located upstream of the outlet and at the confluence of the most downstream tributaries, and Site 3 located approximately at the center of the watershed (Figure 10 a).

565

We ran each model for 24 hours, setting Manning's n uniformly across the watershed with a value of 0.025, and using the bed load transport equation of Meyer-Peter & Müller (1948) with a D_{50} of 4 mm. All other variables during the instantiation of the components had default values. We simulated each rainfall case with and without activating RiverBedDynamics (4 cases in total) to analyze the effect of the selected temporal distribution of rainfall intensity on flow hydraulics (e.g., flow discharge) independent of morphodynamic changes that would also influence the hydraulics (without RiverBedDynamics) and to include the feedbacks between hydraulics and morphological changes (with RiverBedDynamics).

When running only OverlandFlow (i.e., RiverBedDynamics deactivated), the resulting hydrographs for both the steady and intermittent cases had relatively smooth shapes at the three sites (Figure 10b). Compared to the steady case, the intermittent scenario showed earlier and larger peak discharges at every site. For example, at Site 1 under steady conditions, the peak was 54.2 m³/s arriving after 3.6 hours compared to 57.9 m³/s at 2.6 hours under intermittent rainfall.

With RiverBedDynamics activated, the resulting hydrograph had a lower peak discharge compared to hydrographs run using fixed bed elevations. At the outlet, for the steady rainfall condition, the peak discharge was 40.7 m³/s, a reduction of almost 25% compared to the case without bed evolution. For the intermittent case, the peak discharge was 42.1 m³/s, a reduction of almost 27% compared to the case without bed evolution (Figure 10 b). At Site 2, the reductions in peak discharge for the steady and intermittent cases were about 19% when we included effects of bed evolution. At Site 3, the changes in

580



hydrograph shape caused by including bed evolution were relatively small, with the discharge peak decreasing by less than 5% in both steady and intermittent scenarios.

585

Comparing hydrograph shapes, Sites 2 and 3 had a smooth shape, slightly skewed to the left, and with a single peak for both the steady and intermittent cases. Site 1, at the outlet, had similar characteristics to Sites 2 and 3 for the fixed bed elevation case. However, when the bed elevation varied in time, the shape of the hydrograph at Site 1 changed, featuring a double-peak.

590

RiverBedDynamics predicted alternating periods of erosion and deposition at Sites 1 and 2 for both rainfall cases. In the steady scenario, Site 1 initially eroded to -0.544 m from 0.023 m, before depositing to 0.256 m. Site 2 first deposited sediment, increasing elevation by 0.874 m, then eroded to 0.566 m. The intermittent case showed similar patterns, with Site 1 ranging between -0.526 m and 0.262 m, and Site 2 peaking at 0.922 m before reducing to 0.578 m. Site 3 showed

595

negligible changes in both scenarios.

Most bed elevation changes occurred during the first 6 hours of simulation, coinciding with larger discharges and shear stresses. Throughout the watershed, scour and deposition patterns were observed primarily near confluences or areas with changes in local channel direction. The total area experiencing erosion or deposition larger than 1 cm after 24 hours was

600 5100 and 8730 m^2 for the steady and intermittent scenarios, respectively.

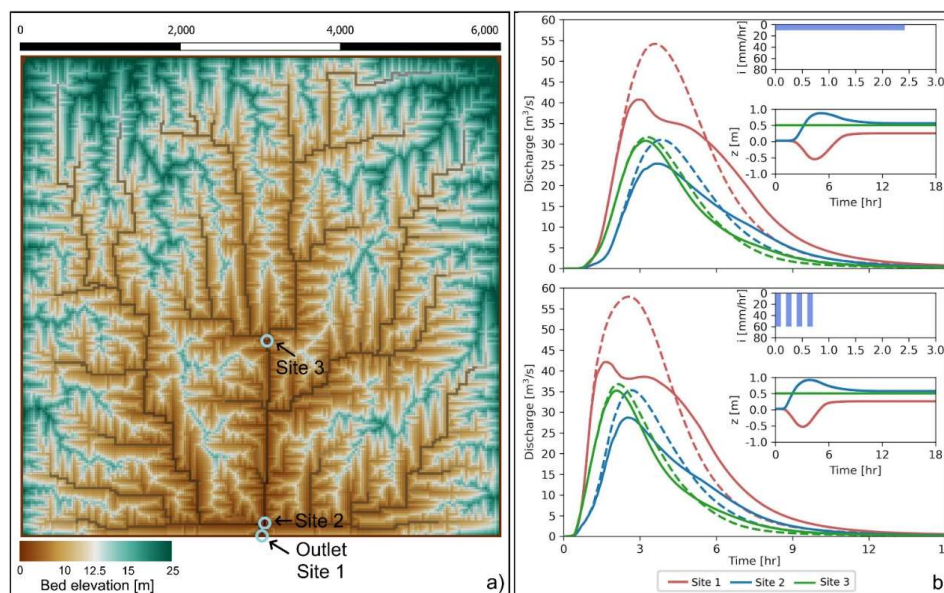




Figure 10: Discharge and bed surface elevation response to different rainfall intermittency scenarios. a) Synthetic square test basin. Three different sites, called Site 1, 2, and 3, were chosen to represent some of the spatial variability within the watershed. 605 b) Hydrographs for steady (upper panel) and intermittent (lower panel) rainfall cases. Inset panels show hyetographs (top) and bed surface elevation (bottom) at the three sites. Dashed lines represent simulations with only OverlandFlow, while solid lines include the effects of bed elevation changes predicted by RiverBedDynamics.

5 Discussion

The results presented in Section 4 demonstrate that RiverBedDynamics can be effectively coupled with a surface hydraulics 610 flow solver, specifically OverlandFlow in this study, to predict the evolution of bed surface properties at a watershed scale. This initial version allows us to simulate bed changes with varying degrees of complexity. For example, when utilizing the bed load transport equation of Meyer-Peter & Müller (1948), only changes in bed elevation can be considered. However, by selecting the equations of Parker (1990) or Wilcock & Crowe (2003), we can track changes in surface and substrate bed elevations as well as GSD over time. Thus, our LEM enables users to include or exclude certain processes and details 615 depending on their specific prediction requirements.

While our new component was developed using OverlandFlow, it can integrate with any flow solver available in Landlab due to the standardized component structure. For example, in very large watersheds, where local details are less crucial than regional changes, the KinwaveOverlandFlowModel could be used to reduce simulation time. Conversely, if small-scale 620 information is required, the OverlandFlow version of Adams et al. (2017) may suffice. It's worth noting that some assumptions that are included in the flow solver of Adams et al. (2017), such as negligible contributions from the advection term of the shallow water equations (Bates et al., 2010; de Almeida et al., 2012), may not be representative in a complex fluvial system. Consequently, a more comprehensive flow model may need to be developed to account for such processes. Regardless, the structure of RiverBedDynamics and other Landlab components facilitates easy model integration.

625 We acknowledge that our model incorporates several simplifying assumptions that could potentially impact simulation accuracy in certain scenarios. First, our solution to the Exner equation, which predicts changes in bed surface elevation, is one of the simplest formulations when working in a 2D approach (Furbish, Fathel, & Schmeeckle, 2017). While more generalized forms of sediment mass balances have been developed and applied (e.g., Juez et al., 2016; Paola & Voller, 2005; 630 Parker et al., 2000), we prioritized a computationally efficient implementation suitable for large watershed applications. In our formulation, we assumed that rectangular elements could define the alignment of a channel as well as general flow directions on hillslopes. However, this may not be representative of channels with significant curvature. Incorporating a curvature coefficient similar to that implemented by Van De Wiel et al. (2007) could lead to more accurate results, especially near river confluences.

635



Second, all our test cases involved channels without macro-roughness elements such as large boulders, vegetation, or any type of flow obstructions that can significantly alter the flow direction. Although we aimed to make RiverBedDynamics as versatile as possible, we have not yet evaluated its performance when subjected to sharp local gradients in shear stress induced by obstacles.

640

Third, we implemented an optional slope-dependent critical shear stress equation (Mueller et al., 2005a), which can be used in the models of Meyer-Peter & Müller (1948) and Fernandez Luque & Van Beek, (1976). We recommend caution when using this option as it both overrides the original τ_c^* values and allows τ_c^* to vary in time as local bed slope change, which may lead to unexpected behavior. This capability was included based on preliminary model simulations where locations with steep elevation gradients, particularly riverbanks, eroded at a faster pace than expected, resulting in artificial channel widening.

645

Furthermore, certain sediment transport phenomena are not included in this first release. For example, RiverBedDynamics does not account for suspended sediment motion or its effects on bed evolution. Additionally, sharp unnatural angles within the river bed can occur because the effects of the angle of repose (sometimes called avalanche or sediment slide models) were not included in our results (Sanchez & Wu, 2011; Song et al., 2020). Finally, we did not incorporate the effects of sediment or particle diffusion (Furbish, Fathel, Schmeeckle, et al., 2017) that may smooth the bed profile, resulting in a more realistic representation (compared to having large bed angles).

650

RiverBedDynamics is unique among Landlab components in its ability to predict sediment deposition using a fractional grain size formulation, making it particularly suited for modeling gravel bed rivers. Other components primarily focus on predicting bed surface elevation changes based on transport-limited or detachment-limited river assumptions. However, this advanced capability comes at a computational cost: simulations using RiverBedDynamics can take up to 1.5 times longer compared to the DetachmentLtdErosion component (even when using the simplest configuration such as MPM, no stratigraphy tracking, and constant GSD). This increased runtime may constrain the total possible simulation time. Although there are no intrinsic limitations on simulation time in RiverBedDynamics, this mechanistic approach may be better suited for modeling relatively small-time scale processes.

655

660

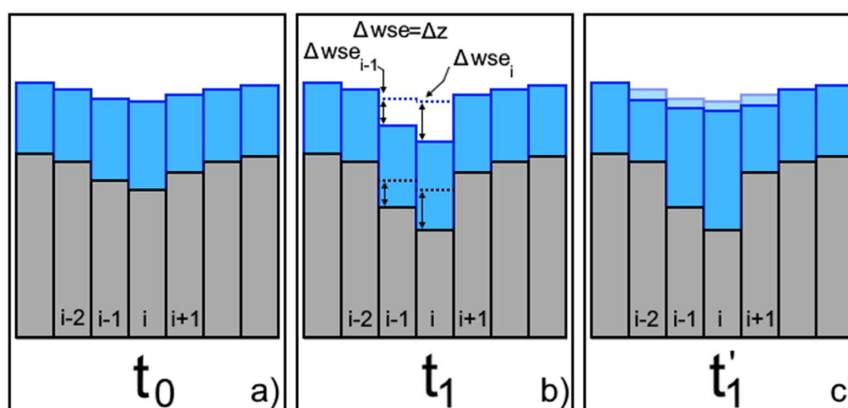
The coupled OverlandFlow-RiverBedDynamics approach in our LEM employs a decoupled method to solve for river bed evolution (Cao et al., 2002; Colombini & Stocchino, 2005). This means that the governing equations are solved separately and serially. Essentially, the flow is "paused" while the RiverBedDynamics component solves the Exner equation during a given time step (Figure 11 a and b). Consequently, the selected time-step must ensure relatively small bed elevation changes to maintain simulation stability. For scenarios involving flow, rainfall, and watershed conditions that generate dramatic

665



elevation changes, an optional local correction can be used to preserve numerical stability and ensure mass conservation
 670 (Figure 11 c).

Since only the water surface elevation is affected after a change in bed surface elevation, not water depth or discharge, it is possible to locally adjust the water depth to account for bed elevation effects. This is done by running OverlandFlow for a few internal cycles to obtain a better water surface elevation around the nodes with bed surface elevation changes (only those
 675 sharing links are corrected). Once the internal cycle finishes, the corrected water depth is mapped onto the grid. In this correction process, flow discharge remains unaltered, but the local velocity is modified. This capability is configured in the driver file, and an example of its use is provided in the example code for the test case used in section 5.4.



680 Figure 11: Illustration of the local water depth correction, an optional capability in RiverBedDynamics. This example demonstrates an erosional case, though the correction also applies to deposition. a) Initial bed and water depth condition (t_0). Grey rectangles represent individual bed nodes, identified by subscript i . Light blue rectangles depict water depth. b) Bed and water surface elevations change at nodes $i - 1$ and i by the end of time t_1 , while water depth remains constant. The bed erosion depth at a specific node equals the decrease in water surface elevation, $\Delta wse_{i-1} = \Delta z_{i-1}$ and $\Delta wse_i = \Delta z_i$, where Δwse and
 685 Δz represent the change in water surface and bed elevation, respectively. Dashed black and blue lines show the change in water surface and bed elevation at t_0 . c) The local correction is applied to water surface elevation for all nodes sharing a link with those that experienced elevation changes (from $i - 2$ to $i + 1$). The time t'_1 denotes an internal cycle; simulation time does not progress during this correction. Transparent light blue areas above the water surface elevation indicate its position at time t_0 .

6 Current capabilities and future enhancements

690 RiverBedDynamics represents a significant advancement in Landlab's modeling capabilities for river systems. As the first component to predict sediment deposition using a fractional grain size formulation, it is particularly suited for modeling gravel bed rivers. Unlike other components that focus primarily on bed surface elevation changes based on transport-limited or detachment-limited assumptions, RiverBedDynamics enables more complex simulations of bed evolution. It tracks



changes in both surface and substrate bed elevations as well as grain size distribution over time, allowing users to model
695 detailed and realistic scenarios of river bed dynamics at watershed scales.

The examples included with RiverBedDynamics demonstrate its versatility, yet they represent only a subset of the diverse
scenarios that can be simulated within the Landlab environment. For instance, integrating RiverBedDynamics with other
components like VegCA opens up new avenues for studying vegetation competition under non-steady sediment transport
700 regimes. This integration capability highlights the component's potential for multidisciplinary research in fluvial
geomorphology and ecology. The model could also be used to understand how changes in climate influence bed evolution
and GSD changes. Future applications could include coupling with bedrock erosion components to investigate how sediment
cover and grain size distributions affect bedrock incision rates, though this would require modifications to incorporate a
bedrock surface. Additionally, the component could be adapted to study sediment sorting and deposition patterns in alluvial
705 fan environments, where changes in channel slope and width strongly influence grain size distribution and depositional
processes.

While RiverBedDynamics already offers powerful modeling capabilities, there are exciting opportunities for future
enhancements. One potential improvement would be implementing a time-varying Manning's roughness that responds to bed
710 grain properties and water depth, such as the model proposed by Limerinos (1970). Additionally, incorporating bank erosion
and channel migration capabilities could improve predictions and make long-term simulations more realistic. To expand the
component's applicability to longer timescales, we could implement a morphological acceleration factor (Morgan et al.,
2020). This approach would allow for less frequent morphology calculations in slowly changing bed processes, extending
the component's use to landscape evolution runs spanning millennia or longer while improving computational efficiency.

715
For applications in mountain river systems, particularly those with high gradient longitudinal slopes, implementing the
equations developed by Schneider et al. (2015) and Yager et al. (2007, 2012) could provide more accurate bed load transport
calculations. This addition would enable explicit inclusion of large roughness elements and sediment supply limited
conditions common in steep streams. Another potential refinement is the inclusion of a critical shear stress that evolves with
720 sediment transport rate, similar to the approach of Johnson (2016). These potential enhancements build upon the strong
foundation of Landlab and in particular our proposed component RiverBedDynamics, expanding their already significant
capabilities in modeling complex river systems across various spatial and temporal scales.



725 7 Conclusion

We presented the first version of RiverBedDynamics, a Landlab component designed and built to simulate 2D sediment transport and river bed evolution with a special focus on gravel-bedded rivers. Coupling RiverBedDynamics with a OverlandFlow has created a LEM capable of providing accurate and detailed predictions of bed surface evolution in terms of elevation and grain size distribution. This new LEM is physically based and solves fundamental governing equations such as the conservation of mass in RiverBedDynamics, and mass and momentum in OverlandFlow, enhancing its reliability for simulating unsteady processes. The new component is flexible enough for short- and long-term simulations depending on the number of processes that can be included in each case. Our LEM's predictions were validated against analytical and previously reported solutions, demonstrating accurate representation of changes in bed surface elevation and grain size distribution. Both purely erosional and depositional cases were evaluated, with processes well captured in each scenario. Additionally, we employed a synthetic watershed to illustrate how the interaction between rainfall intensity distribution and sediment transport processes influences flow discharge and bed surface evolution across the domain.

While we have designed the first version of RiverBedDynamics to be as comprehensive as possible in representing sediment transport processes, there is potential for further enhancements and generalizations to expand its capabilities. Nonetheless, our LEM has demonstrated that the combination of OverlandFlow and RiverBedDynamics offers significant potential for simulating many typical scenarios encountered in practical river management situations and fundamental scientific research. We anticipate that future developments will focus on improving the representation of bank erosion, channel migration, and local angle of repose effects.

745

8 Code and data availability

The source code for RiverBedDynamics is available in a public Zenodo repository at <https://doi.org/10.5281/zenodo.14159914>. This repository contains the latest release version of the component, including all necessary files for running the model within the Landlab framework.

750

The example scripts and data used to create and run the test cases presented in this article are accessible in a separate Zenodo repository: <https://doi.org/10.5281/zenodo.14159904>. This repository includes all the necessary input files, parameters, and scripts to reproduce the results discussed in this paper.



755 Both repositories are open-source and freely available for use, modification, and distribution under MIT License. We encourage users to refer to the README files in each repository for detailed instructions on installation, dependencies, and usage. For any questions regarding the code or data, please contact the corresponding author.

9 Author contribution

760 AM, SA, NG, and EY conceptualized the component and defined its requirements; AM developed the component code; SA performed alpha and beta testing; NG conducted code review and implemented improvements; AM developed the test cases with SA reviewing them; AM wrote the original manuscript draft; SA, NG, and EY reviewed and edited the manuscript.

10 Competing interests

The authors declare that they have no conflict of interest.

765

11 Acknowledgements

This work was supported by NSF award number 1918459 to Anderson and Gasparini. We are grateful for the insightful conversations with Joel P. L. Johnson and Grace Guryan, which were instrumental in initiating this project. We extend our
770 sincere appreciation to Eric Hutton and Mark Piper from CSDMS for their invaluable assistance during the development of this Landlab component.

References

- Adams, J. M., Gasparini, N. M., Hobley, D. E. J., Tucker, G. E., Hutton, E. W. H., Nudurupati, S. S., & Istanbuluoglu, E.
775 (2017). The Landlab v1.0 OverlandFlow component: A Python tool for computing shallow-water flow across watersheds. *Geoscientific Model Development*, 10(4), 1645–1663. <https://doi.org/10.5194/gmd-10-1645-2017>
- Attal, M., Cowie, P. A., Whittaker, A. C., Hobley, D., Tucker, G. E., & Roberts, G. P. (2011). Testing fluvial erosion models using the transient response of bedrock rivers to tectonic forcing in the Apennines, Italy. *Journal of Geophysical Research: Earth Surface*, 116(F2), 2010JF001875. <https://doi.org/10.1029/2010JF001875>



- 780 Barnhart, K. R., Glade, R. C., Shobe, C. M., & Tucker, G. E. (2019). Terrainbento 1.0: A Python package for multi-model analysis in long-term drainage basin evolution. *Geoscientific Model Development*, 12(4), 1267–1297. <https://doi.org/10.5194/gmd-12-1267-2019>
- Barnhart, K. R., Hutton, E. W. H., Tucker, G. E., Gasparini, N. M., Istanbuluoglu, E., Hobbey, D. E. J., Lyons, N. J., Mouchene, M., Nudurupati, S. S., Adams, J. M., & Bandaragoda, C. (2020). Short communication: Landlab v2.0: a software
785 package for Earth surface dynamics. *Earth Surface Dynamics*, 8(2), 379–397. <https://doi.org/10.5194/esurf-8-379-2020>
- Barnhart, K. R., Hutton, E. W. H., Tucker, G. E., M. Gasparini, N., Istanbuluoglu, E., E. J. Hobbey, D., J. Lyons, N., Mouchene, M., Siddhartha Nudurupati, S., M. Adams, J., & Bandaragoda, C. (2020). Short communication: Landlab v2.0: A software package for Earth surface dynamics. *Earth Surface Dynamics*, 8(2), 379–397. [https://doi.org/10.5194/esurf-](https://doi.org/10.5194/esurf-8-379-2020)
790 [8-379-2020](https://doi.org/10.5194/esurf-8-379-2020)
- Barry, J. J., Buffington, J. M., & King, J. G. (2004). A general power equation for predicting bed load transport rates in gravel bed rivers. *Water Resources Research*, 40(10), 1–22. <https://doi.org/10.1029/2004WR003190>
- Bates, P. D., Horritt, M. S., & Fewtrell, T. J. (2010). A simple inertial formulation of the shallow water equations for efficient two-dimensional flood inundation modelling. *Journal of Hydrology*, 387(1–2), 33–45. <https://doi.org/10.1016/j.jhydrol.2010.03.027>
795
- Braun, J., & Willett, S. D. (2013). A very efficient O(n), implicit and parallel method to solve the stream power equation governing fluvial incision and landscape evolution. *Geomorphology*, 180–181, 170–179. <https://doi.org/10.1016/j.geomorph.2012.10.008>
- Campforts, B., Schwanghart, W., & Govers, G. (2017). Accurate simulation of transient landscape evolution by eliminating numerical diffusion: The TTLEM 1.0 model. *Earth Surface Dynamics*, 5(1), 47–66. [https://doi.org/10.5194/esurf-5-](https://doi.org/10.5194/esurf-5-47-2017)
800 [47-2017](https://doi.org/10.5194/esurf-5-47-2017)
- Cao, Z., Day, R., & Egashira, S. (2002). *Coupled and Decoupled Numerical Modeling of Flow and Morphological Evolution in Alluvial Rivers*. 128(3). [https://doi.org/10.1061/\(ASCE\)0733-9429\(2002\)128:3\(306\)](https://doi.org/10.1061/(ASCE)0733-9429(2002)128:3(306))



- Cheng, Z., Hsu, T. J., & Calantoni, J. (2017). SedFoam: A multi-dimensional Eulerian two-phase model for sediment transport
805 and its application to momentary bed failure. *Coastal Engineering*, 119, 32–50.
<https://doi.org/10.1016/j.coastaleng.2016.08.007>
- Colombini, M., & Stocchino, A. (2005). Coupling or decoupling bed and flow dynamics: Fast and slow sediment waves at
high Froude numbers. *Physics of Fluids*, 17(3). <https://doi.org/10.1063/1.1848731>
- Coulthard, T. J. (2001). Landscape evolution models: A software review. *Hydrological Processes*, 15, 165–173.
810 <https://doi.org/doi.org/10.1002/hyp.426>
- Coulthard, T. J., Macklin, M. G., & Kirkby, M. J. (2002). A cellular model of Holocene upland river basin and alluvial fan
evolution. *Earth Surface Processes and Landforms*, 27(3), 269–288. <https://doi.org/10.1002/esp.318>
- de Almeida, G. A. M., Bates, P., Freer, J. E., & Souvignet, M. (2012). Improving the stability of a simple formulation of the
shallow water equations for 2-D flood modeling. *Water Resources Research*, 48(5), 1–14.
815 <https://doi.org/10.1029/2011WR011570>
- Fernandez Luque, R., & Van Beek, R. (1976). Erosion And transport Of bed-load sediment. *Journal of Hydraulic Research*,
14(2), 127–144. <https://doi.org/10.1080/00221687609499677>
- Forte, A. M., Yanites, B. J., & Whipple, K. X. (2016). Complexities of landscape evolution during incision through layered
stratigraphy with contrasts in rock strength. *Earth Surface Processes and Landforms*, 41(12), 1736–1757.
820 <https://doi.org/10.1002/esp.3947>
- Furbish, D. J., Fathel, S. L., & Schmeeckle, M. W. (2017). Particle Motions and Bedload Theory. In D. Tsusumi & J. B.
Laronne (Eds.), *Gravel-Bed Rivers* (pp. 97–120). <https://doi.org/10.1002/9781118971437.ch4>
- Furbish, D. J., Fathel, S. L., Schmeeckle, M. W., Jerolmack, D. J., & Schumer, R. (2017). The elements and richness of particle
diffusion during sediment transport at small timescales. *Earth Surface Processes and Landforms*, 42(1), 214–237.
825 <https://doi.org/10.1002/esp.4084>
- Gasparini, N. M., Tucker, G. E., & Bras, R. L. (2004). Network-scale dynamics of grain-size sorting: Implications for
downstream fining, stream-profile concavity, and drainage basin morphology. *Earth Surface Processes and
Landforms*, 29(4), 401–421. <https://doi.org/10.1002/esp.1031>



- Ghimire, B., & Deng, Z.-Q. (2011). Event flow hydrograph-based method for shear velocity estimation. *Journal of Hydraulic Research*, 49(2), 272–275. <https://doi.org/10.1080/00221686.2011.552463>
- 830
- Goren, L., Willett, S. D., Herman, F., & Braun, J. (2014). Coupled numerical–analytical approach to landscape evolution modeling. *Earth Surface Processes and Landforms*, 39(4), 522–545. <https://doi.org/10.1002/esp.3514>
- Hobley, D. E. J., Adams, J. M., Siddhartha Nudurupati, S., Hutton, E. W. H., Gasparini, N. M., Istanbuluoglu, E., & Tucker, G. E. (2017). Creative computing with Landlab: An open-source toolkit for building, coupling, and exploring two-
- 835
- dimensional numerical models of Earth-surface dynamics. *Earth Surface Dynamics*, 5(1), 21–46. <https://doi.org/10.5194/esurf-5-21-2017>
- Huang, H. Q. (2010). Reformulation of the bed load equation of Meyer-Peter and Müller in light of the linearity theory for alluvial channel flow. *Water Resources Research*, 46(9), 2009WR008974. <https://doi.org/10.1029/2009WR008974>
- Johnson, J. P. L. (2016). Gravel threshold of motion: A state function of sediment transport disequilibrium? *Earth Surface*
- 840
- Dynamics*, 4(3), 685–703. <https://doi.org/10.5194/esurf-4-685-2016>
- Juez, C., Ferrer-Boix, C., Murillo, J., Hassan, M. A., & García-Navarro, P. (2016). A model based on Hirano-Exner equations for two-dimensional transient flows over heterogeneous erodible beds. *Advances in Water Resources*, 87, 1–18. <https://doi.org/10.1016/j.advwatres.2015.10.013>
- Lamb, M. P., Dietrich, W. E., & Venditti, J. G. (2008). Is the critical shields stress for incipient sediment motion dependent
- 845
- on channel-bed slope? *Journal of Geophysical Research: Earth Surface*, 113(2), F02008. <https://doi.org/10.1029/2007JF000831>
- Langston, A. L., & Tucker, G. E. (2018). Developing and exploring a theory for the lateral erosion of bedrock channels for use in landscape evolution models. *Earth Surface Dynamics*, 6(1), 1–27. <https://doi.org/10.5194/esurf-6-1-2018>
- Li, Q., Gasparini, N. M., & Straub, K. M. (2018). Some signals are not the same as they appear: How do erosional landscapes
- 850
- transform tectonic history into sediment flux records? *Geology*, 46(5), 407–410. <https://doi.org/10.1130/G40026.1>
- Limerinos, J. T. (1970). *Determination of the Manning coefficient from measured bed roughness in natural channels* *Roughness in Natural Channels* (p. 53). U.S. Geological Survey.



- Mao, L., Uyttendaele, G. P., Iroumé, A., & Lenzi, M. A. (2008). Field based analysis of sediment entrainment in two high gradient streams located in Alpine and Andine environments. *Geomorphology*, 93(3–4), 368–383.
855 <https://doi.org/10.1016/j.geomorph.2007.03.008>
- Meyer-Peter, E., & Müller, R. (1948). Formulas for Bed-Load Transport. *Proceedings of the 2nd Meeting of the International Association of Hydraulic Research*, 39–64. <https://doi.org/1948-06-07>
- Mitchell, N., & Forte, A. M. (2023). Tectonic advection of contacts enhances landscape transience. *Earth Surface Processes and Landforms*, 48(7), 1450–1469. <https://doi.org/10.1002/esp.5559>
- 860 Morgan, J. A., Kumar, N., Horner-Devine, A. R., Ahrendt, S., Istanbuloglu, E., & Bandaragoda, C. (2020). The use of a morphological acceleration factor in the simulation of large-scale fluvial morphodynamics. *Geomorphology*, 356, 107088. <https://doi.org/10.1016/j.geomorph.2020.107088>
- Mueller, E. R., Pitlick, J., & Nelson, J. M. (2005a). Variation in the reference Shields stress for bed load transport in gravel-bed streams and rivers. *Water Resources Research*, 41(4), 1–10. <https://doi.org/10.1029/2004WR003692>
- 865 Mueller, E. R., Pitlick, J., & Nelson, J. M. (2005b). Variation in the reference Shields stress for bed load transport in gravel-bed streams and rivers. *Water Resources Research*, 41(4), 1–10. <https://doi.org/10.1029/2004WR003692>
- Paola, C., & Voller, V. R. (2005). A generalized Exner equation for sediment mass balance. *Journal of Geophysical Research: Earth Surface*, 110(4), 1–8. <https://doi.org/10.1029/2004JF000274>
- Parker, G. (1990). Surface-based bedload transport relation for gravel rivers. *Journal of Hydraulic Research*, 28(4), 417–436.
870 <https://doi.org/10.1080/00221689009499058>
- Parker, G. (1991). Selective Sorting and abrasion of river gravel I Theory. *Journal of Hydraulic Engineering*, 117(2), 131–147.
- Parker, G., Paola, C., & Leclair, S. (2000). Probabilistic Exner Sediment Continuity Equation for Mixtures with no Active Layer. *Journal of Hydraulic Engineering*, 126(11), 818–826. [https://doi.org/10.1061/\(ASCE\)0733-9429\(2000\)126:11\(818\)](https://doi.org/10.1061/(ASCE)0733-9429(2000)126:11(818))
875



- Pfeiffer, A., Barnhart, K., Czuba, J., & Hutton, E. (2020). NetworkSedimentTransporter: A Landlab component for bed material transport through river networks. *Journal of Open Source Software*, 5(53), 2341. <https://doi.org/10.21105/joss.02341>
- Sanchez, A., & Wu, W. (2011). A non-equilibrium sediment transport model for coastal inlets and navigationChannels. *Journal of Coastal Research*, 59, 39–48. <https://doi.org/10.2112/SI59-005.1>
880
- Schneider, J. M., Rickenmann, D., Turowski, J. M., Bunte, K., & Kirchner, J. W. (2015). Applicability of bed load transport models for mixed-size sediments in steep streams considering macro-roughness. *Water Resource Research*, 51, 5260–5283. <https://doi.org/10.1002/2014WR016417>
- Shobe, C. M., Tucker, G. E., & Barnhart, K. R. (2017). The SPACE 1.0 model: A Landlab component for 2-D calculation of sediment transport, bedrock erosion, and landscape evolution. *Geoscientific Model Development*, 10(12), 4577–4604. <https://doi.org/10.5194/gmd-10-4577-2017>
885
- Smith, H. E. J., Monsalve, A. D., Turowski, J. M., Rickenmann, D., & Yager, E. M. (2023). Controls of local grain size distribution, bed structure and flow conditions on sediment mobility. *Earth Surface Processes and Landforms*, 48(10), 1990–2004. <https://doi.org/10.1002/esp.5599>
- Song, Y., Xu, Y., & Liu, X. (2020). Physically Based Sand Slide Method in Scour Models Based on Slope-Limited Diffusion. *Journal of Hydraulic Engineering*, 146(11), 1–11. [https://doi.org/10.1061/\(asce\)hy.1943-7900.0001814](https://doi.org/10.1061/(asce)hy.1943-7900.0001814)
890
- Temme, A. J. A. M., Armitage, J., Attal, M., van Gorp, W., Coulthard, T. J., & Schoorl, J. M. (2017). Developing, choosing and using landscape evolution models to inform field-based landscape reconstruction studies. *Earth Surface Processes and Landforms*, 42, 2167–2183. <https://doi.org/10.1002/esp.4162>
- Toro-Escobar, C. M., Paola, C., & Parker, G. (1996). Transfer function for the deposition of poorly sorted gravel in response to streambed aggradation. *Journal of Hydraulic Research*, 34(1), 35–53. <https://doi.org/10.1080/00221689609498763>
895
- Tucker, G. E., & Hancock, G. R. (2010). Modelling landscape evolution. *Earth Surface Processes and Landforms*, 35(1), 28–50. <https://doi.org/10.1002/esp.1952>



- Tucker, G. E., Hutton, E. W. H., Piper, M. D., Campforts, B., Gan, T., Barnhart, K. R., Kettner, A. J., Overeem, I., Peckham, S. D., McCready, L., & Syvitski, J. (2022). CSDMS: A community platform for numerical modeling of Earth surface processes. *Geoscientific Model Development*, 15(4), 1413–1439. <https://doi.org/10.5194/gmd-15-1413-2022>
- Tucker, G. E., Lancaster, S. T., Gasparini, N. M., Bras, R. L., & Rybarczyk, S. M. (2001). An object-oriented framework for distributed hydrologic and geomorphic modeling using triangulated irregular networks. *Computers and Geosciences*, 27(8), 959–973. [https://doi.org/10.1016/S0098-3004\(00\)00134-5](https://doi.org/10.1016/S0098-3004(00)00134-5)
- 905 Tucker, G. E., & Slingerland, R. L. (1994). Erosional dynamics, flexural isostasy, and long-lived escarpments: A numerical modeling study. *Journal of Geophysical Research*, 99(B6). <https://doi.org/10.1029/94jb00320>
- Van De Wiel, M. J., Coulthard, T. J., Macklin, M. G., & Lewin, J. (2007). Embedding reach-scale fluvial dynamics within the CAESAR cellular automaton landscape evolution model. *Geomorphology*, 90(3–4), 283–301. <https://doi.org/10.1016/j.geomorph.2006.10.024>
- 910 Whipple, K. X., Forte, A. M., DiBiase, R. A., Gasparini, N. M., & Ouimet, W. B. (2017). Timescales of landscape response to divide migration and drainage capture: Implications for the role of divide mobility in landscape evolution. *Journal of Geophysical Research: Earth Surface*, 122(1), 248–273. <https://doi.org/10.1002/2016JF003973>
- Whipple, K. X., & Tucker, G. E. (2002). Implications of sediment-flux-dependent river incision models for landscape evolution. *Journal of Geophysical Research: Solid Earth*, 107(B2). <https://doi.org/10.1029/2000JB000044>
- 915 Wilcock, P. R., & Crowe, J. C. (2003). Surface-based transport model for mixed-size sediment. *Journal of Hydraulic Engineering*, 129(2), 120–128. [https://doi.org/10.1061/\(ASCE\)0733-9429\(2003\)129:2\(120\)](https://doi.org/10.1061/(ASCE)0733-9429(2003)129:2(120))
- Wong, M., & Parker, G. (2006). Reanalysis and Correction of Bed-Load Relation of Meyer-Peter and Müller Using Their Own Database. *Journal of Hydraulic Engineering*, 132(11), 1159–1168. [https://doi.org/10.1061/\(ASCE\)0733-9429\(2006\)132:11\(1159\)](https://doi.org/10.1061/(ASCE)0733-9429(2006)132:11(1159))
- 920 Yager, E. M., Dietrich, W. E., Kirchner, J. W., & McArdell, B. W. (2012). Prediction of sediment transport in step-pool channels. *Water Resources Research*, 48(1), W01541. <https://doi.org/10.1029/2011WR010829>
- Yager, E. M., Kirchner, J. W., & Dietrich, W. E. (2007). Calculating bed load transport in steep boulder bed channels. *Water Resources Research*, 43(7), W07418. <https://doi.org/10.1029/2006WR005432>