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1 Simulating bank erosion over an extended natural sinuous river reach using a
2 universal slope stability algorithm coupled with a morphodynamic model

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10 **Abstract**

11 Meandering river channels are often associated with cohesive banks. Yet, only a few river modelling
12 packages include geotechnical and plant effects. Existing packages are solely compatible with single-
13 threaded channels, require a specific mesh structure, derive lateral migration rates from hydraulic
14 properties, determine stability based on friction angle, rely on non-physical assumptions to describe cut
15 offs, or exclude floodplain processes and vegetation. In this paper, we evaluate the accuracy of a new
16 geotechnical module that was developed and coupled with Telemac-Mascaret to address these limitations.
17 Innovatively, the newly developed module relies on a fully configurable, universal genetic algorithm with
18 tournament selection that permits it 1) to assess geotechnical stability along potentially unstable slope
19 profiles intersecting liquid-solid boundaries, and 2) to predict the shape and extent of slump blocks while
20 considering mechanical plant effects, bank hydrology, and the hydrostatic pressure caused by the flow.
21 The profiles of unstable banks are altered while ensuring mass conservation. Importantly, the new
22 stability module is independent of mesh structure and can operate efficiently along multi-threaded
23 channels, cut offs, and islands. Data collected along a 1.5 km long reach of the semi-alluvial Medway
24 Creek, Canada, over a period of 3.5 years are used to evaluate the capacity of the coupled model to
25 accurately predict bank retreat in a meandering river channels, and to evaluate the extent to which the

26 new model can be applied to a natural river reach located in a complex environment. Our results indicate
27 that key geotechnical parameters can indeed be adjusted to fit observations, even with a minimal
28 calibration effort; and that the model correctly identifies the location of the most severely eroded bank
29 regions. The combined use of genetic and spatial analysis algorithms, in particular for the evaluation of
30 geotechnical stability independently of the hydrodynamic mesh, permits the consideration of biophysical
31 conditions for an extended river reach with complex bank geometries, with only a minor increase in run
32 time. Further improvements with respect to plant representation could assist scientists in better
33 understanding channel-floodplain interactions, and in evaluating channel designs in river management
34 projects.

35 **Keywords:** Meandering, river morphodynamics, fluvial modelling, and geotechnical slope stability.

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37

38 **1. Introduction**

39 Morphodynamic models have been employed for decades by researchers and practitioners to examine the
40 evolution of alluvial river channels (e.g. [Rinaldi et al., 2008](#); [Tal and Paola, 2010](#); [Ham and Church,](#)
41 [2012](#)). In particular, two-dimensional non-linear and linear models based on the shallow water equations,
42 combined with a computational mesh that can evolve due to sediment transport, are increasingly used to
43 determine the morphological evolution of meandering channels ([Darby et al., 2002](#); [Langendoen et al.,](#)
44 [2016](#)). Most of these models involve a large number of assumptions and simplifications to combine
45 fluvial and bank erosion processes into a runnable solution, often neglecting floodplain heterogeneity in
46 terms of morphology ([Pittaluga and Seminara, 2011](#)), channel bedforms ([Shen, 1984](#); [Parker et al., 2011](#)),
47 multithreading ([Camporeale et al., 2013](#)), sedimentology and stratigraphy ([Simon et al., 2000](#); [Malkinson](#)
48 [and Wittenberg, 2007](#); [Lai et al., 2012](#)), bank hydrology ([Pollen, 2007](#); [Pollen-Bankhead and Simon,](#)
49 [2010](#)), and flow regimes. As a result, the contribution of these processes to channel evolution is poorly
50 understood ([Güneralp and Marston, 2012](#)). In addition, riparian plants, which alter channel/floodplain
51 roughness and provide mechanical soil reinforcement ([Abernethy and Rutherford, 1998](#); [Van de Wiel and](#)
52 [Darby, 2007](#); [Thomas and Pollen-Bankhead, 2010](#)), should be included in morphodynamic models,
53 although this is seldom the case ([Bertoldi et al., 2014](#)). Finally, opportunities to establish spatial
54 connections between floodplain components, for instance between hydrological processes and riparian
55 plants ([Perucca et al., 2007](#); [Mitsch and Gosselink, 2010](#)), are often missed ([Malkinson and Wittenberg,](#)
56 [2007](#); [Lai et al., 2012](#)).

57 A few river models have been enhanced to include bank retreat algorithms (e.g. [El Kadi](#)
58 [Abderrezzak et al. 2016](#)). The linear near-bank excess velocity approach (known as HIPS, from
59 [Hasegawa \(1977\)](#) and [Ikeda et al. \(1981\)](#)) relies on an erodibility coefficient to lump the effects of flow,
60 soil and vegetation properties on bank retreat rates (see [Johannesson and Parker, 1989](#); [Zolezzi and](#)
61 [Seminara, 2001](#); [Posner and Duan, 2012](#)), thus making it impossible to isolate the specific causes for
62 retreat, and to "entirely" simulate long-term planimetric and morphological evolution due to the lack of

63 analytical solution of neck/chute cutoff (Chen and Tang, 2012). In addition, these models do not
64 guarantee sediment continuity, and assume a flat-bedded channel with few perturbations (Coulthard and
65 Van de Wiel, 2006; Pittaluga and Seminara, 2011). When riparian vegetation is considered, its effect is
66 typically limited to altering bed roughness, although a few notable exceptions exist where vegetation was
67 connected to other floodplain processes (e.g. Collins et al., 2004; Perucca et al., 2007; Iwasaki et al.,
68 2016). Models based on HIPS concepts do, however, have the advantage of allowing for long reaches to
69 be simulated at relatively low computational cost (e.g. Schwenk et al., 2015). A few non-linear
70 morphodynamic models have also been coupled to physically-based bank erosion modules (e.g.
71 Darby et al., 2002; Lai et al., 2012; Langendoen et al., 2016). However, several limitations remain
72 regarding these solutions. The most severe ones are probably their incompatibility with long
73 spatiotemporal scales (Pittaluga and Seminara, 2011), the lack of physically-based equations in the
74 implemented bank stability assessment and retreat processes, and the integration of assumptions such as
75 arbitrary bank and planform geometries (e.g. Duan and Julien, 2010; Langendoen et al., 2016). The main
76 difficulty seems associated with the inclusion of processes acting at different spatiotemporal scales
77 compared to the shallow-water flow equations employed in the hydrodynamic models (Williams et al.,
78 2016a).

79 The integration of geotechnical algorithms applicable over long river reaches in a natural
80 environment with complex floodplain has seldom been attempted (Rousseau et al., 2014a,b; Evangelista
81 et al., 2015). This paper presents a novel set of algorithms implemented in a new physics-based,
82 deterministic model of channel-floodplain co-evolution. The model is capable of simulating mass wasting
83 events, including river bank failures, while also taking into account the specific biophysical context. Here,
84 it is integrated into a two-dimensional unstructured grid morphodynamic model (Telemac-Mascaret), but
85 it could also be implemented in other modelling software, with relatively minor adjustments. The novel
86 aspects included in the present modelling investigation is fivefold: 1) physics-based algorithms allowing
87 to parameterize lateral erosion using physical, measurable quantities; 2) a genetic algorithm that can

88 automatically select between rotational and translational failure mechanisms, depending on local
89 biophysical conditions; and allow 3) larger spatial and temporal scales than commonly employed in
90 physics-based bank erosion modelling due to greater computational efficiency; 4) a bank erosion module
91 independent of mesh structure, i.e. there is no need to impose a body-fitted coordinate system; and
92 5) model calibration using data from a complex natural site. Furthermore, because the model includes the
93 interaction between an alluvial river channel and its (vegetated or non-vegetated) floodplain, it is able to
94 simulate lateral river channel adjustments that can lead to the development of meandering, wandering or
95 braided river planform geometries. This paper thus directly addresses the issues identified by [Williams et](#)
96 [al. \(2016b\)](#) who noted that "Future model development efforts should be directed toward improving the
97 realism of bank erosion processes in the model. In particular, the bank erosion scheme needs to be made
98 independent of grid resolution and orientation." (p. 6639)

99 **2. Overview of model components**

100 The geotechnical and riparian vegetation modules presented here were integrated into the Open Telemac-
101 Mascaret suite of mathematical solvers to include additional fluvial processes, namely lateral adjustments
102 through mass wasting of river banks and the effects of floodplain vegetation on geotechnical stability
103 ([Rousseau et al., 2014a, b](#)).

104 *2.1. Hydrodynamics*

105 The two-dimensional (2D) version of Telemac was selected to minimize computation time, although the
106 three-dimensional (3D) version could easily be coupled to the new modules if required in future projects.
107 The equations governing fluid motion, in their non-conservative form and Cartesian coordinates, are
108 described in [Galland et al. \(1991\)](#). The [Smagorinsky \(1963\)](#) model was selected to consider turbulent
109 viscosity whilst minimizing computational effort. The default advection scheme, i.e. the method of
110 characteristics, was selected for horizontal flow velocities and depth. For boundary conditions, flow
111 discharge and uniform velocities were imposed at the channel inlet with a free surface elevation at the

112 outlet. In all simulations, the inlet of the flow comprised the wet nodes on the left domain side, whereas
113 the outlet included all the mesh nodes located on the right domain side so that the outlet may adjust its
114 location during a simulation.

115 *2.2. Sediment transport and bed deformation*

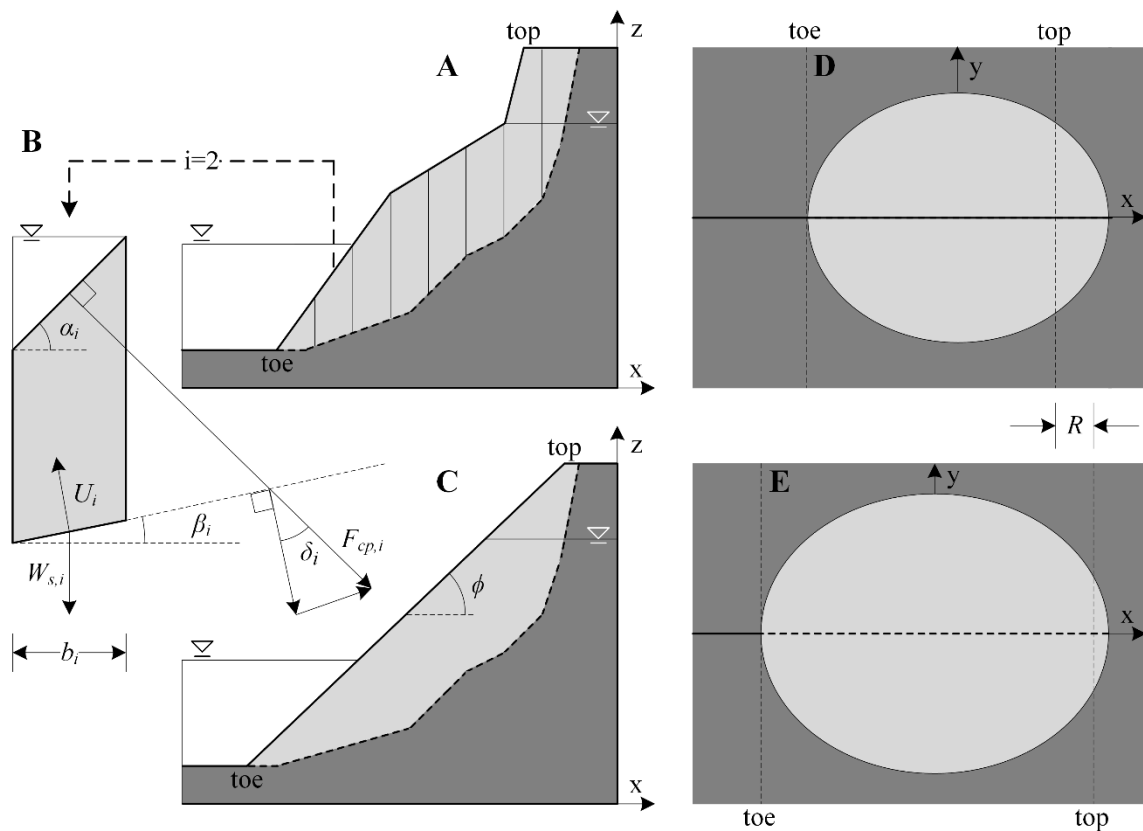
116 Sediment transport is calculated by the module Sisyphe. The [Meyer-Peter and Müeller \(1948\)](#) bed load
117 formula was selected to calculate transport rates during simulations due to its compatibility with respect
118 to sediment grain size at our field site (see below). Hiding/exposure is calculated using [Egiazaroff \(1965\)](#).
119 The formulae related to the configurations considered in this study are thoroughly described in [Villaret](#)
120 [\(2010\)](#). Bed evolution due to bed load transport is calculated using the Exner equation.

121 The sediment transport module Sisyphe includes an algorithm that can simulate sediment slide; this
122 was deactivated to let geotechnically stable river banks be steeper than the friction angle of sediment, and
123 more importantly, to prevent having two modules competing for a single process. An option was enabled
124 to include curvature effects on the direction of particle entrainment to compensate for the fact that the
125 flow is depth-averaged. This feature was enabled in all simulations due to its relevance for the study of
126 meandering processes. The effects of transport magnitude and direction on local topography are estimated
127 using [Koch and Flokstra \(1981\)](#).

128 *2.3. Lateral adjustments*

129 The primary objective of this research was to develop an alternative methodology to simulate river bank
130 retreat within a computational fluid dynamics (CFD) model, using a set of algorithms that is compatible
131 with unstructured meshes, for a wide spectrum of alluvial rivers, and with any modelling code offering
132 finite element spatial discretization. This new framework relies almost completely on vector-based spatial
133 analysis. In the generic framework, the stability sub-model ignores the location of channel boundaries.
134 Instead, it performs stability assessments across the floodplain, considering flow conditions, antecedent
135 soil moisture, soil and sedimentological properties, and plant cover (if the vegetation module is enabled,
136 see below). The channel planform will evolve when the banks are unstable along its boundaries.

137 The module is divided into five components. A landscape analysis algorithm generates a network
 138 of transects along which slope stability assessments are performed during a morphodynamic simulation
 139 (see the inset in Fig. 1). Although this algorithm can be configured to detect potentially unstable slopes
 140 anywhere across the simulation domain, for the current study an option was enabled to perform the
 141 analysis strictly along the external river bank of meander bends through edge detection. A genetic
 142 algorithm searches for the geometry of the most likely failure profile and returns the lowest safety factor
 143 (F_s , the ratio of shear strength to shear stress – see Eq. 3-4 below) encountered. The geotechnical module



144
 145 **Fig. 1.** Bank slope stability assessment using Bishop's simplified method of slices. A) Initial profile,
 146 indicating stable bank region (dark gray) and failure block (light gray). B) Stability is calculated based on
 147 forces acting on vertical slices through the failure block (Eq. 3-4). C) Profile following deposition of the
 148 failure block at the friction angle of the bank material. Note that this is a two-dimensional transect of a
 149 three-dimensional bank; the volume of the failure block in C is the same as in A. The upper limit of the
 150 failure block in C, i.e. a planar 3D surface within an elliptical zone, is adjusted to ensure that the two
 151 failure block volumes are equal. Elliptical zones affected by D) block failure and E) deposition. The
 152 character R indicates the distance of bank retreat.

153 includes a river bank hydrology module that computes water table elevation in the river bank as a
154 function of flow stage and hydraulic conductivity. Finally, the algorithm analyses the geotechnical
155 stability of a river bank in search of a failure block geometry that minimizes the safety factor (during the
156 erosion phase, the unknown is the lower extent of the failure block, represented by the dashed line in
157 Fig. 1A), removes the unstable failure block and deposits the material downslope at the friction angle
158 (Fig. 1C, also see section 2.3.5 below), and updates the computational mesh nodes while ensuring mass
159 conservation (the unknown is the upper extent of the altered unstable block during the deposition phase)
160 (see Rousseau et al. (2014a,b) for more details). The upper extent of the post-failure block is a three-
161 dimensional, planar surface with an elliptical shape. The mesh nodes affected by a failure are those
162 located in the elliptical erosion and deposition zones (Fig. 1D and 1E) with the vertical displacement at a
163 node inversely proportional to its horizontal distance on the stability-analysis transect.

164 2.3.1. Terrain analysis

165 The geotechnical model evaluates terrain stability along a large number of analysis transects carefully
166 placed across the landscape. Each transect is oriented in the direction of the steepest ascent and adjusted
167 in length to extend from the lowest to the highest elevation in this direction. Transect orientation is done
168 by spinning a transect of length $trLen$ around its center of mass, and selecting the angle that minimizes the
169 surface elevation's root mean square error between the points comprised in two transects located on each
170 side of a trial transect at a distance $trLen / 2$. A smoothing stage adjusts the direction of each transect
171 with respect to the direction of their immediate neighbours. Each transect is extended until an increase in
172 length no longer leads to any substantial elevation change, defined by a threshold parameter (10° in our
173 simulations) to improve efficiency, as otherwise geotechnical stability would be evaluated for very
174 shallow banks that are unlikely to erode. Each transect is then transposed into a 2D bank profile, which is
175 analysed for stability using the method of slices and powered by a genetic algorithm, both of which are
176 described in details in the following sections. By default, transects regenerate at each iteration of the
177 geotechnical module. However, re-generation was disabled in our simulations to facilitate the statistical

178 comparison of retreat rates between scenarios. This strategy could be adopted due to the relatively low
 179 retreat rates and short simulations in our study. Although the location of transect centers was constant in
 180 time, each transect elongated and rotated to adjust to the evolving bank morphology.

181 Several options are available in the geotechnical module to define transect density, distribution
 182 pattern and admissibility criteria, e.g. in terms of wetness/dryness and length. In the context of a study
 183 involving lateral erosion in meandering channels, the edge detection algorithm was selected; it distributes
 184 transects at equal distance along solid-liquid boundaries. Its independence of mesh structure means that it
 185 can efficiently detect river banks along multi-threaded channels, cut offs, and islands.

186 2.3.2. Genetic algorithm

187 Any slope stability analysis includes an algorithm that devises a set of potential slip surfaces to be
 188 evaluated for their geotechnical stability. Given a two-dimensional geotechnical stability analysis, a
 189 solution with identifier id is a series of connected nodes delineating the lower limit of an unstable soil
 190 block, i.e. the dashed line in Fig. 1A. Therefore, a solution can be described by the following vector:

$$191 \quad \vec{S}_{id} = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_{n-1}, \vec{v}_n\} \quad (1)$$

192 where \vec{v}_i is the node at rank i along a slip surface. The solution with the lowest F_s value is the most likely
 193 to occur.

194 Grid-search patterns are usually employed to list potential slip surfaces. For instance, this can be
 195 achieved by varying the location of the centre of the arc describing the shape of a circular slip surface,
 196 along with its radius. Here, a genetic algorithm with tournament selection, improved from the work of Li
 197 et al. (2010), was implemented in the geotechnical module to converge towards a critical solution more
 198 rapidly. A *child* solution is created by combining two existing solutions, i and j , such that:

$$199 \quad \vec{S}_{child} = \eta \vec{S}_i + (\eta - 1) \vec{S}_j \quad (2)$$

200 where $\eta = [0,1]$ is a randomly generated cross-over ratio. During cross-over, mutation has a probability of
 201 happening, in which case a randomly selected node comprised in *child* solution is displaced. A set of

202 matching rules, namely partner exclusivity, child count policy, and prevention of breeding between
 203 relatives, allows the variability within the pool of solutions to be optimized (Li et al., 2010). For example,
 204 a child count policy means that two solutions S_i and S_j (a solution is a potential slip surface or the dashed
 205 line in Fig. 1B) can produce a pre-defined maximum number of solutions after randomly mixing their
 206 geometrical characteristics. The lack of such a policy limits genetic variability, and thus increases the
 207 time required to converge toward a stable solution, i.e. finding the failure block associated with the lowest
 208 safety factor. Finally, a user-specified migration rate dictates the probability for a solution to be created
 209 randomly rather than being the result of a cross-over.

210 In the current context, we can define a generation as the set of n solutions that were created from an
 211 initial population. After each generation, the most critical slip surface(s) are kept, the least critical are
 212 discarded, and new randomly selected surfaces survive to the next round. The search process terminates
 213 when the most critical slip surface remains unaltered for a number of consecutive generations.

214 2.3.3. Slope stability assessment

215 Bishop's (1955) modified method of slices (Fig. 1) was slightly adjusted to quantify the geotechnical
 216 stability of the soil along a transect while considering the flow's confining pressure and soil pore-water
 217 pressure. Combined with the genetic algorithm, it can produce planar, circular and non-circular slip
 218 surfaces. Given a 2D bank profile and potential slip surface, the following set of equations must be solved
 219 by iteration:

$$220 \quad F_s = \frac{\sum_{i=1}^n \frac{cb_i + (\gamma_{s,i} + F_{cp,i} - U_i b_i) \tan \phi}{m_i}}{\sum_{i=1}^n \gamma_{s,i} \sin \beta_i + F_{cp,i} \sin \delta_i} \quad (3)$$

$$221 \quad m_i = \cos \beta_i + \frac{\sin \beta_i \cdot \tan \phi}{F_s} \quad (4)$$

222 where F_s = safety factor; $\gamma_{s,i}$ = weight of soil material in slice i out of n ; U_i = the pore water pressure at the
 223 base of a slice of width b_i , basal angle β_i and top angle α_i ; δ_i = angle between the result of hydrostatic

224 confining force and normal to failure plane; ϕ = friction angle of the soil material; and m_i = m-term in
 225 Bishop formula. Pore-water pressure is given by:

$$226 \quad U = \frac{\rho g}{z_{wt} - z_b} \quad (5)$$

227 where g = acceleration due to gravity, z_{wt} = elevation of the water table, and z_b = elevation at the base of a
 228 slice. A first approximation of F_s value is done using the ordinary method of slices (Fellenius, 1927). The
 229 confined water pressure is given by:

$$230 \quad F_{cp,i} = \gamma_{w,i} \cos \beta_i \quad (6)$$

231 where $\gamma_{w,i}$ = weight of water. Any solution resulting in a safety factor lower than unity is said to be
 232 unstable and is expected to result in a slope failure.

233 2.3.4. River bank hydrology

234 A saturated river bank, combined with a falling flow stage, can trigger mass wasting events (Thorne,
 235 1982). To account for the lag effect between free surface and water table elevations, a simple river bank
 236 hydrology module is used to calculate water table elevation. According to this module, water table
 237 elevation (z'_{wt}) at a time $t=t_0+\Delta t$ is given by:

$$238 \quad z'_{wt} = z_{fs} - (z_{fs} - z_{wt})e^{-k\Delta t} \quad (7)$$

239 where t_0 = time at the previous iteration, Δt = time step, $t=t_0+\Delta t$ = time at the current iteration, z_{wt} = water
 240 table elevation at time t_0 , z'_{wt} = water table elevation at time t ; z_{fs} = flow surface elevation at time t_0 ; and
 241 k = rate of convergence of the water table elevation toward z_{fs} . The constant k is adjusted according to the
 242 hydraulic conductivity of the bank material, and thus represents how quickly the water table adapts to a
 243 change in the river's flow stage. Two k -values are required per simulation: one for the rising limb of a
 244 flood hydrograph, and one for its falling limb (see values below).

245

246 2.3.5. Slump block removal and deposition

247 The slope stability analysis performed along each transect returns the side profile, i.e. the two-
 248 dimensional curve representing the critical slip surface (Fig. 1A). Each mesh node affected by mass
 249 wasting relocates according to its position $(d_B/r_B, d_A/r_A)$ with respect to the boundaries of an elliptical
 250 erosion surface (Fig. 2). The ellipse has the length of the unstable section of the analysis transect and a
 251 user-defined relative width, i.e. r_B/r_A in Fig. 2; a value of 0.75 was selected for our simulations to cover
 252 the bank region located between a transect's immediate neighbours. A mesh node located along the edge
 253 of the ellipse is not affected by a failure. Conversely, displacement computed is greatest along the transect
 254 for each d_A/r_A value. A mesh node located within the ellipse has a vertical displacement (dz) that is
 255 computed as a linear function of the distance between the transect and the edge of the ellipse, in the
 256 direction orthogonal to the transect, i.e.:

$$257 \quad dz = dz_A \cdot \frac{d_B}{r_B \sqrt{1 - \left(\frac{d_A}{r_A}\right)^2}} \quad (8)$$

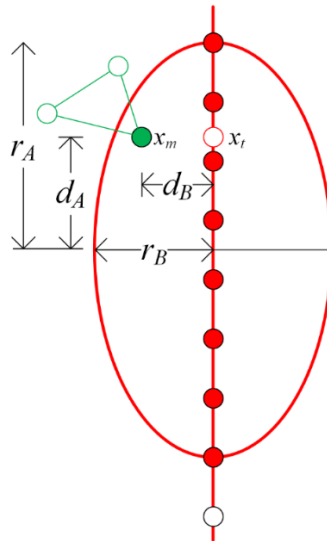
258 where r_A and $r_B = r_A \cdot k_f$ = lengths of the semi-axes A and B , k_f = width-to-length ratio of all ellipses
 259 defining an erosion or deposition zone, d_A and d_B = distances from the ellipse's centre to the mesh node
 260 along each semi-axis, dz_A = elevation change at a distance d_A from the centre of the ellipse in the direction
 261 of the mesh node x along the axis A . The value of dz_A is obtained by interpolating elevation change at
 262 node x_m using the two nearest transect nodes. Mesh elements that are intersecting the transect snap to the
 263 slip surface (lower line in Fig. 1A) after a vertical translation and/or a rotation around an axis orthogonal
 264 to the two-dimensional profile. The volume of the unstable block is calculated by subtracting the pre- and
 265 post-failure computational meshes, assuming that the neighbouring transects are stable. The unstable
 266 slope material deposits in an elliptical zone at the toe of the slope at the friction angle of the bank
 267 material. An algorithm similar to the erosion algorithm is used, with one important distinction: a solver is
 268 required to ensure that the volume of eroded soil is equal to the deposited volume, and thus guarantees
 269 mass conservation. Note that the elliptical zone where deposition occurs is usually not identical to the

270 ellipse in which erosion occurs since material moves down slope during a failure.

271 2.3.6. Soil properties

272 In its simplest configuration mode, the geotechnical module defines bank material in terms of soil class
273 and degree of compaction. The other relevant sedimentological parameters, i.e. mean grain size, mass
274 density, friction angle, cohesion and porosity, are calculated using tabular data available in NAVFAC
275 (1986), Swiss Standard SN 670 010b (1999), and MnDOT (2015). The soil material below water table is
276 assumed to be saturated; it is partially saturated if fine-textured and located above the water table while
277 being affected by capillary rise; and it is dry otherwise. Default values can be over-ridden by the model
278 user. Although the module is capable of recognizing multiple soil layers, bank material was assumed to
279 have uniform properties in our simulations.

280



281 **Fig. 2.** Geometry of an erosion or deposition zone (Fig. 1D and 1E) within an elliptical zone with flatness
282 r_B / r_A . Any mesh node x_m , located within the ellipse, as well as the red-filled nodes along the analysis
283 transect are unstable. In erosion mode, the elevation change at mesh node x_m depends on the distance d_B .
284 However, in deposition mode, the elevation of transect node x_t is assigned to mesh node x_m for each
285 tridimensional planar surface representing a potential depositional fan surface.
286

287 2.3.7. Riparian vegetation

288 At each geotechnical iteration, the plant evolution module (see Rousseau et al. (2014a) for details)
289 transfers information to the geotechnical module regarding the physiological plant properties that can

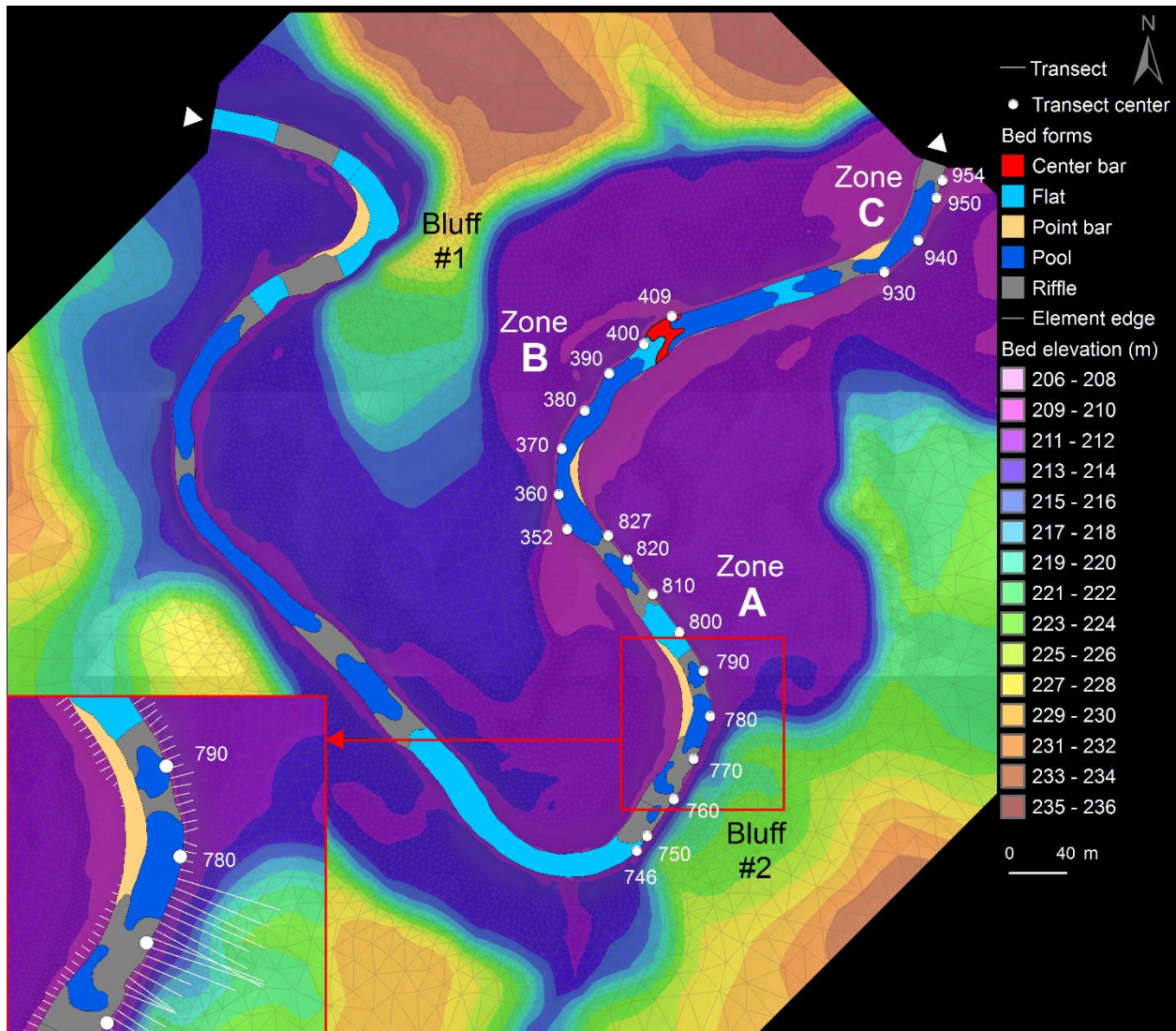
290 influence the mechanical properties of the river bank: depth and radius of rooting zone; cohesion due to
291 roots; trunk height, spacing and diameter. The plant evolution module also includes functionalities to
292 generate a plant cover and manage plant growth. In this study, however, dry mesh nodes were covered
293 with trees from a single species associated with a single set of physiological properties that remained
294 constant in time.

295 **3. Method**

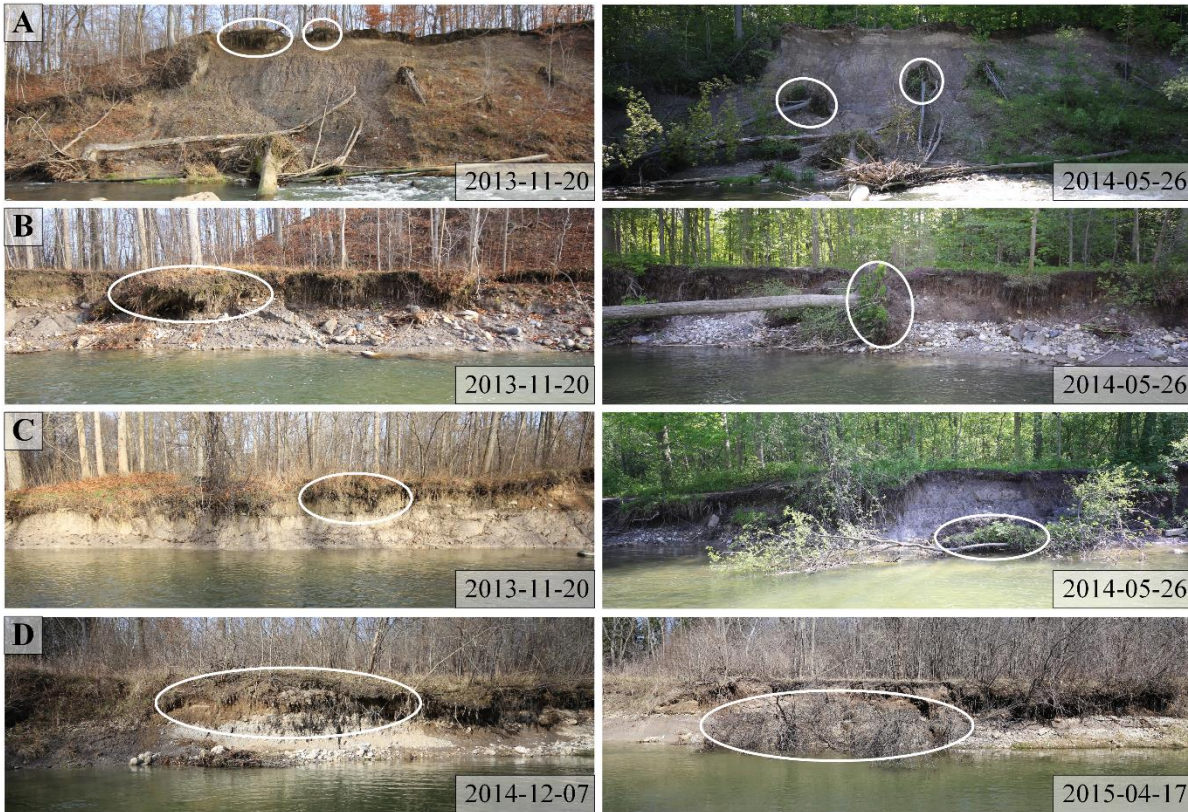
296 3.1. Field data

297 The model is evaluated against hydrological and morphological data collected along a 1.5 km-long reach
298 of the semi-alluvial river Medway Creek, London, Ontario (Fig. 3). Within this site, three zones were
299 examined more closely due to signs of previous erosion episodes; these zones will be referred to as A
300 (transects 746-827), B (transects (352-409), and C (transects 929-954). The 20-m wide channel is in a
301 post-glacial valley covered by diverse assemblages of deciduous and coniferous trees, shrubs and
302 herbaceous species. The density of mature trees within the riparian area is higher in zone A. In zone B the
303 tree density is noticeably lower and the proportion of shrubs higher, perhaps due to their proximity to an
304 area that is seasonally flooded and occupied by beavers. Zone C includes mature trees between transects
305 932-940, but grass and shrubs in the downstream transects. Bank height along the external bends typically
306 varies between 2-4 m, with two bluffs substantially increasing this value locally to 20 m (Fig. 3). This
307 stream was selected due to the substantial observed erosion along certain banks, but also because its flow
308 has been monitored for decades. Bankfull discharge (1.5-year recurrence interval, based on gauging
309 station 02GD008 just downstream of the study reach) is about 43 m³/s. In addition, sedimentology was
310 examined by N. Bergman (personal communication, 8 December 2013) and rendered available for this
311 project. Basic flow measurements (depth and velocity) were taken along the inlet and outlet cross sections
312 of the study reach at low stage for the purpose of calibrating the flow and the energy slope prior to
313 running numerical simulations. The difference between the predicted increase in free surface elevation at
314 the outlet (1.58 m) from low (1.15 m³/s) to high flow discharge (60 m³/s) was less than 7% compared with

315 data from the gauging station (1.48 m), which is an acceptable error that is not expected to have an effect
 316 on the geotechnical modelling outcome of predicted river bank failures.



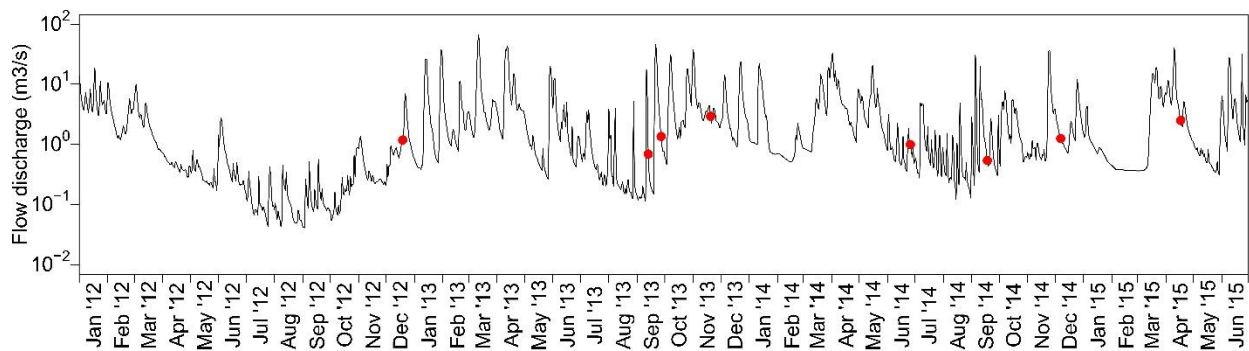
317
 318 **Fig. 3.** Topography of the river and floodplain at Medway Creek, London, Ontario. The outline of
 319 computational elements appears in light grey. Three zones were studied in greater detail: Zone A
 320 (transects 746-827), Zone B (transects 352-409), and the downstream Zone C (transects 929-954). The
 321 two white arrows indicate the position of liquid boundaries and flow direction. Bed form types
 322 correspond to the locations where the substrate was sampled by N. Bergman (personal communication, 9
 323 June 2016). Note that the center bar is partially covered with long grass and shrubs. Inset: Detail on
 324 transects that are used in slope stability assessments for zone A.



325
 326 **Fig. 4.** Photographs of the most unstable river banks along the study reach between January 2012
 327 and June 2015, before and after failure. The corresponding transects in Fig. 3 are A) 760, B) 781, C)
 328 807, and D) 393-397. White ellipses highlight trees fallen due to bank erosion.

329 Substantial efforts were put into surveying and monitoring channel morphology in this study. In
 330 November 2012, we collected over 5000 topography points of the channel bed and banks in the study
 331 reach using a differential GPS (resolution of 1 cm vertically and 1.5 cm horizontally). These points were
 332 combined with a 1-m resolution LiDAR Digital Elevation Model (DEM) (The University of Western
 333 Ontario, 2006) to create a DEM of the channel and floodplain. Photographs of the banks were taken after
 334 each flood with discharge (Q) $\geq 15\text{m}^3/\text{s}$ between February 2012 and April 2015 when the river banks were
 335 visible (e.g. no snow cover, no high flow) (Fig. 4). A consumer-grade camera with geotagging capabilities
 336 (Canon 60D) with wide-angle lens (Canon 10-20 mm) was employed along three banks that appeared
 337 unstable. A Garmin eTrek Legend CX hand-held GPS, mounted on the camera's hot shoe, recorded the
 338 position of each photograph. Photogrammetry analysis was performed on seven photograph sets taken in
 339 zone A (Fig. 3), corresponding to flooding events recorded between September 2013 and April 2015

340 (Fig. 4, 5), in an attempt to estimate the amount of bank retreat in a non-intrusive way since the research
341 site is located in a protected area. Targets were placed on six trees along unstable reaches to facilitate the
342 analysis in the photogrammetric software Agisoft PhotoScan. Two of these trees were subject to bank
343 failures, and were evacuated by subsequent floods (Fig. 4B), which rendered the photogrammetry analysis
344 more tedious and less accurate than expected. Although photographs were taken in zones B and C,
345 photogrammetry was not attempted due to the lack of bank retreat in these zones. The photogrammetry
346 analysis reveals a bank retreat up to 2.4 m in zone A, which is compatible with visual cues (Fig. 4).



347 **Fig. 5.** Flood hydrograph between January 2012 and June 2015, based on gauging station 02GD008
348 located just downstream of the study reach. Red dots indicate the dates where measurements of bank
349 morphology and characterization of bank condition took place.
350

351 Data acquired by N. Bergman (personal communication, 9 June 2016) in the same study reach
352 revealed substantial spatial variations in grain size distribution (GSD) and various bed form types (riffles,
353 pools, bars, flats, steps, and bluffs) (Fig. 3). Trial simulations, using a unique grain size distribution to
354 describe bed substrate in the whole study reach, indicated the importance of varying GSD spatially. With
355 a single curve, pools quickly filled with sediment, leading to a homogeneous longitudinal and lateral
356 bathymetry very different from field observations. In order to simplify our simulations, grain size
357 distribution data were re-organized into three classes using the package 'rPart' (Therneau et al., 2015) in
358 the software R (R Core Team, 2013). The resulting GSD used in the simulations is presented in Table 1.
359 River banks exhibit three distinct soil layers along the study reach. The lower part consists of glacial till,
360 which is buried under an equally thick sand layer, covered itself by a thin organic layer with dense root

361 network. However, in the absence of accurate sedimentological data for our study site, and for simplicity,
 362 we set up our model with uniform textured bank material with a single layer corresponding to the USCS
 363 class ML (silt, given a mean grain size diameter of 0.0234 mm). By doing so, we hypothesize that the
 364 lower layer is more resistant than the middle and upper layers, and thus that the till determines bank
 365 strength and retreat, which is compatible with field observations.

366 **Table 1.** Grain size distribution curves.

Category	Pool		Riffle		Flat		Step		Bar	
	Size (mm)	Fraction (%)	Size (mm)	Fraction (%)	Size (mm)	Fraction (%)	Size (mm)	Fraction (%)	Size (mm)	Fraction (%)
Fine	4.1	35	20.9	27	5.0	20	20.5	17	3.0	28
Medium	66.9	43	96.6	29	57.7	43	98.1	22	63.4	49
Coarse	271.2	22	250.9	44	189.0	37	270.3	61	239.7	23

367

368 3.2. Numerical setup

369 3.2.1. Mesh generation

370 The size of the triangular elements of the numerical mesh varied spatially to use a higher resolution in
 371 areas most likely affected by mass wasting (along steep river banks in bends) and on the channel bed to
 372 ensure that the simulated flow is grid independent. A moderate resolution was selected for the floodplain,
 373 and a low resolution for the valley walls. Grid-independence sensitivity analyses were completed to
 374 determine the appropriate number of nodes to include in the simulation domain (Roache et al., 1986; Lane
 375 et al., 2005; Biron et al., 2007). The number of mesh nodes was 67,780 (0.09 m²/element), 18,841 (1.20
 376 m²/element), 17,003 (3.67 m²/element), and 4135 (48.69 m²/element), respectively in the steep-bank, bed,
 377 floodplain and valley side zones. The time step was set to 0.1 s to ensure a Courant number close to unity.

378

379 3.2.2. Calibration and boundary conditions

380 3.2.2.1. Flow

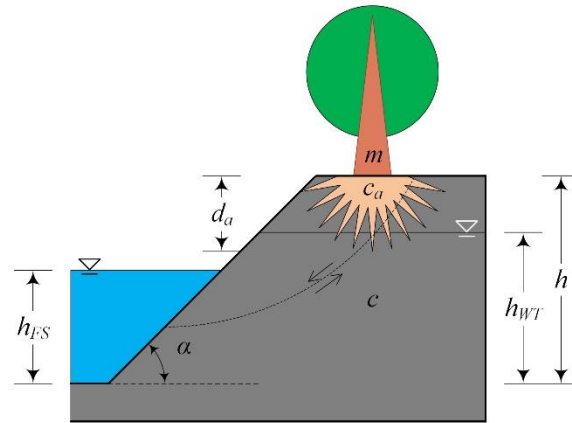
381 A fixed-flat-bed simulation was first run with a low flow discharge of $Q = 1.15 \text{ m}^3/\text{s}$ to adjust the
382 elevation of the water surface at the inlet of the simulation domain to the measured inlet value at Medway
383 Creek. This was done by varying bed roughness coefficient; a Strickler-Manning (n) value of 0.039 was
384 selected for the entire bed. This procedure, which is common in CFD modelling (e.g. Bates et al., 1997;
385 Rameshwaran et al., 2013), allows adjustment of the energy slope to fit experimental measurements
386 (Vidal et al., 2007). At the field site, high-magnitude flooding events generally last a few days. For
387 instance, a $66.2 \text{ m}^3/\text{s}$ peak discharge occurred during a flood event that lasted nearly six days. A time-
388 contracted gamma-distributed hydrograph was therefore fitted to the gauging station data to capture the
389 intensity and shape of this event while limiting simulation length to 12 hours. The gamma equation is
390 given by:

391
$$Q_t = Q_0 + (Q_P - Q_0) \cdot e^m \cdot (t/t_P)^m \cdot e^{-m t/t_P} \quad (12)$$

392 where Q_t = flow discharge at time t , Q_0 = initial discharge, Q_P = peak discharge, m = shape parameter, and
393 t_P = peak time. In all simulations, parameters values were $Q_0 = 7.5 \text{ m}^3/\text{s}$, $Q_P = 60 \text{ m}^3/\text{s}$, $m = 4.8$, and $t_P =$
394 3 h.

395 3.2.2.2. Sediment

396 A mobile-bed simulation with fixed banks and steady bankfull discharge ($Q = 43 \text{ m}^3/\text{s}$) was run to
397 determine the transport equations and substrate properties. The friction angle of the sediment was 36.5° ,
398 bed porosity 0.4, and mass density of sediment taken as $2650 \text{ kg}/\text{m}^3$. This simulation revealed a stable bed
399 with minor adjustments over a 24-hour period. During all simulations, the sediment rate at the inlet was
400 equal to the outlet rate.



401 **Fig. 6.** Simplified bank profile and variables for the evaluation of model behaviour. Sensitivity analysis
 402 was conducted for the following variables: bank angle ($^{\circ}$) and height (h), soil cohesion (c), height of free
 403 surface (h_{FS}) and water table (h_{WT}), trunk mass (m), apparent cohesion due to roots (c_a), and root crown
 404 radius (d_a). The dotted arc represents a hypothetical slip surface.
 405

406 *3.2.2.3. Geotechnical slope stability and vegetation*

407 Calibration was done in three ways. First, a series of bank stability predictions along a river bank with a
 408 simple profile (Fig. 6) was undertaken to explore model behaviour for a range of biophysical condition, to
 409 identify the most sensitive variables, and to estimate parameter values to use in subsequent three-
 410 dimensional simulations. For the evaluations that included vegetation effects, a single Sugar Maple tree
 411 located at the bank top was considered. This tree was assumed to be positioned on the slump block, thus
 412 increasing the weight of a slump block intersecting bank top. Tree length and root radius were calculated
 413 from trunk diameter using empirical equations presented in Kenefic and Nyland (1999) and Tubbs (1977),
 414 respectively for length and radius. Wood density was assumed to be 690 kg/m^3 (Green et al., 2007). The
 415 range of parameter values considered for each sensitivity analysis (geotechnical and geotechnical-
 416 vegetation) covered those found at the field site. Due to the non-linearity of Eq. 3-4 and stochastic
 417 behaviour of the genetic algorithm, results were analyzed using machine learning algorithms in R
 418 (R Core Team, 2013). The importance of variables was quantified using the 'randomForest' package
 419 (Breiman et al., 2015), whereas classification trees were built using the 'rPart' package (Therneau et al.,
 420 2015). In the latter analysis type, tree complexity was selected in such a way that the standard error be

421

Table 2. Parameters of the geotechnical model.

<i>Properties of bank material</i>	
USCS class	ML
Mean grain size (mm)	0.0234
Cohesion (kPa)	0.25
Compaction	75%
<i>Transects</i>	
Length (initial; minimum) (m)	5.0; 1.0
Number of nodes (initial; additional on bank top)	65; 3
Number of mesh nodes and trial angles during orientation	9; 33
<i>Stability assessment</i>	
Minimum block width (% of transect length)	75.0
Minimum block profile area (m ²)	1·10 ⁻²
Number of vertical block slices	2 ⁵
Safety factor precision; mass balance precision (m ³)	1·10 ⁻⁴
Mass balance precision (m ³)	1·10 ⁻⁴
<i>Genetics</i>	
Population size	48
Number of generations (minimum; maximum)	16; 32
Number of generations without improvement (maximum)	4
Mutation rate (%)	12.5
Migration rate (%)	65.0
Options: Inbreeding; polygamy	no; no

422

423 smaller than the error difference between consecutive levels of complexity. Second, a series of one-
 424 iteration mobile bed-and-banks simulations was completed to help determine the value of key
 425 geotechnical parameters: cohesion, USCS class, mean grain size and compaction. The objective was to
 426 ensure that no mass failure occurs for a set of parameters representing the (low) flow conditions
 427 encountered during the week when the initial channel survey was completed. Bank stability was found to
 428 be sensitive to the four parameters with the exception of compaction. The parameter values that produced
 429 the fewest failures were selected (Table 2). Third, a series of 2:45 hour-simulations (corresponding to the
 430 peak values of the gamma-distributed curve representing the hydrograph), were run to further examine the
 431 model's sensitivity to soil cohesion. Although the geotechnical module has the capability to calculate soil

432 **Table 3.** Parameters varied during numerical simulations. Trunk spacing is equal to twice rooting depth
 433 d_a .

Configuration	Vegetation parameters					
	a_{TREE}	d_{BASE}	h_{TREE}	m_{TRUNK}	d_a	c_a
HYD	-	-	-	-	-	-
SED	-	-	-	-	-	-
GTC	-	-	-	-	-	-
RVG _{SMALL}	12.5	8.7	12.1	25	1.00	0.025
RVG _{MEDIUM}	23.2	20.9	21.1	250	1.25	0.250
RVG _{LARGE}	50.3	55.7	29.8	2500	1.75	1.250

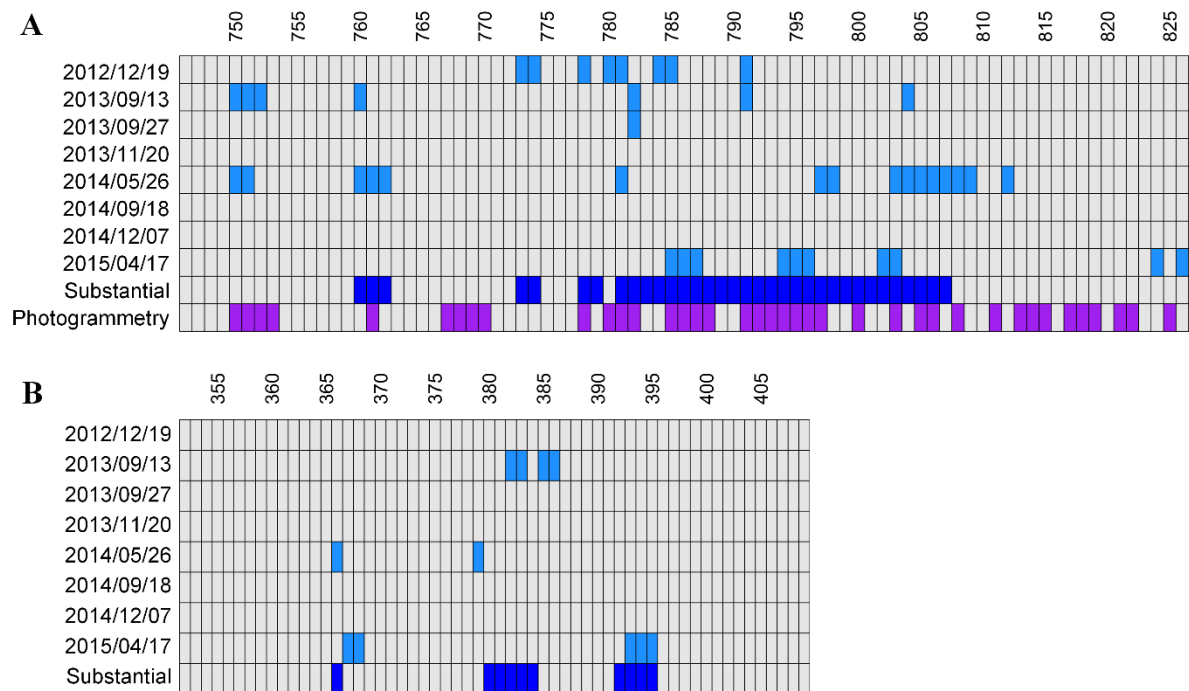
Legend: HYD = a hydraulic-only simulation; SED = a HYD simulation with sediment transport; GTC = a SED simulation with mass wasting; RVG = a GTC simulation with a single tree type of size specified in the subscript; a_{TREE} = tree age (years); d_{BASE} = basal trunk diameter (cm); h_{TREE} = tree height (m); m_{TRUNK} = trunk mass (kg); d_a = rooting depth (m); and c_a = apparent cohesion due to roots (kPa).

434 cohesion based on USCS class and degree of saturation, we note a substantially discrepancy between the
 435 range of bank cohesion values typically encountered in nature (9-67 kPa for USCS class ML) and those
 436 employed while calibrating the geotechnical module (0.25-1.50 kPa). Thus, this variable should be
 437 considered as a numerical parameter rather than an input variable, in a similar way to the roughness
 438 coefficient for hydrodynamic calibration. The high degree of compaction was assumed to match the
 439 geological past of this post-glacial environment. Regarding bank hydrology, water table elevation
 440 adjustment rates, k , of 0.056 and 0.018 were used, respectively for the rising and falling limbs of the
 441 hydrograph. These values were obtained by fitting Eq. 7 to data acquired by Needelman (2013) in a river
 442 bank of similar composition. The values assigned to key parameters of the genetic algorithm are provided
 443 in Table 2. These parameters were adjusted to minimize computation time whilst maximizing precision.

444 3.3. Simulations and analysis procedure

445 Excluding the calibration simulations, a total of six simulations were run with the biophysical parameters
 446 described listed in Tables 2-3. Physiological plant properties were varied in a way such as to provide a
 447 range of stabilization power; threshold parameter values were estimated using machine learning
 448 algorithms applied to the output of geotechnical slope stability assessments along a simplified river bank

449 profile. A high value of apparent cohesion (c_a) was attributed to a large, 50-year old Sugar Maple tree,
 450 whilst the value assigned to small and medium trees was proportional to the basal trunk area of the large
 451 tree.



452 **Fig. 7.** Occurrence of river bank failures along zones A and B based on photographs. A light blue box
 453 corresponds to an observed geotechnical failure for a specific date and transect. Dark blue boxes
 454 correspond to substantial total retreat ($\geq 0.1 \text{ m}^2$) during the observation period, i.e. between 2012/12/19
 455 and 2015/04/17, and include both fluvial and geotechnical processes that led to the removal of bank
 456 material along a transect. Purple boxes indicate transects where retreat was detected by photogrammetry
 457 for the same period (provided for zone A only).
 458

459 4. Results

460 4.1. Field observations

461 A total of 25 flooding events with a magnitude of $Q \sim 15 \text{ m}^3/\text{s}$ or above were recorded during the study
 462 period (2012-2015) (see Fig. 5). The largest number of bank failures was noticed on May 26th 2014,
 463 following a winter that included 6 floods with peak discharge varying between 14.4-33.0 m^3/s , and after
 464 receiving 45.3 mm of rain, and April 17th 2015 (associated with 18.9 and 40.3 m^3/s floods). Zone A was
 465 subject to a larger number of failures (42) than zone B (11) (Fig. 7), whereas a single minor failure
 466 occurred along transect 953 of zone C on November 20th, 2013 (not shown). Photogrammetry analysis

467 performed for zone A of the study reach reveals a bank retreat up to 2.4 m in the eroded areas (purple
468 column in Fig. 7). Assuming that bank evolution occurred where calculated retreat rates were ≥ 10 cm,
469 the accuracy of photogrammetric measurements, relative to field observations, was 61%; this low value is
470 due to a low number of benchmarks. In addition, riparian vegetation is responsible for false positives
471 downstream of transect 812. The accuracy climbs at 76% when excluding these transects. Nevertheless,
472 the magnitude of retreat rates is comparable to visual observations made on the field.

473 4.2. Accuracy quantification and evaluation

474 In order to facilitate the comparison of numerical simulations with morphological changes observed at the
475 field site, the accuracy of each prediction was quantified using Youden's (1950) index:

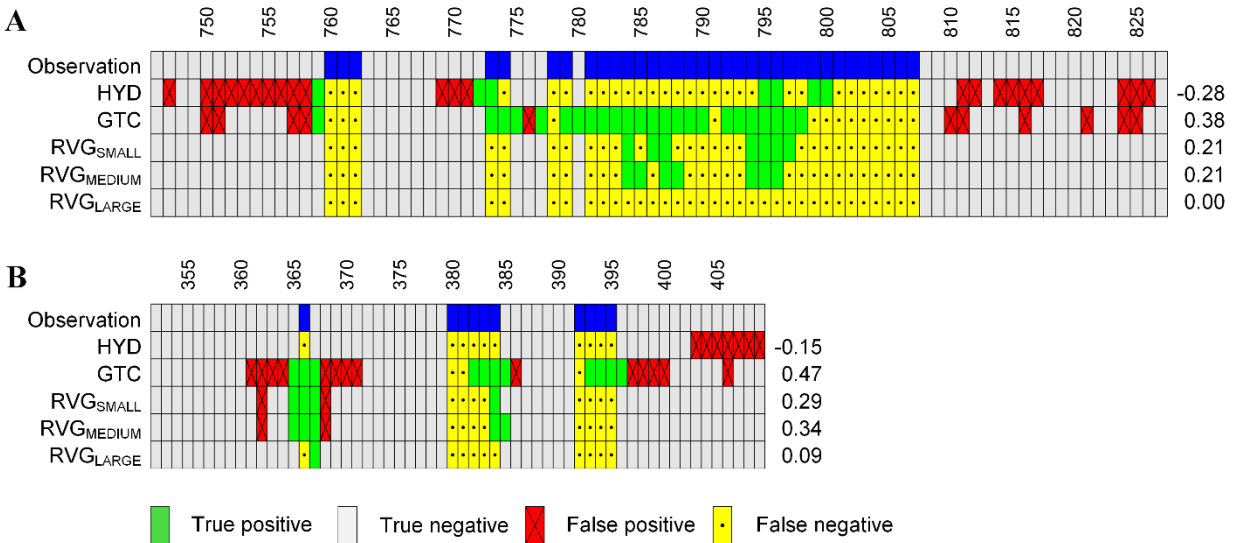
$$476 \quad J = SN + SP - 1 = \frac{TP}{TP + FN} + \frac{TN}{TN + FP} - 1 \quad (13)$$

477 where SN = sensitivity, SP = specificity, TP = number of true positive, TN = number of true negatives,
478 FP = number of false positives, and FN = number of false negatives. This rating method provides a
479 constant range of indices, between 1.0 (correct prediction for every transect) and -1.0 (wrong prediction
480 for every transect), which facilitates the comparison of predictive accuracy amongst configurations and
481 river reaches. In our analysis, an error of one transect spacing (~ 3.5 m) is tolerated, which means that a
482 prediction contributes to TP if a failure was observed along a given transect or along its directly
483 neighbouring transects.

484 4.3. Effects of model components

485 4.3.1. Hydraulics

486 Given the scoring system described above, using maximum bed shear stress along a channel cross-section
487 as an indicator of bank evolution is not very successful. Indeed, using a threshold value of 25 Pa results in
488 Youden's indices of -0.28 and -0.15, respectively for zones A and B (rows "HYD" in Fig. 8). This
489 threshold value was selected because it maximizes the overall accuracy of hydraulic simulations. Very
490 few failures are correctly predicted (true positives) and several are incorrectly predicted (false positives).



491
 492 **Fig. 8.** Comparison of bank failures observed within the study reach between January 2012 and June 2015
 493 to numerical predictions. The x-axis labels correspond to the transects shown in Fig. 3 for zones A and B.
 494 Zone C is not shown due to the lack of substantial bank retreat through mass wasting during the
 495 observation period. The values on the right side of each table correspond to the Youden's index attributed
 496 to each numerical prediction, calculated using Eq. 13. The "observation" row corresponds to the
 497 "substantial" row in Fig. 7. HYD = a hydraulic-only simulation; GTC = a simulation with sediment
 498 transport and mass wasting; RVG = a GTC simulation with a single tree type of size specified in the
 499 subscript. The subscripts SMALL, MEDIUM and LARGE refer to tree size (see Table 3 for the
 500 physiological properties of plant cover).

501 *4.3.2. Sediment transport*

502 The result of the simulation with bedload transport only, i.e. with the geotechnical and vegetation
 503 modules disabled, reveals a fairly stable channel with changes located primarily along channel margins.
 504 Nevertheless, looking at the total volume of displaced material, we note an initial adjustment of bed
 505 morphology during the first iteration of the simulation followed by a slow evolution rate that is
 506 proportional to flow discharge during the rising and falling limbs of the hydrograph. Insignificant
 507 bathymetric changes, combined with rapid channel stabilization, indicate that the coupled
 508 morphodynamic model Telemac2D-Sisyphe, i.e. without geotechnical algorithms, is unlikely to simulate
 509 lateral retreat in our study reach.

510

511 4.3.3. *Geotechnical processes*

512 Simulations with different parameter sets were run to evaluate the sensitivity of the coupled geotechnical
513 model to key parameters. For instance, doubling the adjustment rate of the water table elevation results in
514 very similar erosion patterns. However, a rise in water table elevation within the river bank is required to
515 create mass wasting in the upper bluff area. Although a steady low flow discharge of 7.5 m³/s is sufficient
516 to trigger the retreat of low banks, a peak discharge of 60 m³/s, i.e. similar to the maximum value
517 recorded during the observation period (66.2 m³/s on March 12th, 2013), when combined with a gamma-
518 shaped hydrograph curve, improves fit with observations. In particular, a flow discharge of 60 m³/s
519 affects the tall bluff (transects 746-762 in Fig. 3) and increases the length of the eroded bank subject to
520 mass wasting (transects 778-807 in Fig. 8a). Similarly, soil cohesion values of 0.25, 0.50, and 1.00 kPa
521 were evaluated, with a value of 0.25 kPa resulting in the best match with field observations.

522 Enabling the geotechnical module results in much improved predictions of retreat location
523 compared to using a threshold bed shear stress value. This reflects in the calculated accuracy values, i.e.
524 Youden's indices of 0.38 (GTC) vs. -0.28 (HYD) in zone A, and 0.47 vs. -0.15 in zone B (Fig. 8B).
525 Despite the presence of several false positives, configuration GTC correctly predicts the location of the
526 river banks where acute erosion was observed in the field (row GTC in Fig. 8). Two of the most unstable
527 bank regions along zone A are correctly identified (Fig. 4A-B). The model may not be able to identify
528 transects 799-807 (Fig. 4C and 8A) because most failures along this bank occurred only once the
529 upstream 50-meter region had retreated substantially, i.e. after 18 floods with a magnitude of at least 15
530 m³/s over a period of 1.5 year (Fig. 7). Although configuration GTC results in the largest number of true
531 positives, it also involves several false positives between transects 810-825 (Fig. 8A). This may be
532 explained by the fact that the presence of riparian vegetation is only indirectly, partially accounted for
533 with an adjustment of soil cohesion in configuration GTC.

534

535 4.3.4. Riparian vegetation

536 The vegetation module developed for this study influences the degree of bank stability by increasing soil
 537 cohesion (up to a certain depth under soil surface, depending on plant physiology), and by increasing the
 538 mass of a slump block (tree species only). These mechanical actions in turn affect friction forces along the
 539 slip surface (e.g. dashed line in Fig. 1A), stabilizing or destabilizing a river bank, depending on bank
 540 geometry and local biophysical conditions.

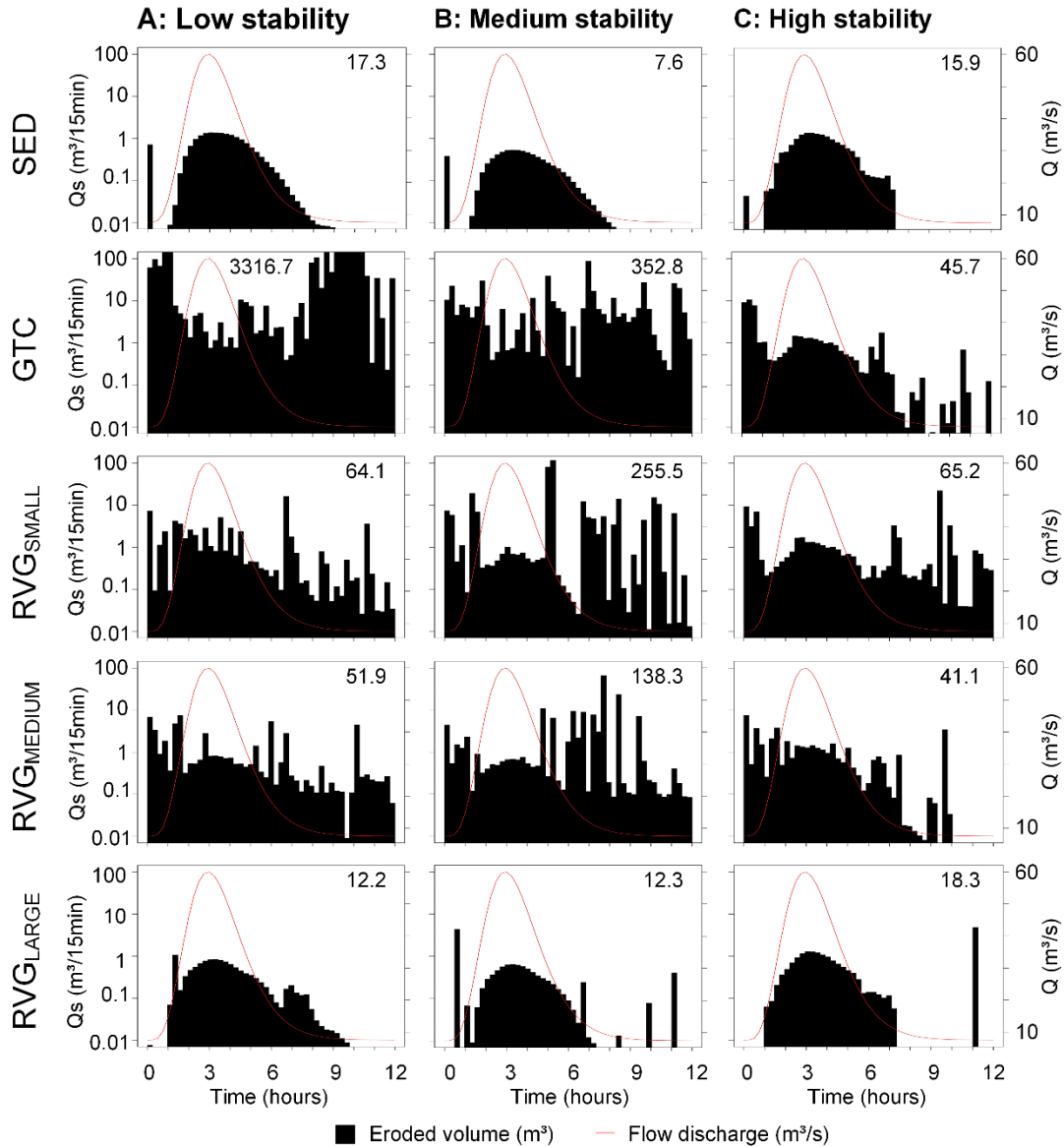
541 All three configurations involving vegetation, i.e. RVG_{SMALL} , RVG_{MEDIUM} and RVG_{LARGE} , induce a
 542 drastic reduction in the number of failures (both false positives and true positives), relative to the basic
 543 geotechnical scenario (row GTC in Fig. 8). Configurations with small and medium trees result in
 544 Youden's indices lower than that of the GTC configuration in zone A. Although this difference is less
 545 important in zone B, configurations RVG_{SMALL} and RVG_{MEDIUM} fail to predict the correct extent of the
 546 large unstable zone (transects 778-807 in Fig. 8A), and the occurrence of failures in top of the bluff
 547 (Fig. 4A and transects 760-762 in Fig. 8A). In addition, the largest failure (in volume) recorded in zone B
 548 during the observation period is not predicted by the model (transects 392-395 in Fig. 8B). This may be
 549 due to the increased stability in the model which was using trees, whereas at the field site for this zone
 550 shrubs are present. The scenario with the largest trees almost completely eliminated bank instabilities,
 551 with a progression in mechanical reinforcement with an increase in tree size. Although vegetation

552 **Table 4.** Number of failures and absolute volume change in zones A, B and C of the study reach. Zone A
 553 corresponds to transects 746-827, zone B to transects 352-409, and zone C to transects 929-954 (Fig. 3).

Configuration	Number of failures			Volume change (m ³)		
	A	B	C	A	B	C
SED	-	-	-	17.3	7.6	15.9
GTC	590	199	108	3316.7	352.8	45.7
RVG_{SMALL}	199	100	166	64.1	255.5	65.2
RVG_{MEDIUM}	108	94	151	51.9	138.3	41.1
RVG_{LARGE}	0	6	1	12.2	12.3	18.3

Legend: SED = a hydraulic simulation with sediment transport; GTC = a SED simulation with mass wasting; RVG = a GTC simulation with a single tree type of size specified by the subscript.

554 considerably reduces the number of failures and volume of displaced floodplain material, it does not
 555 necessarily reduce the mean failure volume (Table 4). For instance, 151 failures produced a total
 556 displacement of 41.1 m³ in zone C under the RVG_{MEDIUM} configuration, whereas a single failure displaced
 557 18.3 m³ under RVG_{LARGE}. Therefore, vegetation cover triggers a non-linear response that seems to be
 558 exacerbated by complex river bank geometry, combined with the stochastic nature of the genetic

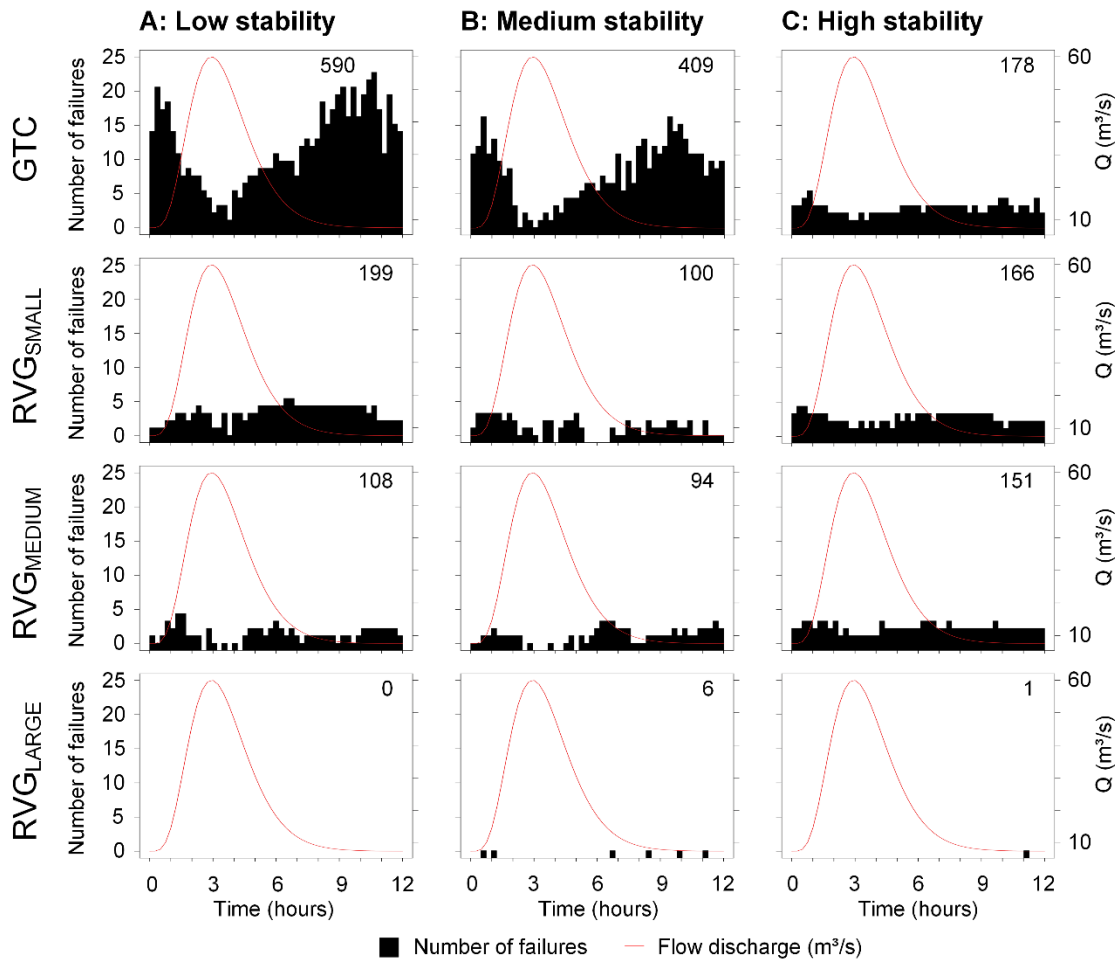


559
 560 **Fig. 9.** Floodplain material displacement (Q_s on the primary y-axis) and flow discharge (Q on the
 561 secondary y-axis) during numerical simulations as a function of simulation time. Columns correspond to
 562 zones A (low bank stability), B (medium bank stability) and C (high bank stability) in Fig. 3. In-plot
 563 annotations are the cumulative, simulated displaced volumes (m³).

564 algorithm. Non-linearity is expected to increase even more with spatiotemporal variations in
 565 sedimentology, and plant physiology and assemblage.

566 *4.4. Timing and magnitude of bank retreat*

567 Neglecting soil displacement occurring at the onset of simulations (which may be attributed to bed
 568 adjustment with respect to flow dynamics), the SED configuration produces displacement rates matching
 569 the shape and phase of the gamma-distributed hydrograph, with maximum values reaching 1.3, 0.6 and
 570 1.4 m³ per 15-minute period, respectively for zones A, B and C (Fig. 9). Activating the geotechnical
 571 module alters the timing of mass wasting and total volume of displaced soil material, and produces a



572 **Fig. 10.** Number of failures (on the primary y-axis) and flow discharge (Q on the secondary y-axis)
 573 during numerical simulations as a function of simulation time. Columns correspond to zones A (low bank
 574 stability), B (medium bank stability) and C (high bank stability) in Fig. 3. In-plot annotations are the total
 575 number of simulated failures.
 576

577 complex signal emerging from the overlap of bed load entrainment and mass wasting processes. Although
578 we note similar patterns in the geomorphic response to the 12-hour flood event amongst zones and
579 amongst configurations (Fig. 9), displacement occurs primarily near the end of the falling limb, i.e. when
580 flow discharge has returned to pre-flood magnitude. The lag between flow and soil displacement peaks
581 seems to be caused by the indirect consideration of soil hydraulic conductivity in the geotechnical
582 modules, which is done by defining water table convergence coefficients. The similarity in response
583 between zones A and B (Fig. 9-10) may be attributable to similarities in bank geometry, height and cover
584 along unstable transects. In particular, zone C has a lower tree density than zones A and B, and does not
585 exhibit three distinctive soil strata. Finally, note that a substantial reduction in the number of failures does
586 not automatically translate into a substantial reduction in the volume of displaced soil material (e.g. Fig. 9
587 vs. Fig. 10).

588 The volume of displaced soil material is inversely proportional to the maturity of the vegetation
589 cover, which demonstrates the substantial contribution of physiological plant properties on channel
590 dynamics (Fig. 9). Small and medium trees reduce the incidence of mass wasting in zones A and B, but
591 have very little effect in zone C (Fig. 10). Note that a patch of young trees provides enough soil
592 reinforcement to substantially reduce the number of failures. The timing of displaced material in
593 configuration RVG_{LARGE} is similar to the timing in SED, with the distinction that vegetation slightly
594 reduces soil displacement rate (Fig. 9). This may be attributed to the increase in friction caused by the
595 presence of riparian vegetation on river banks, which highlights the importance of including flow-
596 vegetation interaction.

597 **5. Discussion**

598 This research confirms that a morphodynamic model lacking a dedicated geotechnical component is
599 unable to simulate lateral adjustments in a meandering channel with cohesive banks. In our simulations,
600 the sediment transport module alone predicts little bank retreat for a study reach that was observed to be

601 unstable. In addition, using a threshold bed shear stress value appears to be a weak predictor for the
602 location of bank failures (row HYD in Fig. 8).

603 Lateral erosion processes have been included into fluvial models using various approaches, and it is
604 not straightforward to compare them directly with the modules presented here, particularly since very few
605 have attempted to simulate complex natural cases (e.g. Van de Wiel and Darby, 2004). Bishop's (1955)
606 method of slices, used here to quantify river bank stability, was also employed in other fluvial studies, but
607 often with the intent of examining the geotechnical stability at the river bank and reach scales, not
608 connecting it to a hydraulic model (e.g. Van de Wiel and Darby, 2007; Langendoen and Simon, 2008;
609 Midgley et al., 2012). A few models consider geotechnical soil properties in the calculation of retreat
610 rates, but these are generally kept constant during a simulation. For instance, the incipient collapse angle
611 is not affected by flow stage in Asahi et al. (2013).

612 A major advantage of our model is the integration of a river bank hydrology module and the
613 consideration of spatial variations in soil moisture within the bank during stability evaluations.
614 Furthermore, it allows the complexity of natural rivers to be fully represented, whereas most other models
615 impose idealized bank profiles (e.g. Eke et al., 2014; Langendoen et al., 2016) and planform geometries
616 (e.g. Crosato, 2007). Several of these implementations assume (without enforcing) mass conservation
617 during bank retreat, and by doing so, ignore the role of basal control on river bank stability (Thorne,
618 1982). There are, however, some existing retreat models that incorporate a deposition algorithm (e.g. Van
619 de Wiel and Darby, 2004; Dulal et al., 2010).

620 A major advantage of the proposed model is its genetic algorithm, which selects the failure
621 mechanism (rotational or translational) as it determines the extent of the slump block. This feature is
622 particularly relevant to the study of meandering river channels, which develop in cohesive floodplains
623 (Anderson and Anderson, 2010), as rotational failures tend to be associated with cohesive soils (Thorne,
624 1982). Conversely, most existing bank retreat formulations assume non-cohesive soils, and therefore
625 translational failures. The consideration of biophysical conditions (e.g. soil composition and moisture,

626 flow stage, water table elevation, plant cover) by the genetic algorithm is also compatible with the
627 episodic nature of bank retreat in natural rivers (Abernethy and Rutherford, 1998; Pollen-Bankhead and
628 Simon, 2010). Therefore, it is not clear whether existing morphodynamic models, often generating
629 geometrically 'idealized' meandering planforms, can be applied to complex river environments.

630 Bank retreat along Medway Creek was discontinuous in time, space and magnitude during the
631 observation period (Fig. 7 and 8). These irregularities are fairly well simulated by our model, which
632 seems to suggest the necessity to include a sophisticated geotechnical algorithm in morphodynamic
633 models for improved applicability to natural rivers. The proposed modelling approach, which is more
634 detailed and physics-based than previous implementations in the description of salient processes,
635 therefore reduces the gap between theory and practical applications for complex river environments by
636 integrating vector-based geospatial treatment, not only in data surveying as Güneralp and Marston (2012)
637 suggest, but at the heart of the modelling algorithm.

638 The integration of physics-based stability equations also facilitates the fragmentation of river bank
639 erosion into distinct processes (e.g. fluvial and geotechnical), forces (e.g. hydrostatic pressure, pore-water
640 pressure, slump block weight) and components (e.g. riparian plants). Fragmentation, which is one of the
641 key novelties of the proposed morphodynamic modelling approach, enables to examine causal
642 relationships, with parameter values defined in terms of distinct, ideally measurable, physical quantities.
643 Existing planform meander models, including the HIPS formulation (Hasegawa, 1977; Ikeda et al., 1981),
644 simulate lateral retreat as a function of excess near-bank shear stress or excess velocity (e.g. Güneralp et
645 al., 2012; Zolezzi et al., 2012), which prevents the direct manipulation of variables influencing lateral
646 erosion.

647 A similar observation can be made regarding the inclusion of plant effects. Most existing models
648 indirectly consider vegetation (but see Van de Wiel and Darby (2004, 2007) for a simplified physics-
649 based approach) by increasing the threshold stream power required to initiate sediment transport (e.g.
650 Murray and Paola, 2003) or by altering bed roughness based on plant physiology and arrangement, which

651 reduces near-bed velocities (e.g. [Crosato and Saleh, 2011](#)), or by attributing an added cohesion value due
652 to presence of roots while ignoring surcharge effect (e.g. [Eaton and Giles, 2009](#)). Fuzzy concepts have
653 also been used to lump, quantify and integrate the stabilization effects of riparian plants. Modifying bank
654 erodibility based on biomass density (e.g. [Camporeale et al., 2013](#)) allows to relate planform patterns to
655 biomass density and vegetation growth rate ([Perucca et al., 2007](#)). However, this numerical, immaterial,
656 and partially subjective conceptualization of floodplain vegetation limits the application of these models
657 to real-world investigations as biomass does not unambiguously translate into a set of measurable
658 physiological plant properties or alterations of soil strength.

659 By fragmenting geotechnical and vegetation processes and forces, the model presented in this paper
660 facilitates the identification of causes of bank failure. However, the current version of the model fails to
661 recognize the normal range of cohesion values associated with soil and roots in nature, which indicates
662 that it is not yet capable of accounting for the overall complexity of the natural environment, although the
663 overall approach constitutes a step in this direction.

664 Describing fluvial processes using physically-correct equations considerably increases run time,
665 and therefore, severely limits the spatiotemporal scales that can be investigated. The inclusion of the non-
666 linear version of the shallow-water equations and quantification of geotechnical stability based on a limit
667 equilibrium method, combined with the selection of appropriate cell size and time step to ensure
668 numerical stability and the validity of predictions (e.g. grid-independence testing using [Biron et al.](#)
669 [\(2007\)](#); Courant number below unity), contribute to this situation. Despite the activation of parallel
670 processing and the use of a genetic solver for the geotechnical computations, each simulation took
671 approximately 4.5 days to run on a computer equipped with double hex-core processors. As an indication,
672 the coupled model spent 18.0%, 69.5%, and 12.5% of its time in the hydraulic, sediment, and
673 geotechnical modules, respectively. Thus, further improvements in algorithmic efficiency are needed for
674 river scientists and practitioners to employ the type of model presented here to perform simulations over
675 long spatiotemporal scales.

676 In this study, and for the sake of efficiency, we explored parameter sensitivity using machine
677 learning algorithms on a series of one-iteration simulation results, and identified threshold values for key
678 parameters. Such short 'trial' simulations helped identifying the range of reasonable parameter values, but
679 did not allow for the interaction between processes and the full spectrum of combinations of biophysical
680 conditions to be taken into account. This may explain why activating the vegetation module enhances
681 stability more drastically than expected, leading to less accurate predictions than a strictly geotechnical
682 simulation. The outcome may have differed with a sophisticated calibration/fitting method capable of
683 determining soil and apparent (due to roots) cohesion values maximizing correlation between observed
684 and predicted failures.

685 The proposed bank retreat approach differs from previous implementations in its enhanced
686 applicability to complex natural river environments, as most geotechnical models simulating river banks
687 failures are limited in scale to a few tens of meters (e.g. [Thomas and Pollen-Bankhead, 2010](#); [Midgley et al., 2012](#)), whilst morphodynamic models mainly focus on flume-size channels with non-cohesive banks
688 (e.g. [Langendoen et al., 2016](#)) and assume non-vegetated floodplains (e.g. [Pittaluga and Seminara, 2011](#);
689 [Asahi et al., 2013](#)). Therefore, with a few exceptions such as the model developed by [Lai et al. \(2015\)](#), it
690 is one of the very few models that can accurately predict bank erosion episodes at a scale relevant to the
691 management and restoration of river reaches. Validating the meandering planform properties and
692 dynamics predicted by pseudo-empirical approaches is often limited to a coarse comparison of
693 behavioural evolutionary traits against those visible in historical aerial photographs (e.g. [Camporeale et al., 2005](#);
694 [Perucca et al., 2007](#); [Duan and Julian, 2010](#)) or against datasets from analogue flume
695 experiments (e.g. [Duan et al., 2001](#); [Rüther and Olsen, 2007](#)). Conversely, the fairly subtle planform
696 changes associated with physics-based models are not necessarily detectable on airborne imagery.
697 Although a sophisticated, physics-based model applicable to natural vegetated rivers provides more
698 detailed mechanistic information related to river bank erosion processes, it also requires exhaustive data
699 measurements for both parameterization and validation, which are seldom available.
700

701 **6. Conclusion**

702 New geotechnical and vegetation modules were developed and coupled to the hydrodynamic module
703 Telemac-2D and sediment transport module Sisyphé to create a universal and more physics-based
704 representation of the evolution of alluvial river channels at the kilometer scale. The agreement between
705 observed and predicted river bank failures in our simulations is encouraging, and comparison with field
706 data revealed a marked improvement in bank failure predictions when adding the geotechnical module,
707 albeit with limited gain in accuracy when adding the vegetation module.

708 The main novelty of the developed modules is the possibility to parameterize lateral erosion using
709 physical, measurable quantities such as geotechnical soil properties (texture, cohesion, compaction and
710 porosity) and physiological plant properties (wood density, trunk diameter and length, and root strength
711 and depth). Importantly, and innovatively, the geotechnical evaluations are done independently of the
712 hydrodynamic mesh, which allows single- or multi-threaded alluvial river types of any scale (flume,
713 stream, and river) to be studied for cohesive and non-cohesive soils, with or without aquatic or terrestrial
714 plants. The consideration of a floodplain rather than strictly a channel allows to consider palaeochannels
715 and to include the effects of mass wasting events occurring away from the channel, for instance, along
716 valley walls. Furthermore, the effect of soil water content, in particular the effects of the imbalance
717 between free surface and water table elevations during a flooding event, is an important addition to the
718 model since it is recognized as critical to river bank stability. Finally, the use of a fully configurable
719 genetic algorithm with tournament selection keeps runtimes close to those of a morphodynamic model
720 lacking a lateral erosion algorithm while allowing to efficiently locate planar, circular and non-circular
721 slip surfaces through a single algorithm. Our solution addresses some of the critiques of existing
722 morphodynamic models such as the lack of physics in lateral erosion algorithms, the omission of mass
723 wasting and vegetation processes, and their incapacity to consider multi-threaded channels.

724 To enhance the applicability of the developed model for river-related management issues, future
725 developments should focus on improving the physical representation of vegetation, in particular, with

726 respect to the cohabitation of multiple individuals and species in a single simulation cell, a situation that
727 is usually encountered in a natural setting. An adaptive mesh, such as the one built by [Langendoen et al.](#)
728 [\(2016\)](#), would also likely be more efficient for physics-based 2D and 3D models. Most importantly, there
729 is a serious need for accessible time series of biophysical, hydraulic, topographic and sedimentological
730 datasets to calibrate and validate models such as ours. As highlighted by [Rousseau et al. \(2016\)](#), creating a
731 central repository providing universal validation cases for morphodynamic modelling would be extremely
732 valuable.

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