		Mis enforme: Anglais(ÉtatsUnis)
	Estimating the disequilibrium in denudation rates due to divide	
	migration at the scale of river basins.	
	Timothée Sassolas-Serrayet ¹ , Rodolphe Cattin ¹ , Matthieu Ferry ¹ , Vincent Godard ² , Martine Simoes ³	 Mis en forme: Français(France)
	¹ Géosciences Montpellier, Université de Montpellier and CNRS-UMR 5243, Montpellier-34095, France.	 Mis enforme: Français(France)
5	² Aix Marseille Univ, CNRS, IRD, INRA, Coll France, CEREGE, Aix-en-Provence, France ³ Université de Paris, Institut de physique du globe de Paris, CNRS, F-75005 Paris, France	
	<i>Correspondence to</i> : Timothée Sassolas-Serrayet (timothee.sassolas-serrayet@umontpellier.fr)	 Mis enforme: Anglais(ÉtatsUnis)
	correspondence to: rinoutee bassous benayer (inioutee sussous senayer e uniontee iniontee iniontee bassous benayer (inioutee sussous senayer e uniontee iniontee iniontee bassous senayer e uniontee iniontee bassous senayer e uniontee bassous sena	
	Abstract. Basin-averaged denudation rates may locally exhibit a wide dispersion, even in areas where the topographic	 Mis enforme: Anglais(ÉtatsUnis)
I	steady state is supposedly achieved regionally. This dispersion is often attributed to the accuracy of the data or to some	 Mis en for me: Police :GrasAnglais
10	degree of natural variability of the signal, but it which can also be attributed related to stochastic processes such as	(États Unis)
	landsliding. Another physical explanation to this dispersion is that of local and transient disequilibrium between tectonic	
	forcing and erosion at the scale of catchments. Recent worksstudies have shown that divide migration can potentially induce	
1	such perturbations and they propose reliable metrics to assess divide mobility based on cross-divide contrasts in headwater	
	topographic features. Here, we use a set of landscape evolution models assuming spatially uniform uplift, rock strength and	
15	rainfall to assess the effect of divide mobility on basin-wide denudation rates. We propose the use of basin-averaged	
	aggressivity metrics based on cross-divide contrasts (1) in channel-head χ , an integral function of position in the channel	 Mis enforme: Anglais(ÉtatsUnis)
	network, (2) in channel local slopegradient and in elevation.(3) in channel height, measured at a reference drainage area.	
	From our simulations, we show that the metric based on differences in $\frac{\text{channel-head elevations across divides}}{\chi}$ is the	
	leastmost reliable to diagnose local disequilibrium. For the The other metrics, our results suggest a nonlinear relationship	
20	with are more suitable for relatively active tectonic regions as mountain belts, where contrasts in local gradient and	
	elevation are more important. We find that the ratio of basin denudation associated with drainage migration to uplift, which	
	can reach up to a factor of two, regardless of the imposed uplift rate, erodibility, diffusivity coefficient and critical hillslope	
	gradient. A comparison with field observations in the Great Smoky Mountains (southern Appalachians, USA) underlines the	
	difficulty of using the metric based on $\frac{1}{2}\chi$, which depends on the -poorly constrained elevation of the outlet of the	
25	investigated catchment. Regardless-of- the considered metrics, we show that observed dispersion is controlled by catchment	
1	size: a smaller basin may be more sensitive to divide migration and hence to disequilibrium. Our results thus highlight the	
	relevance of divide stability analysis from digital elevation models as a fundamental preliminary step for basin-wide	
	denudation rate studies based on cosmogenic radionuclide concentrations.	

1 Introduction

- 30 Topographic steady state, in which average topography is constant over time, is one of the key concepts of modern geomorphology (e.g. Gilbert, 1877; Hack 1960; Montgomery, 2001). Though simple, this paradigm provides a useful framework to study landscape evolution related to tectonic and/or climatic forcing (e.g. Willett et al., 2001; Reinhart and Ellis, 2015), to spatial variations in rock strength (Perne et al., 2017) or to the geometry of active crustal structures (Lave and Avouac, 2001; Stolar et al., 2007; Scherler et al., 2014; Le Roux-Mallouf et al., 2015). To define topographic steady state,
- 35 the temporal and spatial scales of the processes involved are essential parameters. Compared to large scale geodynamic processes operating over 1-100 Myr timescales, river incision and sediment transport are rapid processes driving landscapes
 to stable forms over this long timescale, whereas rapid climatic fluctuations during the Quaternary may prevent modern landscapes from reaching the occurrence of steady-state conditions in modern landscapes (Whipple, 2001).

The timescale of river divide migration has received increasing attention in the recent years. Although rivers exhibit a rapid 40 adjustment to tectonic or climatic changes to maintain their profiles, Whipple et al. (2017) show that divides continue to 40 migrate over time periods of 10^{6} - 10^{7} years, suggesting as response to the same changes. This suggests that long-term 41 transience might be pervasive in the planar structure of landscapes<u>, even in the absence of new variations in landscape</u> 42 <u>characteristics or forcings (e.g. tectonic or climate)</u> (Hasbargen and Paola, 2000; Hasbargen and Paola, 2003; Pelletier, 2004<u>Dahlquist et al., 2018</u>). In addition to the influence of spatial variability of rock uplift rate, rock strength or rainfall (e.g.

45 Reiner et al., 2003; Godard et al., 2006; Miller et al., 2013), this long timescale could also explain the persistence of spatial variations in denudation rates observed in tectonically inactive orogens in spite of a supposedly and theoretically topographic achieved steady state is achieved (Willett et al., 2014).

As an example of this, in the Great Smoky Mountains in the southern Appalachians, uplift and erosion rates integrated over varying time periods from 10s kyr to 100 Myr give a similar average magnitude of $\frac{0.025 \cdot 0.030 \text{ mm.yr}^4}{2 \cdot 2 \cdot 0.030 \text{ mm.yr}^4}$

- 50 (Matmon et al., 2003a, b and Portenga and Bierman, 2011). These results suggest a regional quasi-topographic steady state over the last ~180 Myr, maintained by the isostatic response of the thickened crust since the end of the Appalachian orogeny
 (Matmon et al., 2003 a, 2003 a, b). Beyond this average value, individual basin-wide denudation rates exhibit a strong dispersion (up to a factor of two, Fig. 1), which is not related to spatial variation in rainfall or in erodibility of the substrate (Matmon et al., 2003b). In a recent study, Willett et al. (2014) assess divide mobility from the contrast in the channel head
- 55 topographic metric χ, taken here as a proxy for steady-state river profile elevation (Perron and Royden, 2012; Royden and Perron, 2013), and propose an explanation in which a significant part of the observed dispersion in denudation rates could be due to drainage divide migration associated with contrasting erosion rates across divides.

Divide migration is often assessed through the metric χ (Willett et al, 2014). More recently, to characterize divide migrations Forte and Whipple (2018) introduced other metrics, referred as "Gilbert metrics" (Gilbert, 1877), based on the cross-divide

60 contrast in channel head local slopegradient and elevation-in order to characterize divide migrations. This last study indeed focused on cross-divide contrasts in headwater basin shape. Here, we model divide migrations and propose to extend these

Mis enforme: Anglais(ÉtatsUnis)

 approaches by modeling divide migration and by developing new metrics ofto assess divide stability at the scale of the entire
 Mis enforme: Anglais(ÉtatsUnis)

 watershed, which are an expansion of the aggressivity metric initially suggested by Willett et al. (2014). We use these
 Mis enforme: Anglais(ÉtatsUnis)

 65
 numerical landscape evolution models, taking into account both hillslope diffusion and fluvial incision. For the sake of
 simplicity and to avoid the influence of other factors such as topography, lithology, climate or vegetation, we restrict our

 analysis to synthetic orogens with spatially uniform uplift, rock strength and rainfall. After a brief presentation of the used
 landscape evolution model (LEM), we describe the methods developed to assess basin-wide denudation rates and

 70
 investigate transient time and location of morphologic adjustments to divide migrations. We explore the relevance and

 romplementarity of tested relative stability metrics between neighbouringneighboring basins. We then investigate the impact

 of uplift rate_erodibility and hillslope processprocesses on the dynamics of divide migration and associated denudation rates.

Finally, we apply our approach to the basin-wide denudation rates dataset of Matmon (2003a,b) in the case of the Great Smoky Mountains and propose new criteria to guide future sampling strategies to assess basin-wide denudation rates from 75 river sands.

2 Methods

2.1 Landscape Evolution Model (LEM)

We use TTLEM (TopoToolbox Landscape Evolution Model) (Campforts et al., 2017), a landscape evolution model based on the Matlab function library TopoToolbox 2 (Schwanghart and Scherler, 2014). This LEM uses a finite volume method (Campforts and Govers, 2015) to solve the following equation of mass conservation for rock/regolith subject to uplift and denudation:

$$\frac{\partial z}{\partial t} = \left(\frac{\partial z}{\partial t}\right)_{td} + \frac{\left(\frac{-U + \left(\frac{\partial z}{\partial t}\right)_{fluv}}{\int fluv} - \frac{for A < A_{\varepsilon}}{\int \rho_{\varepsilon}} \frac{\rho_{r}}{\rho_{\varepsilon}}U + \left(\frac{\partial z}{\partial t}\right)_{hill} + \left(\frac{\partial z}{\partial t}\right)_{fluv}}{\int fluv},$$
(1)

85

80

where $\partial z/\partial t$ is the variation of elevation with time, $(\partial z/\partial t)_{td}$ is the change of elevation due to tectonic horizontal advection, U is the rock uplift rate-, ρ_r/ρ_s is the density ratio between the bedrock and the regolith, A is the upstream area and A_{z} is. We use a linear formulation of hillslope diffusion (Culling, 1963) limited by a critical drainage area which corresponds to the transition between hillslope and fluvial processes. slope S_c :

90 Hillslope denudation is given by a non-linear formulation (Roering et al., 1999):

Mis en forme: Anglais(ÉtatsUnis) Mis en forme: Anglais(ÉtatsUnis)

$$\left(\frac{\partial z}{\partial t}\right)_{hill} = -\nabla q_s \qquad \text{with} \qquad q_s = -\frac{\frac{D\nabla z}{ds}}{\frac{1-\left(\frac{d\nabla z}{ds}\right)^2}{s_z}} D\nabla z,$$
(2)

- 95 where g_s is the flux of soil-regolith material. This flux rate increases to infinity when slope tends to a critical value S_c . When slopes values exceed S_c , they are readjusted to the critical value by using a modified version of the excess topography algorithm (Blöthe et al., 2015). The diffusivity D gives the rate of soil-regolith material creep. Its magnitude ranges from 10^{-3} to 10^{-1} m².yr⁻¹ in natural settings and varies with soil thickness, lithology and vegetation (Roering et al., 1999; Jungers et al., 2009; West et al., 2013; Richardson et al., 2019). Hillslope diffusion is implemented in TTLEM using an implicit scheme,
- 100 which is unconditionally stable at large time steps (Perron, 2011).

Pelletier, 2008). Non-linear diffusion formulation (Perron, 2011) is also implemented in TTLEM. However, we favored the use of a linear diffusion with a critical slope, which is more convenient for the time step used in our simulations (=5000 yr) and the set of parameters considered (see section 2.2). Due to the relatively coarse spatial resolution of our models (= 90m), any of these diffusion formulations generate negligible topographic differences on the direct vicinity of crest lines (Roering

105 et al., 1999. Campforts et al., 2017) and do not affect our results (see Fig. S1). Fluvial incision is calculated with a stream power law:

$$\left(\frac{\partial z}{\partial t}\right)_{f \, l u \, \nu} = -K A^m \left(\frac{\partial z}{\partial x_\Gamma}\right)^n \,, \tag{3}$$

110 where K is the erodibility coefficient reflecting climate, hydraulic roughness, sediment load and lithology. Its value ranges between 10^{16} and 10^{0} m^(1-2m), yr⁻¹ (Kirby and Whipple, 2001; Harel et al., 2016). A is the upstream area. x_{Γ} is the along stream distance from the outlet of the river. m and n are two parameters which that are usually reported as a m/n ratio ranging between 0.35 and 0.8. River The river incision law is implemented in TTLEM using an explicit scheme based on a higher-order flux-limiting finite volume method that is total variation diminishing (TVD-FVM) [see Campforts and Govers 115] (2015) and Campforts et al. (2017) for further details]. Its main advantage is to eliminate numerical diffusion, which is

present in most other schemes solving differential equations of river incision. This last point has a significant impact on the accuracy of basin-wide simulated denudation rates, making TTLEM a well-suited LEM for the purpose of this study.

Mis en forme: Anglais(ÉtatsU	nis)
Mis enforme: Exposant	
Mis en forme: Anglais(États U	nis)
Mis en forme: Anglais(ÉtatsU	nis)

Mis en forme: Anglais(ÉtatsUnis)

Mis enforme: Anglais(ÉtatsUnis)

2.2 Modeling approach and assumptions

2.2.1 Geometry and meshing

120 Since the computation is performed using a discretized land surface, smaller mesh sizes will-lead to detailed topography but will lengthen the computation time and memory requirements. Hereinafter, we consider a reference square landscape model of 50 km side with a grid resolution of 90 m, which is a good compromise between computation time (3-5 hours on a PC workstation) and the total amount of basins that can be studied (>1000). Our results are not affected when the meshgrid resolution is increased to 30 m nor are they when the model size is multiplied by four (100x100 km) (See (see Fig. S1S2).

125 2.2.2 Boundary conditions

In order to isolate the effect of divide migrations on the variability of basin-wide denudation rates, we explore simple models with constant and spatially uniform uplift and precipitation rates and we assume no horizontal advection $(\partial z/\partial t)_{td} = 0$. We use a Dirichlet boundary condition: simulation edges are not affected by uplift on a one pixel band to represent a stable bas e level for rivers. The model presents no initial topography, except for gaussian noise ranging between 0 and 50 m so as to initiate a random fluvial network.

130

2.2.3 Set of parameters

Firstly, we consider a reference model with parameters commonly used for moderately active orogens: an uplift rate U of 0.1 mm.yr⁻¹, a diffusivity D of 10^2 m².yr⁻¹ (Roering et al., 1999), a threshold slope S_c of 30° (Burbank et al., 1996; Montgomery et Brandon, 2002; Binnie et al., 2007), a m/n ratio of 0.5 with m = 0.5 and n = 1, an erodibility coefficient K of $\frac{5 \times 10^6}{5 \times 10^6}$ $m^{+}1x10^{-5}$ m^(1-2m)-yr⁻¹, a ρ_r / ρ_s ratio of 1.3 and a critical drainage area A_{z} of 0.2 km² (Montgomery et al., 1993).

135

140

Secondly, all other parameters held constant, we investigate the specific impact of uplift rate, erodibility and hillslope processes in other models by varying U, K, D and S_c between 0.501 and 21 mm. $vr^{-1}, 5.10^{5}$ m^(1-2m). vr^{-1} and 5.10^{6} m^(1-2m). vr^{-1} . 10^{-3} and 10^{-1} m².yr⁻¹ and 20° and 40° , respectively.

In order to better constrain the variability of our results under similar conditions, we run for each model five simulations using the same parameters, but with different initial random topographies.

2.2.4. Timescale

The total duration of simulations is 10 Myr, which is nearly one order of magnitude longer than the theoretical time to reach general topographic equilibrium for our set of parameters (Willett, 2001)... The implicit scheme used to simulate non-linear hillslope processes provides stable solutions regardless of the time step. In contrast, the explicit scheme used to model fluvial

145 incision requires a time step that satisfies the Courant-Friedrich-Lewy criterion. Hereinafter, we choose a time step Δt of $\frac{5005000}{5000}$ yr for hillslope diffusion. Our results are not affected when using a smaller Δt (i.e. 1000 yr) (see Fig. S1). Incision

Mis enforme: Anglais(ÉtatsUnis)

Mis en forme: Anglais(ÉtatsUnis)
Mis enforme: Anglais(ÉtatsUnis)
Mis en forme: Anglais(ÉtatsUnis)
Mis en forme: Anglais(ÉtatsUnis)
Mis en forme: Anglais(ÉtatsUnis)
Mis en forme: Anglais(ÉtatsUnis)
Mis en forme: Anglais(ÉtatsUnis)
Mis en forme: Anglais(ÉtatsUnis)
Mis en forme: Anglais(ÉtatsUnis)
Mis en forme: Anglais(ÉtatsUnis)
Mis en forme: Anglais(ÉtatsUnis)
Mis en forme: Anglais(ÉtatsUnis)
Mis en forme: Anglais(ÉtatsUnis)
Mis en forme: Anglais(ÉtatsUnis)

computation is nested in this time step and uses another time step that is automatically determined to assure model stability (Campforts et al., 2017).

2.3 Basin-wide denudation rates and aggressivity metrics

150 **2.3.1 Basin-wide denudation rates**

155

170

We derive basins from the synthetic DEMs (Digital Elevation Models) using an accumulation map computed with a single flow direction algorithm implemented in TopoToolbox (Schwanghart and Scherler, 2014). Next, we calculate for each basin the variation in <u>average</u> elevation over a time interval of 10 kyr, which averages the results over 20 time steps. The drainage network <u>can migratemigrates</u> during the simulation, so we only survey the basins that keep the same outlet location during this time interval. Furthermore, due to divide mobility, the geometry of watersheds can also change. Hence, we measure the average difference in elevation inside the basin perimeter after 10 kyr. Here we only assess the surface uplift U_s (England

and Molnar, 1990). To obtain approximate the real-denudation rates $E_{\rm s}$ for each basin, we sum the surface uplift $U_{\rm s}$ with the rock uplift rate $U_{\rm sesumed}$ in our simulation and divide the result by the time interval, which is an approximation. However, ealeulating incremental denudation rates over each time step is prohibitive in terms of computation time for the number of

- 160 extracted basins. By considering the relatively small period over which we integrate denudation (10 kyr), we then assume that these approximations have a negligible impact on the results at first order. Calculated that way, the denudation rate, If the basin is in a topographic steady state, U_s is equal to zero and E is equal to the background uplift rate. Thus, a positive (negative) value of U_s traduce a deficit (an excess) of denudation. Calculated that way, E is sensitive to divide migration but also to transient features like knickpoints that migrate along the river network. In our simulations, knickpoints may develop due to (1) the dissection of the initial flat surface or (2) discrete drainage captures (see Sec. 3.1). We use the knickpointfinder
 - algorithm implemented in TopoToolbox (Schwanghart and Scherler, 2014) to identify the affected basins.

2.3.2 From cross-divide metrics to basin averaged aggressivity metrics

Most recent studies have focused on the relationship between drainage divide mobility and headwater across-divide contrast in either χ , slopegradient, elevation or local relief values (e.g.-Whipple et al., 2017; Forte and Whipple, 2018). Here, in line with Willett et al. (2014, see Supp. Mat. therein) we focus on the specific influence of divide migration on denudation rates

at the scale of the entire stream basin. Our approach aims to integrate cross-divide contrasts in drainage network properties along the entire basin perimeter. We then obtain basin-averaged aggressivity metrics that determine if a watershed is either growing or shrinking (Willett et al. 2014).

First we calculate χ , local slope and elevation map for each channel pixel. χ First, we assess χ , local topographic gradient *G* and height *H* of the drainage network at a reference drainage area A_{ref} (Fig. 2). Ideally, A_{ref} must be equal to the area at which channelization occurs (Forte and Whipple, 2018). However, it is challenging to locate the accurate position of channel

Mis enforme: Anglais(ÉtatsUnis)

Mis en forme: Anglais(ÉtatsUnis) Mis en forme: Anglais(ÉtatsUnis) Mis en forme: Anglais(ÉtatsUnis)

Mis enforme: Anglais(ÉtatsUnis)

{	Mis enforme: Couleurde police : Noir
{	Mis en forme: Anglais(ÉtatsUnis)
	Mis en forme: Couleurde police :

Noir

heads (Clubb et al., 2016). Hence, we use a constant value of A_{ref} set to 1 km². The parameter χ , is an integral function of position along the channel network (Perron and Royden, 2012) described by the equation:

180
$$\chi = \int_{x_b}^x \left(\frac{A_0}{A(x)}\right)^{\frac{m}{n}} dx$$

185

(4)

where A(x) is the upstream drainage area at the location x, A_0 is an arbitrary scaling area set to 1 km². The *m* over *n* ratio refers here to the reference concavity of an equilibrated river profile. Its value is set to 0.5 in accordance with the model parameters. For each independent drainage network, we integrate χ from the outlet x_{b_1} located at the model boundary (< 1 m high), to the channel heads. We then Local gradient is determined for each DEM pixel from its eight-connected neighbors.

Height is simply extracted from the DEM.

Then, we calculate the difference in channel head χ , local slope of metrics $(\Delta \chi, \Delta G$ and elevation ΔH) across each first order basin the segments of divide segment. The shared by two reference basins. Finally, the aggressivity metric is finally obtained by averaging these first order across-divide differences along the perimeter of each extracted sampled basin (Fig. S-2). This way, the sign of the aggressivity metric in a basin corresponds to the difference of the averaged value of considered metric

- 190 way, the sign of the aggressivity metric in a basin corresponds to the difference of the averaged value of considered metric (Channel head χ , slope or elevation difference ($\Delta \chi \Delta G$ and ΔH) in this basin with respect to hisits neighbours. This method has the advantage to ponderate the weight of individual divide segments by the number of pixels they contain, and then to provide a robust assessment of the basin aggressivity. Aggressivity metrics based on χ , G and H are hereafter referred to as $\Delta \chi_{av} \Delta G_{av}$ and ΔH_{av} , respectively, which provide an accurate assessment of the basin aggressivity. However, due to
- 195 topology issues, some parts of the perimeter of the sampled basins may be not shared by two reference basins (Fig. 2). We quantify this incompleteness by assessing the ratio of documented pixels over the total amount of pixel along the basin perimeter. We refer to this ratio as the "confidence index" CI, assuming that an higher CI is associated with a more robust basin aggresivity assessment.

Mis enforme: Anglais(ÉtatsUnis) Mis en forme: Anglais(ÉtatsUnis) Mis enforme: Anglais(ÉtatsUnis)

Mis en forme: Anglais(ÉtatsUnis)

3 Results

200 **3.1 Evolution of reference model**

A detailed analysis of the DEM suggests that during the initial phase, the flat initial surface (Fig. $2a_{3a}$) is progressively uplifted to form a plateau. At the same time the edges of this plateau are gradually regressively eroded by drainage networks that spread from the base level toward the center of the model (Figs. $2b_{3b}$ and c). This transient landscape is completely dissected after 2 Myr. From this time and until the end of the simulation, landscape changes are mainly due to competition

²⁰⁵ between watersheds, resulting in continuous divide migrations with decreasing intensity as the model is moving toward a total topographic equilibrium (Fig. 2d3d to f; Supplementary Video n°1).

WeTo define the time period of regional steady state, we measure the average elevation, the maximum elevation and the average denudation rate over the entire model for each time step (Fig. 3a) and 4a). We identify two distinct stages during the evolution of our reference simulation. During the first million years, due to long wavelength topographic building, the calculated landscapes are far from steady state. This leads to a major increase of the mean elevation from 0 m to ca. 70075

calculated landscapes are far from steady state. This leads to a major increase of the mean elevation from 0 m to ca. 70075
 m. In a second stage, this trend reverses and the mean elevation decreases asymptotically toward ca. 60060 m until the end of the simulation.

The evolution of the maximum elevation follows the same pattern but can be affected by temporal changes in the location and altitude of higherhighest peaks. The maximum elevation increases between 0 and ca. $\frac{2200250}{2200250}$ m over the first $\frac{2}{2}$ Myr

215 (Figs. 3a<u>3Myr (Fig. 4a</u>) then decreases progressively to reachremain at ca. 1600200 m atduring the endrest of the simulation. We compute the average denudation rate from the tectonic prock uplift rate and from average elevation change over the entire model between two time steps ÷:

$$(\Delta z / \Delta t)_{av} = U - E_{av} ,$$

220

where $(\Delta z/\Delta t)_{av}$ is the average surface uplift over the entire model on a time-step Δt , U-is the imposed uniform uplift rate Mis enforme: Anglais(ÉtatsUnis) Mis enforme: Anglais(ÉtatsUnis) (0.1 mm.yr⁻¹) and E_{av} is the average "real" denudation rate. During the first 0.25 Myr, the mean denudation rate falls Mis enforme: Anglais(ÉtatsUnis) abruptly from ca. 0.6 mm, yr⁻¹ to nearly 0 mm, yr⁻¹ as a consequence of diffusion over the initial flat topography. After that Mis enforme: Anglais(ÉtatsUnis) time and until the first 1 Myr, the mean denudation rate increases but remains lower than the uplift rate, leading to the 225 increase in average elevation over this time period. Next in In the following 0.51 Myr. Fax exceeds the uplift rate to reach up Mis enforme: Anglais(ÉtatsUnis) to 1.030.104 mm.yr⁻¹. It then before it gently decreases to 0.1 mm.yr⁻¹ duringuntil the restend of the simulation. This shows that topography tends to - but never reaches - a strict steady state over the simulation time. Abrupt changes in E_{av} after ca. 2Mis enforme: Anglais(ÉtatsUnis) Myr, 2.5 Myr and 3.2 Myr can also be highlighted (Fig. 3b). These brief variations 2.5, 3.5, 4.5 and 9.5 Myr (red circles in Figure 4b) are related to major local captures in the drainage network, which can be observed during the model evolution 230 (red circles in Fig. 2d3e and f and Supplementary Video n°1). On the basis of Based on these results, we will consider that quasi-a regional topographic steady-state is reached between 1.5 and 2 Ma, when the plateau relict topography is totally eroded and F_{av} begins to decrease (Figs. $\frac{23}{24}$ and $\frac{34}{24}$). This time is Mis enforme: Anglais(ÉtatsUnis) consistent with the time required to reach topographic steady state proposed from models with constant uplift rate and no horizontal advection (Willett et al., 2001). 3.2 Basin-wide denudation rates variability 235 We calculate basin-wide denudation rates E upstream of each stable drainage network confluence after 2.5 Myr, 5 Myr and Mis enforme: Anglais(ÉtatsUnis) 10 Myr of simulation (Figs. 4a, b and c). As explained in the method section, we compiled the results obtained for five runs ampled basins.5a, b and c, in order to

(5)

240 basin size. This As exposed by Forte & Whipple (2018), the erosion rate contrasts across divides is spatially limited to areas			
 very near the divides. Thus, the variability is maximum for small basins (ca. 1 km³) and decreases with increasing basin area. In our approach, small basins are nested in larger ones, Hence, these results can be related to the averaging of demudation rates along the drininge network, in agreement with the measurements of Matmon et al. (2005b). This variability also decreases with time (Figs. 4,-bord-5_2), c) For basins with an excess of demudation relative to the upilit rate <i>l</i>, the <i>f</i>/<i>l</i> ratio and the only 2_1_2 for 2_5 Myr but only 2_1 after 5.5 Myr and 1.52 after 10 Myr, Basins with a demudation excess sha tard and containe excess. All the admudation excess sha tard and containe excess. All the admudation factors also gives and the ratio is a second with a capture event visible in figure dh. For basins with a deficit of demudation, therefue, evolution of the ratio is Less obvious. It can be been than 0.255 after 2.5 Myr, but increases shiftly to 0.4-after 4.55 und; 10 Myr. These results reflect a significant spatial variability of the difference between basin-wide demudation intes and 0.5-sher; 10 Myr, plutinate. To assess more accurately the temporal evolution of this variability, we excess of the difference between basin-wide admudation in excess. After 1.5 Ma, basins in deficit (Fig. 4451). Until 1.5 Ma, basins are located on the plateau where demudation rates is null. This leads to a loc MAD for basins with an area of 1.2 km²-and 1.02 km² for co. 0.422 mmyr⁻¹ (so ca. 0.422 mmyr⁻¹) 20 ca. 0.4607 mmyr⁻¹ 20 ca. 0.4607 mmyr⁻¹ for the difference between basin wide demudation rates are upilit mate. The server is a deficit e shift an excess, the MAD value decreases through time, depending on drainage area: iron ca. 0.422 mmyr⁻¹ 20 ca. 0.4627 mmyr⁻² 20 ca. 0.4627 mmyr⁻¹ for basins with an area of 1.2 km²-and 1.02 km² form ca. 0.422 mmyr⁻¹ or ca. 0.492 mmyr⁻¹ for the basins of decided by these contrasts are of the 200km². These createshere these and upilit mate in our ref		respectively). Regardless of the duration, we observe a significant variability in the calculated denudation rates depending on	Mis en forme: Anglais(Roy aume-Un)
 In our approach, small basins are nested in larger ones. Hence, these results can be related to the averaging of denudation rates along the drainage network, in agreement with the measurements of Matmon et al. (2005b). This variability also decreases with time (Figs. 4t, burd Succ). For basins with an excess of denudation relative to the upfit rate (<i>L</i>, the <i>E</i>/<i>I</i>) ratio decreases with time (Figs. 4t, burd Succ). For basins with an excess of denudation relative to the upfit rate (<i>L</i>, the <i>E</i>/<i>I</i>) ratio decreases with time (Figs. 4t, burd Succ). For basins with an excess of denudation relative to the upfit rate (<i>L</i>, the <i>E</i>/<i>I</i>) ratio due of the general trend at 10 Ma (Fig. 5c) are associated with a capture event visible in figure 4b. For basins with a deficit of denudation, thick evolution of the ratio is less obvious. It can be lower than 0.455 after 2.5 Myr, but increases slightly to demudation mates and 0.55 there results reflect a significant spatial variability of the difference between basin-wide demudation in the state of the plate and the denudation for the variability, we excess orin deficit (Fig. 4451). Until 1.5 Ma, basins are located on the plateau where demudation rate is multi. This leads to a sociated on the plateau where demudation rate is multi. This leads to a sociated on the plateau where demudation rate is multi. This leads to a sociated or the plateau where demudation rate is multi. This leads to a sociated or the plateau where demudation rate is multi. This leads to a sociated the relates and would be demudation rates. After 1.5 Ma, basins in deficit eshibit an esymptotic increase in MAD from nearly -0.215 to -0.051 mm yr⁻¹ regard less of the area class considered. Sociated the measurement of 402-000 km⁻¹. These results mether to englishing from ca. 0.142 mm yr⁻¹ for the difference between basin with a mate and uplift mate in our reference models largest basins. We also see a coherent evolution of this difference over the simulation time, consistent with the direction of	240	basin size. This As exposed by Forte & Whipple (2018), the erosion rate contrasts across divides is spatially limited to areas	
 rates along the drainage network, in agreement with the measurements of Matmon et al. (2003b). This variability also decreases with time (Figs. 4p. band-5a.cs). For basins with an excess of denudation relative to the uplift rate <i>J</i>. the <i>F</i>/<i>J</i> ratio can each up to 42.5 after 2.5 Myr but only 2 after 5 Myr and 1.52 after 10 Myr. Basins with a deficit of denudation, twelve evolution of this is as a cassance basin with a denudation excess. After 1.5 Myr. but increases slightly to 0.4 after 45 myr for three distinct categories of basin sizes: 1-2 km², nlo-20 km² wet han denudation in excess or in deficit (Fig. 4452). Until 1.5 Ma, basins are located on the plateau where denudation rates is null. This kads to a low MAD for basins with a denudation deficit and to the absence of basins with a denudation excess. After 1.5 Ma, basins in excess, the MAD value decreases through time, depending mmyr⁴, regardless of the access considered. Semmyr⁴ 25 to ca. 04507 mmyr⁴ for basins with a denudation deficit and to the absence of basins with a denudation excess. After 1.5 Ma, basins in excess, the MAD value decreases through time, depending mmyr⁴, regardless of the and cass considered. Semmyr⁴ 25 to ca. 04507 mmyr⁴ for basins with a denudation rates is neither homogeneous nor randomly distributed (Fig. 5462). This contasts with an encor of 100-200 km². We take see conterent evolution of this difference over the simulation time, consistent with the model progression toward etotation of drainage basins with a constant is in bedwaterchannel <i>g</i>, stoperanding and elevationing drainage are at 500 km². These constast are indexed within more form the uplift mate by consistent with the direction of divide mobility obtained from our model. One may note that the higher the contast in these sea coherent evolution of this difference over the simulation time, consistent with the model progression toward etotast on topographic equilibrium. These splone of sampled basin in this dataset cotatins a binkippoint. Th		very near the divides. Thus, the variability is maximum for small basins (ca. 1 km ²) and decreases with increasing basin area.	Mis en forme: Anglais(Roy aume-Un)
decreases with time (Figs. 4m ⁺ end-5m ⁺ sc.). For basins with an excess of denudation relative to the uplift rate <i>U</i> , the <i>E/U</i> ratio can reach up to \$22_5 after 2.5 Myr but only 2 after 5 Myr and 1.52 after 10 Myr. Basins with a denudation excess that stand out of the general trend at 10 Mar (Fig. 5c) are associated with a capture event visible in figure 4b. For basins with a deficit of denudation, thicke evolution of the ratio is less obvious. It can be lower than 0.455 after 2.5 Myr, but increases slightly, to 0.4 effect + 50 until 10 Myr. These results reflect a significant spatial variability of the difference between basin-wide demudation rates and 0.6 effect 10 Myr. These results reflect a significant spatial variability of the difference between basin-wide demudation intes and 0.6 effect 10 Myr. These results reflect a significant spatial variability of the difference between basin-wide demudation intes and 0.6 effect 10 Myr. These results reflect a significant spatial variability of the difference between basin-wide demudation rates and 0.6 effect 10 Mar. These results are basin set to cated on the plateau where denudation recess. After 1.5 Ma, basins in deficit exhibit an examptionic increase in MAD from the uplift rate by considering separately basins with a denudation is excess or in deficit (Fig. 4452). Until 1.5 Ma, basins are located on the plateau where denudation recess. After 1.5 Ma, basins, with a denudation effect and to the absence of basins with a denudation excess. After 1.5 Ma, basins, with a mean of 100 200 km². These results reflect a significant spatial variability of 10.20 km² after 1.5 Ma, basins, with a mean of 100 200 km². These results reflect a significant spatial variability of 10.20 km² after 1.5 Ma, basins, with a deficit exhibit an examption with demudation rates is neither homogeneous nor randomly distributed (Fig. 5c). The location of drainage basins with demudation rates is neither homogeneous nor randomly distributed (Fig. 5c). The location of drainage basins with d		In our approach, small basins are nested in larger ones. Hence, these results can be related to the averaging of denudation	
 can reach up to 22_5 after 2.5 Myr but only 2 after 5 Myr and 1.52 after 10 Myr. Basins with a denudation excess that stand out of the general tend at 10 Ma (Fig. 5-1) are associated with a capture event visible in figure 4b. For basins with a deficit of denudation, this due events of the analysis (EtatsUnis) Q 1-ster 5_0 until 10 Myr. These results reflect a significant spatial variability of the difference between basin-wide demudation rates and 6-ster 10 Myrphilin auc. To assess more accurately the temporal evolution of this variability. We then estimate the mean absolute deviation (MAD) from the uplif rate by considering separately basins with a denudation in excess of in deficit (Fig. 4451). Until 1.5 Ma, basins are located on the plateau where denudation excess. After 1.5 Ma, basins in deficit and to the absence of basins with a denudation excess. After 1.5 Ma, basins are located on the plateau where denudation tare ac class considered. For basins with a derudation fits and 10-20 km² at 0.04501 mm.yr⁴, egardRess of the are class considered. Consistent with an area of 1-02 km² at 0.04501 mm.yr⁴, egardRess of the area class. Considered at 100 200km². These results reflect a significant spatial variability of 10-20 km² at 0 ca. 0.04501 mm.yr⁴ for basins with a area of 1-02 km² at 0.04501 mm.yr⁴ in our of-ence mode b-largest basins. We also see a coherent evolution of this difference over the simulation time, consistent with the model progression toward atotel deviation divide mobility obtained from our model. One may note that the higher the contrast in these modes progression toward atotel deviation divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the data set contains a link/cpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates in this data		rates along the drainage network, in agreement with the measurements of Matmon et al. (2003b). This variability also	Mis en forme: A nglais(États U nis)
out of the general trend at 10 Ma (Fig. Sc) are associated with a capture event visible in figure 4b, For basins with a deficit of denudation, the face volution of the ratio is less obvious. It can be lower than 0.255 after 2.5 Myr, but increases slightly to 0.4-effer-50 until 10 Myr, These results reflect a significant spatial variability of the difference between basin-wide denudation attes and 0.5-effer 10 Myrpifif nute. To assess more accurately the temporal evolution of this variability. we excess or in deficit (Fig. 4451). Until 1.5 Ma, basins are located on the plateau where denudation rates is notified to the absence of basins with a denudation in excess. After 1.5 Ma, basins in deficit exhibit an expension (MAD) from the uplift rate by considering separately basins with a denudation in excess. If or basins with a area of 1-2 km ² and 10.20 km ² and from ca. 0.420 mmyr ⁴ to ca. 0.4507 mmyr ⁴ (25 C) to ca. 0.4607 mmyr ⁴ for basins with a area of 1-2 km ² and 10.20 km ² and from ca. 0.7 to ca. 0.001 mmyr ⁴ for basins with an area of 1-2 km ² and 10.20 km ² and from ca. 0.7 to ca. 0.001 mmyr ⁴ for basins. We also see a coherent evolution of this difference over the simulation time, consistent with the model progression toward evolution of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide the bight the divide migration grediked by these contrasts are to pographic equilibrium. These Mone of sampled basin in this data set contains a lankclopoint. Thus, these results based on simulation rates is primarily controlled bas will a basin such as expland (shink) show higher (lower) 200 consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the assis with a constant poperities as well as constant boundary conditions conting uniform		decreases with time (Figs. $\frac{4a}{b}$ and $\frac{5a}{2}$ c). For basins with an excess of denudation relative to the uplift rate U , the E/U ratio	Mis en forme: A nglais(États U nis)
 bit of the second in the product in the product in the second of the second of the product in the second of the product in the second of the product in the second of the second of the product in the second of the sec	245	can reach up to 32.5 after 2.5 Myr but only 2 after 5 Myr and 1.57 after 10 Myr. Basins with a denudation excess that stand	Mis en forme: Anglais(ÉtatsUnis)
 0.4 = fer = 56 until 10 Myr, These results reflect a significant spatial variability of the difference between basin-wide denudation rates and 0.5 after 10 Myruphin rate. To assess more accurately the temporal evolution of this variability, we calculate <i>f</i> every 0.5 Myr for three distinct categories of basin sizes: 1-2 km², 10-20 km² and 100-200 km². We then estimate the mean absolute deviation (MAD) from the uplift rate by considering separately basins with a denudation in excess or in deficit (Fig. 4452). Until 1.5 Ma, basins are located on the plateau where denudation rate is null. This leads to a low MAD for basins with a denudation deficit and to the absence of basins with a denudation excess. After 1.5 Ma, basins in deficit exhibit an asymptotic increase in MAD from nearly -0.4516 to -0.4501 mmyr⁻¹, regardless of the area class considered. For basins in excess, the MAD value decreases through time, depending on drainage area : from ca. 0.76 to ca. 0.0490 mmyr⁻¹ for basins with an area of 1-2 km² and 102 0 km², from ca. 0.442 mm yr⁻¹ to ca. 0.0507 mm yr⁻¹ for basins with an area of 1-2 km² and 102 0 km², from ca. 0.442 mm yr⁻¹ to ca. 0.0507 mm yr⁻¹ for basins with an area of 1-2 km² and 102 0 km², from ca. 0.442 mm yr⁻¹ to ca. 0.0507 mm yr⁻¹ for basins with an area of 1-2 km² and 102 0 km², from ca. 0.442 mm yr⁻¹ to ca. 0.0507 mm yr⁻¹ for basins with an area of 1-2 km² and 102 0 km², from ca. 0.442 mm yr⁻¹ to ca. 0.0507 mm yr⁻¹ for basins with an area of 1-2 km² and 102 0 km², from ca. 0.442 mm yr⁻¹ to cincides with migrating drainage divides (Fig. 5m² and) with cross-divide contrasts is neither homogeneous nor randomly distributed (Fig. 5m² and 100 contary to the denudation rates is neither homogeneous nor randomly distributed (Fig. 5m² and 100 contary to more from the equilibrium value of 0.1 mm yr⁻¹ coincides with migrating drainage divides (Fig. 5m² and 0 fort and Whipple (2018), the divide migrations predicted by the		out of the general trend at 10 Ma (Fig. 5c) are associated with a capture event visible in figure 4b. For basins with a deficit of	Mis en forme: Anglais(ÉtatsUnis)
 denudation rates, and 0.5-mer-10 Myraphift rate. To assess more accurately the temporal evolution of this variability, we calculate <u>f</u> every 0.5 Myr for three distinct categories of basin sizes: 1-2 km², 10-20 km² and 100-200 km². We then estimate the mean absolute deviation (MAD) from the uplift rate by considering separately basins with a denudation in excess or in deficit (Fig. 445(1). Until 1.5 Ma, basins are located on the plateau where denudation rate is null. This leads to a low MAD for basins with a denudation deficit and to the absence of basins with a denudation excess. After 1.5 Ma, basins in deficit exhibit an exymptotie-increase in MAD from mearly -0.215 to -0.05021 mm.yr⁻¹, regardless of the area class considered. For basins in excess, the MAD value decreases through time, depending on drainage area : from ca. 0.28 mm.yr⁻¹25 to ca. 0.05027 mm.yr⁻¹ for basins with an area of 1-0 200 km² and 100-200 km² from ca. 0.442 mm.yr⁻¹ to ca. 0.05027 mm.yr⁻¹ for basins with an area of 1-0 200 km² and 100-200 km² from ca. 0.442 mm.yr⁻¹ to ca. 0.05027 mm.yr⁻¹ for basins with an even of 100-200 km² and 100-200 km² from ca. 0.442 mm.yr⁻¹ to ca. 0.05027 mm.yr⁻¹ for basins with events basin wide denudation rates and uplift mate in our reference modely-largest basins. We also see a coherent evolution of this difference over the simulation time, consistent with the model progression toward e-total topographic equilibrium. The spatial variability of the denudation rates is neither homogeneous nor randomly distributed (Fig. 5m²/₂₀). The location of divide mobility obtained from our model. One may note that the higher the contrast in heedwaterchannel <u>y</u>, slopegradient and elevationheight (Figs. 5b, et and <u>6</u>, d). Following Willett et al. (2014) and Forte and Whilpple (2018), the divide migrations predicted by these contrasts are consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in heesw		denudation, this the evolution of the ratio is less obvious. It can be lower than 0.255 after 2.5 Myr, but increases slightly to	
 cakulate <u>E</u> every 0.5 Myr for three distinct categories of basin sizes: 1-2 km², 10-20 km² and 100-200 km². We then estimate the mean absolute deviation (MAD) from the uplift rate by considering separately basins with a denudation in excess or in deficit (Fig. 4452). Until 1.5 Ma, basins are located on the plateau where denudation rate is null. This leads to a low MAD for basins with a denudation deficit and to the absence of basins with a denudation excess. After 1.5 Ma, basins in deficit exhibit an exymptotic-increase in MAD from nearly -0.215 to -0.6501 mm.yr⁻¹, regardless of the area class considered. For basins in excess, the MAD value decreases through time, depending on drainage era i from ca. 0.28 mm.yr⁻¹ 25 to ca. 0.0507 mm.yr⁻¹ for the difference between basin wide denudation time, and on 1.9 L km² end 10.2 0 km² from ca. 0.412 mm.yr⁻¹ to ca. 0.0507 mm.yr⁻¹ for basins with an area of 1-2 km² end 10.2 0 km² from ca. 0.412 mm.yr⁻¹ to ca. 0.0507 mm.yr⁻¹ for basins with an area of 1-2 km² end 10.2 0 km² from ca. 0.412 mm.yr⁻¹ to ca. 0.0507 mm.yr⁻¹ for basins with an era of 100-200km². These results reflect a significant spatial variability or 10.2 km² and from ca. 0.7 to ca. 0.041 mm.yr⁻¹ for the difference between basin wide demudation rates and uplif rate in our reference modely-largest basins. We also see a coherent evolution of this difference over the simulation time, consistent with the model progression toward attended for drainage divides (Fig. 2e3d) and with cross-divide contrasts in headwaterchannel <u>y</u>, sbegeradient and elevationheight (Figs. 5b; end <u>6</u> d). Following Willett et al. (2014) and Forte and Whipple (2018), the divide migrations predicted by these contrast in these parameters across the divide, the higher the deviation rate from ur model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the denudation rates a spreperise as well as constant boundary conditions conf		0.4 after 56 until 10 Myr. These results reflect a significant spatial variability of the difference between basin-wide	
 estimate the mean absolute deviation (MAD) from the uplift rate by considering separately basins with a denudation in excess or in deficit (Fig. 4454). Until 1.5 Ma, basins are located on the plateau where denudation rate is null. This leads to a low MAD for basins with a denudation deficit and to the absence of basins with a denudation excess. After 1.5 Ma, basins in deficit exhibit an exymptotic increase in MAD from nearly -0.215 to -0.0501 mm.yr⁻¹, regardless of the area class considered. For basins in excess, the MAD value decreases through time, depending on drainage area : from ca. 0.38 mm.yr⁻¹ for to asins with an area of 1-2 km² and 10.20 km²/₂ from ca. 0.442 mm.yr⁻¹ to ca. 0.4507 mm.yr⁻¹ for basins with an area of 1-2 km² and 10.20 km²/₂ from ca. 0.422 mm.yr⁻¹ to ca. 0.0501 mm.yr⁻¹ for basins with an area of 1-0200km². These results reflect a significant spatial variability of 10.20 km² and from ca. 0.7 to ca. 0.041 mm.yr⁻¹ for the difference between basin-wide denudation rates and uplift rate in our reference models largest basins. We also see a coherent evolution of this difference over the simulation time, consistent with the model progression toward a total topographic equilibrium. The spatial variability of the denudation rates is neither homogeneous nor randomly distributed (Fig. 5450). The location of drainage basins with denudation rates is neither homogeneous nor randomly distributed (Fig. 5450). The location of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from tropographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrin		denudation rates and 0.5 after 10 Myruplift rate. To assess more accurately the temporal evolution of this variability, we	
 estimate the mean absolute deviation (MAD) from the uplift rate by considering separately basins with a denudation in excess or in deficit (Fig. 4454). Until 1.5 Ma, basins are located on the plateau where denudation rate is null. This leads to a low MAD for basins with a denudation deficit and to the absence of basins with a denudation excess. After 1.5 Ma, basins in deficit exhibit an asymptotic increase in MAD from nearly -0.215 to -0.0504 mm.yr⁴, regardless of the area class considered. For basins in excess, the MAD value decreases through time, depending on drainage area : from ca. 0.28 mm.yr⁴ for basins with an area of 1-2 km² and 10.20 km²c₂ from ca. 0.442 mm.yr⁴ to ca. 0.4507 mm.yr⁴ for basins with an area of 1-020km². These results reflect a significant spatial variability of 10.20 km² and from ca. 0.7 to ca. 0.041 mm.yr⁴ for bedifference between basin wide demudation tates and uplift mate in our reference models largest hasins. We also see a coherent evolution of this difference over the simulation time, consistent with the model progression toward a total topographic equilibrium. The spatial variability of the denudation rates is neither homogeneous nor randomly distributed (Fig. 5450). The location of drainage basins with denudation rates is neither homogeneous nor randomly distributed (Fig. 5450). The location of divide most-divide contrasts in headwaterschannel <i>g</i>, sipograficient and elevatiomheight (Figs. 5h, cand 6h, d). Following Willett et al. (2014) and Forte and Whipple (2018), the divide migrations predicted by these contrasts are consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant	250	calculate E every 0.5 Myr for three distinct categories of basin sizes: 1-2 km ² , 10-20 km ² and 100-200 km ² . We then	Mis en forme: Anglais(ÉtatsUnis)
 low MAD for basins with a denudation deficit and to the absence of basins with a denudation excess. After 1.5 Ma, basins in deficit exhibit an exymptotic increase in MAD from nearly -0.215 to -0.9501 mm.yr¹, regardless of the area class considered. For basins in excess, the MAD value decreases through time, depending on drainage area : from ca. 0.28 mm.yr⁴ 25 to ca. 0.6507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.6507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr² coincides with migrating drainage divides (Fig. 5, 2, 2, 2) and with cross-divide contrasts in headwaterchannel χ, slopegradient and elevationheight (Figs. 5, 5, e and for basins with and erction of divide mobility obt	I		
 low MAD for basins with a denudation deficit and to the absence of basins with a denudation excess. After 1.5 Ma, basins in deficit exhibit an exymptotic increase in MAD from nearly -0.215 to -0.9501 mm.yr¹, regardless of the area class considered. For basins in excess, the MAD value decreases through time, depending on drainage area : from ca. 0.28 mm.yr⁴ 25 to ca. 0.6507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.6507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr¹ to ca. 0.0507 mm.yr¹ for basins with an area of 1-2 km² and 10 20 km² 2, from ca. 0.142 mm.yr² coincides with migrating drainage divides (Fig. 5, 2, 2, 2) and with cross-divide contrasts in headwaterchannel χ, slopegradient and elevationheight (Figs. 5, 5, e and for basins with and erction of divide mobility obt	1	excess or in deficit (Fig. 445d). Until 1.5 Ma, basins are located on the plateau where denudation rate is null. This leads to a	
 For basins in excess, the MAD value decreases through time, depending on drainage area : from ca. 0.28 mm.yr⁴ j2 to ca. 0.6907 mm.yr⁴ for basins with an area of 1-2 km² and 10-20 km²,² from ca. 0.442 mm.yr⁴ to ca. 0.6907 mm.yr⁴ for basins with an area of 100 200km². These results reflect a significant spatial variability of 10-20 km² and from ca. 0.7 to ca. 0.04 mm.yr⁴ for the difference between basin wide denudation rates and uplift rate in our reference models. largest basins. We also see a coherent evolution of this difference over the simulation time, consistent with the model progression toward a total topographic equilibrium. The spatial variability of the denudation rates is neither homogeneous nor randomly distributed (Fig. 5m⁶a). The location of drainage basins with denudation rates is neither homogeneous nor randomly distributed (Fig. 5m⁶a). The location of divides (Fig. 2e3d) and with cross-divide contrasts in headwater_channel <i>χ</i>, slopegradient and elevationheight (Figs. 5b, cand 6b-d). Following Willett et al. (2014) and Forte and Whipple (2018), the divide migrations predicted by these contrasts are consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) 270 denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology 	I	low MAD for basins with a denudation deficit and to the absence of basins with a denudation excess. After 1.5 Ma, basins in	
 For basins in excess, the MAD value decreases through time, depending on drainage area : from ca. 0.28 mm.yr⁴ j2 to ca. 0.6907 mm.yr⁴ for basins with an area of 1-2 km² and 10-20 km²,² from ca. 0.442 mm.yr⁴ to ca. 0.6907 mm.yr⁴ for basins with an area of 100 200km². These results reflect a significant spatial variability of 10-20 km² and from ca. 0.7 to ca. 0.04 mm.yr⁴ for the difference between basin wide denudation rates and uplift rate in our reference models. largest basins. We also see a coherent evolution of this difference over the simulation time, consistent with the model progression toward a total topographic equilibrium. The spatial variability of the denudation rates is neither homogeneous nor randomly distributed (Fig. 5m⁶a). The location of drainage basins with denudation rates is neither homogeneous nor randomly distributed (Fig. 5m⁶a). The location of divides (Fig. 2e3d) and with cross-divide contrasts in headwater_channel <i>χ</i>, slopegradient and elevationheight (Figs. 5b, cand 6b-d). Following Willett et al. (2014) and Forte and Whipple (2018), the divide migrations predicted by these contrasts are consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) 270 denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology 	1	deficit exhibit an asymptotic-increase in MAD from nearly -0.215 to -0.0504 mm.yr ⁻¹ , regardless of the area class considered.	
 0.6507 mm.yr⁻¹ for basins with an area of 1-2 km² and 10-20 km² ²/₂ from ca. 0.442 mm.yr⁻¹ to ca. 0.0507 mm.yr⁻¹ for basins with an area of 100-200km². These results reflect a significant spatial variability of 10-20 km² and from ca. 0.7 to ca. 0.04 mm.yr⁻¹ for the difference between basin wide denudation rates and uplift rate in our reference models. Jargest basins. We also see a coherent evolution of this difference over the simulation time, consistent with the model progression toward a total topographic equilibrium. The spatial variability of the denudation rates is neither homogeneous nor randomly distributed (Fig. 5a6a). The location of drainage basins with denudation rates far from the equilibrium value of 0.1 mm.yr⁻¹ coincides with migrating drainage divides (Fig. 2e3d) and with cross-divide contrasts in headwater.channel x, slopegradient and elevationheight (Figs. 5b, e and 6b-d). Following Willett et al. (2014) and Forte and Whipple (2018), the divide migrations predicted by these contrasts are consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) 270 denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology 	255	For basins in excess, the MAD value decreases through time, depending on drainage area : from ca. $0.\frac{28 \text{ mm.yr}^4 25}{25}$ to ca.	
 with an area of 100 200km?. These results reflect a significant spatial variability of 10-20 km² and from ca. 0.7 to ca. 0.04 mm.yr¹ for the difference between basin-wide denudation rates and uplift rate in our reference models.largest basins. We also see a coherent evolution of this difference over the simulation time, consistent with the model progression toward a total topographic equilibrium. The spatial variability of the denudation rates is neither homogeneous nor randomly distributed (Fig. 546a). The location of drainage basins with denudation rates far from the equilibrium value of 0.1 mm.yr¹ coincides with migrating drainage divides (Fig. 2e3d) and with cross-divide contrasts in headwaterchannel <i>x</i>, slopegradient and elevationheight (Figs. 55, e and db-d). Following Willett et al. (2014) and Forte and Whipple (2018), the divide migrations predicted by these contrasts are consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology 			
Imm_yr^1 for the difference between basin wide denudation rates and uplift rate in our reference models-largest basins. We also see a coherent evolution of this difference over the simulation time, consistent with the model progression toward a totalWe atotal260topographic equilibrium.261The spatial variability of the denudation rates is neither homogeneous nor randomly distributed (Fig. Se6a). The location of drainage basins with denudation rates far from the equilibrium value of Q.1 mm.yr^1 coincides with migrating drainage divides (Fig. Se3d) and with cross-divide contrasts in headwaterchannel χ. slopegradient and elevationheight (Figs. Sb-c and 6b-d). Following Willett et al. (2014) and Forte and Whipple (2018), the divide migrations predicted by these contrasts are consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower)270270270271			
 also see a coherent evolution of this difference over the simulation time, consistent with the model progression toward a total topographic equilibrium. The spatial variability of the denudation rates is neither homogeneous nor randomly distributed (Fig. 5a6a). The location of drainage basins with denudation rates far from the equilibrium value of 0.1 mm.yr⁻¹ coincides with migrating drainage divides (Fig. 2a3d) and with cross-divide contrasts in headwaterchannel <i>x</i>, slopegradient and elevationheight (Figs. 5b, c and 6b-d). Following Willett et al. (2014) and Forte and Whipple (2018), the divide migrations predicted by these contrasts are consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) 270 denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology 			
 topographic equilibrium. The spatial variability of the denudation rates is neither homogeneous nor randomly distributed (Fig. 546a). The location of drainage basins with denudation rates far from the equilibrium value of Q.1 mm.yr⁻¹ coincides with migrating drainage divides (Fig. 2e3d) and with cross-divide contrasts in headwaterchannel <u>x</u>, slopegradient and elevationheight (Figs. 5b, e and 6b-d). Following Willett et al. (2014) and Forte and Whipple (2018), the divide migrations predicted by these contrasts are consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology 			
 The spatial variability of the denudation rates is neither homogeneous nor randomly distributed (Fig. 5a.6a). The location of drainage basins with denudation rates far from the equilibrium value of <u>0</u>,1 mm.yr⁻¹ coincides with migrating drainage divides (Fig. 2e3d) and with cross-divide contrasts in headwaterchannel <u>x</u>, slopegradient and elevationheight (Figs. 5b, c and 6b-d). Following Willett et al. (2014) and Forte and Whipple (2018), the divide migrations predicted by these contrasts are consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) 270 denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology 	260		
 drainage basins with denudation rates far from the equilibrium value of 0.1 mm.yr⁻¹ coincides with migrating drainage divides (Fig. 2e3d) and with cross-divide contrasts in headwaterchannel x, slopegradient and elevationheight (Figs. 5b, e and 6b-d). Following Willett et al. (2014) and Forte and Whipple (2018), the divide migrations predicted by these contrasts are consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) 270 denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology 	1	The spatial variability of the denudation rates is neither homogeneous nor randomly distributed (Fig. 5a6a). The location of	
divides (Fig. 2e3d) and with cross-divide contrasts in headwaterchannel χ , slopegradient and elevationheight (Figs. 5b, e and 6b-d). Following Willett et al. (2014) and Forte and Whipple (2018), the divide migrations predicted by these contrasts are consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) 270 denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology			
6b-d). Following Willett et al. (2014) and Forte and Whipple (2018), the divide migrations predicted by these contrasts are 265 consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in these 265 parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) 270 denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology			Mis en forme: Anglais/États Unic)
265 consistent with the direction of divide mobility obtained from our model. One may note that the higher the contrast in these parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) 270 denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology			
parameters across the divide, the higher the deviation of the denudation rate from the uplift rate, and therefore from topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) 270 denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology	265		
 topographic equilibrium. TheseNone of sampled basin in this data-set contains a knickpoint. Thus, these results based on simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology 	200		
 simulations assuming uniform and constant properties as well as constant boundary conditions confirm that the dispersion observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology 	I		
observed in denudation rates is primarily controlled by divide migration. Basins that expand (shrink) show higher (lower) 270 denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology			
270 denudation rates compared to uplift rate, and are hereafter referred to as aggressors (victims), following the terminology			
	270		
Audpred by whiett et al. 2014). Mis enforme: Anglais(EtatsUnis)	270		
			Mis enforme: Anglais(EtatsUnis)

	3.3 Deviation of denudation rates from the uplift rate, and basin aggressivity	
	Willett et al. (2014) showed that the basin-averaged cross-divide contrast in χ , could be used to deduce an aggressivity	Mis en forme: A nglais(États U nis)
I	metric for basins. We extend this basin-scale approach to the Gilbert's metrics recently proposed by Forte and Whipple	
275	(2018) including cross-divide contrast in headwater slope and elevation. Hereinafter, we refer to these aggressivity metrics-	
	based on cross-divide contrast in headwater χ , headwater slope (gradient) and headwater elevation (height) as 4χ , $4G$ and	
	<u>4H-gradient and elevationrespectively.</u>	
	We here assess the relationship between the E/U ratio and these aggressivity metrics. First, to exclude variability related to	Mis en forme: Anglais(États Unis)
	both basin area and time, we focus on a single class of basins with a size of $\frac{10 \cdot 202 \cdot 4}{10 \cdot 202 \cdot 4}$ km ² gathered from five computed	Mis enforme: Anglais(ÉtatsUnis)
280	reference models after a simulation duration of 2.5 Myr. Denudation rates may be affected by knickpoints, which are a	
	source of transient perturbation at the scale of the catchment. Therefore, in order to focus only on perturbations associated	
	with drainage divide dynamics, basins that contain knickpoints are ignored. In agreement with cross-divide metrics tested by	
	Forte and Whipple (2018), aggressorgraphs in figure 7 must be divided into four quadrants. Aggressor (victim) basins have	
	negative (positive) $\frac{4\chi}{\chi_{av}}$ and $\frac{4H\Delta H_{av}}{H_{av}}$ values and conversely positive (negative) $\frac{4G\Delta G_{av}}{G_{av}}$ value: (Fig. 2). Theoretically,	
285	aggressor (victim) basins have higher (lower) denudation rates than the underlying uplift rate. Therefore graphs in figure 6	
	must be divided into four quadrants, with aggressors situated in the lower-left one, and victims in the higher-right one. This	
	result is verified for ca. $91, 8881$ %, 52 % and 8281 % of basins for aggressivity metric based on headwater χ , slope and	
	elevation values, $\Delta \chi_{av} \Delta G_{av}$ and ΔH_{av} respectively (Figs. 6a, b and c). Several basins depart significantly from the expected	
	quadrants for 4G-and 4H: these exhibit significant knickpoints in their drainage network that increase measured denudation	
290	rates. For this limited dataset, the evolution between E/U and both $\frac{4}{\sqrt{2}}\chi_{av}$ and $\frac{4G}{2} can \Delta H_{av}$ may be defined by a linear	Mis en forme: Anglais(ÉtatsUnis)
	relationship (Figs. 6a and b). Part of the dispersion observed around this first order trend may be explained by	Mis enforme: Anglais(ÉtatsUnis)
	approximations in the calculation of denudation rates and aggressivity metrics.Fig. 7b). Compared to other metrics, 4H	
	seems to be ΔG_{av} is less sensitive to drainage migration and shows a more scattered distribution (Fig. 6c).	
I	In natural settings, the stage of evolution of landscapes cannot be easily defined and the total amount of basins with a	
295	specific size may be limited. The large dataset provided by from our model modeling can provide further insights by gathering	
1	the results obtained every 0.5 Myr for seven classes of basin areas expanding geometrically with a multiplying factor of 2	
	from 1-2 to 64-128 km ² (Figs. 6d, e and f). Denudation rates can be affected by knickpoints, which are an additional source	
	of transient perturbation at the seale of the eatchment. Therefore, in order to focus on perturbations associated with drainage	
	divide dynamics, basins that contain knickpoints are ignored. 7b). Basins that contain knickpoints are discarded from the	
300	analysis. Our results highlight the major control of basin size on the dispersion E/U. When all classes of drainage areas are	Mis enforme: Anglais(ÉtatsUnis)
	combined together, we still obtain a clear relationship between aggressivity metrics and E/U , with 77 %, 56 % and 78 % of	Mis en forme: A nglais(États U nis)
	basins lying in aggressors or victims quadrants for $\Delta \chi_{av}$, ΔG_{av} and ΔH_{av} , respectively (Fig 7b). Our results highlight the	
	major control of basin size on the dispersion E/U_2A_{χ} and ΔG and E/U (Fig 6d and c). One may note however Part of the	Mis en forme: A nglais(États U nis)
	variability intrinsic for each class of basin area may in turn be explained by heterogeneities in aggressivity between different	Mis en forme: A nglais(États U nis)
I	10	

- 305 parts of a basin. Figure 7c shows that this dispersion is related to the standard deviation of aggressivity metrics. $\Delta \chi_{std} \Delta G_{std}$ and ΔH_{std} . In other words, basins where different divide segments migrate at different rates or in different directions are more scattered. The lower the confidence index, the more scattered the results (Fig. 7d). Thus, some dispersion may come from approximations due to undocumented divide segments performed when averaging metrics differences between reference basins (Fig. 2). One may note two different trends for victim and aggressor basins. Aggressors show a more
- scattered distribution for $4 \chi \Delta \chi_{av}$ and $4 G \Delta G_{av}$ metrics. When compared to victims, these basins have hillslopes closer to the critical value *Sc* (Fig. S3). Hence, the dispersion may be explained by the non-linear relationship existing between denudation rates and basin slope (Montgomery and Brandon, 2002; Binnie et al., 2015). Therefore, a simple linear trend is no longer sufficient to properly fit our results. Hence we consider victim and aggressor basins independently assuming two

Mis en forme: Anglais(ÉtatsUnis)

315 4 Discussion

4.1 Sensitivity tests

linear trends between-E/U-and each aggressivity metrie.

The reference model involves various parameters related to uplift, <u>fluvial incision and</u> hillslope denudation-and fluvial incision. A systematic analysis of all parameters-trade-off between allparameters is out of the scope of this manuscript. Here, we do not investigate the effect of erodibility, because it remains poorly constrained in nature and because our study is

320 mostly focused on channel head properties. In this section, we assess the sensitivity of the results to both tectonic and hillslopeerosion processes, by studying the specific impact of uplift <u>U</u>, erodibility <u>K</u>, diffusivity <u>D</u> and critical hillslope gradient <u>Sc</u> taken separately. Varying these parameters may change the simulation time required to erode the plateau associated with the initial boundary conditions. In this section, to reduce sensitivity dependence on these initial conditions. We only consider results obtained between 5 Ma and 10 Ma.

325 4.1.1 Sensitivity to uplift rate

We vary the tectonictest rock uplift rate from rates of 0.501 mm.yr^{-1} , 0.1 mm.yr^{-1} (hereafter called reference model) and 1 mm.yr^{-1} to 2 mm.yr^{-1} , which is incover the range observed in moderately active mountain belts; e.g. the Alborz, the Alps or the Caucasus of a large variety of geodynamic settings (Champagnac et al., 2009; Djamour et al., 2010; Vincent et al., 2011), 2012). It is well-known that a river responds to a fall in base level (due to changes in rock uplift rate or other forcing)

by cutting downward into its bed, deepening and widening its active channel. In our simulations, changes in uplift rate lead to variations in the geometrydensity of the drainage network. Compared to the reference model, an uplift rate of 21 mm.yr⁻¹ (0.501 mm.yr⁻¹) results in an a decrease (increase (decrease) of river channelization - inversely proportional to drainage density - which induces larger (smaller) river basins, a lower (higher) range of values for 4χ and 4G and a higher (lower) range of values for 4H (Fig. 7). More importantly, our simulations suggest a linear. These results are consistent with

Mis enforme: Exposant

- previous studies that show an inverse relationship between tectonic upliftdrainage density and denudationerosion rates. As a 335 result, assuming no climate tectonic feedback we obtain no significant changes in equilibrium topography when using a threshold slope for diffusion processes (Tucker and Bras, 1998; Clubb et al., 2016). An increase in uplift rate favors river entrenchment leading to increase the range of ΔG_{av} and ΔH_{av} (Fig. 8). relationship between the calculated aggressivityHence, these two Gilbert's metrics and the E/U-ratio for appear to be well suited to diagnose local disequilibrium 340
- for higher uplift rates ranging between 0.5(i.e. ≥ 1 mm.yr⁻¹ and 2 mm.yr⁻¹. Conversely increase uplift rate induces a lower range of values for $\Delta \chi_{av}$. This last observation is explained by the decrease of drainage density, and associated stream length.

<u>Maximum variability of E/U reaches a factor regardless the assumed uplift rate between 1 mm.yr⁻¹ and 0.01 mm.yr⁻¹. The</u> observed small differences suggest that limited uplift rate promote diffusive processes (see Sect. 4.1.3).

345 4.1.2 Influence of erodibility

Fluvial erosion is proportional to the erodibility coefficient K that may reflect, among others, rock strength and climate. We let this parameter vary between 5.10^6 m^(1-2m), yr⁻¹ and 5.10^5 m^(1-2m), yr⁻¹. As expected from (Eq.1 and 3), we find that erodibility and uplift rates have opposite effects. Lower (higher) values of erodibility lead to higher (lower) average topography. Thus, an increase (decrease) in erodibility decreases (increases) the range of all aggressivity metrics (Fig. 9).

350

Lower values of erodibility also increase the range of the E/U ratio. Models with higher (lower) erodibility reach a quasitopographic steady state earlier (at a later stage). Hence, differences in the variability of E/U may be related to different stages of evolution for each models over the period we consider (5 to 10 Ma) (Fig. 5d).

4.1.3 Influence of hillslope processes

	Hillslope denudation is proportional to the diffusivity coefficient <i>D</i> and depends on the critical slope SeSc (Eq. 2). To test Mis enforme: Anglais(ÉtatsUnis)
355	the effect of hillslope processes, we first varylet p vary between 10^{-3} and 10^{-4} m ² .yr ⁻¹ . We find and 10^{-1} m ² .yr ⁻¹ . Compared to Mis enforme: Anglais(ÉtatsUnis)
	the reference model, we find no major difference with the reference model over that range of values for D (Fig. differences
	in the case of a lower diffusivity (i.e. 0.001 m ² .yr ⁻¹) (Fig. 10c). In contrast, for models with higher diffusivity coefficient (i.e.
	0.1 m ² .yr ⁻¹), this parameter has a significant effect on both the range of E/U and the aggressivity metric ΔG_{av} (Fig. 10a).
	This result derives from a stronger impact of diffusive processes in the upstream parts of the drainage network, consistent
360	with the observations described in section 4.1.1.8)-Denudation intensity also varies inversely to the square of the eritical
	hillslope gradient Se (Eq. 2). Assuming a critical slope between 20° and 40°, we find that Sc is a parameter influencing (Mis enforme: Anglais(ÉtatsUnis)
	aggressivity metrics (Fig. 9). As in the case of variations in uplift rate, changes in Se lead to differences in the organization
	of the drainage network. Compared to the reference model, a low Sc leads to large river channelization, reducing the range
	of all aggressivity metrics, but does not affect significantly the relationship between the <i>E/U</i> ratio and the studied metrics. Mis enforme: Anglais(ÉtatsUnis)
I	Mis en forme: A nglais(États U nis)

365 These observations are consistent with a landscape where hillslopes are completely determined by the assumed critical slope value. (Fig.11).

Altogether, these sensitivity tests demonstrate the robustness of our findings: $\frac{4}{3}$ and $\frac{46}{4}$. Regardless of the tested parameter values, we observe a relationship between aggressivity metrics and deviation of denudation rates from uplift rates. Thus,

370 aggressivity metrics are, to the first-order, reliable metrics to assess the effect of divide mobility on basin-wide denudation rates inferred from simulations. In the following section, we apply this approach to field observations and discuss the consequences for sampling and interpretation.

4.2 Implication Implications for the interpretation of basin-wide denudation rate interpretation rates

Over the last decades, measurements of cosmogenic radionuclides (CRN) concentration in alluvial sediments (see Granger et 375 al., 2013 and references therein), of suspended sediments (Gabet et al., 2008) and of detrital thermochronology (Huntington and Hodges, 2006) have become common practices to assess basin-wide denudation rates. However, their interpretation remains debated, even in settings where topographic steady state is supposedly achieved regionally.

4.2.1 Application to the Great Smoky Mountains

As previously mentioned (Matmon et al., 2003a, b), while the Great Smoky Mountains in the southern Appalachians are 380 expected to be in a quasi-topographic steady state, basin-wide denudation rates show a strong dispersion up to a factor of 2two in comparison withto the estimated uplift rate (0.025 toca. 0.03 mm.yr⁻¹, see Fig. 1). Basins in We use the Great Smoky Mountains exhibit average slope values ranging between 20 and 30° (Matmon et al., 2003a). This may suggest that hillslopes of repose. On the basis of the consider that our approach is applicabl the<u>data associated with 40</u> basins studied originally sampled by Matmon et al., (2003a, b), even though estimated uplift rates 385 differ by nearly two orders of magnitude when compared to the uplift rates used in our models. We use the data associated with 40 basins originally sampled by Matmon et al., (2003a) and for which denudation rates were re-calculated by Portenga and Bierman (2011). Following our method, we calculate the three basin-averaged aggressivity metrics $\frac{4\chi}{4G} \frac{4G}{\chi_{aux}} \Delta G_{aux}$ and $\Delta H\Delta H_{av}$ associated with these 40 catch ments (Fig. 1012; See also Fig. S4S3). The calculation of χ requires to define the elevation of the catchment outlets H_b and the m/n ratio (Eq. 4). As underlined by Forte and Whipple (2018), the choice of

Mis enforme: Anglais(ÉtatsUnis)

Mis enforme: Anglais(ÉtatsUnis) Mis enforme: Anglais(ÉtatsUnis)

Mis enforme: Anglais(ÉtatsUnis)

Mis enforme: Anglais(ÉtatsUnis)

Mis enforme: Anglais(ÉtatsUnis)

Mis en forme: Anglais(ÉtatsUnis)

Mis enforme: Anglais(ÉtatsUnis)

Mis enforme: Anglais(ÉtatsUnis)

- 390 the "correct" outlet elevation is non-trivial in natural settings. We first consider a local base level given by the Tennessee river<u>River</u>. To test the relevance of this choice, we also testuse a base level located at a fixed <u>arbitrary</u> elevation $H_b = 400$ m. We use assume the same m/n ratio value of 0.45 as used in the by Willett et al. (2014) study for the Great Smoky Mountains. For all calculated metrics, the majority (ca. 7558% for $4G\Delta G_{av}$ and 55ca. 66% for $4H\gamma\Delta H_{av}$) of the basins are located in the expected quadrants (see Fig. 7). However, more attention must be given to the results based on 4χ , $\Delta \chi_{av}$. For this metric, 62ca. 58% of the analyzed basins lie in the expected quadrant when we consider the Tennessee river as the local base level
- 395

versus $-\frac{67ca}{68}$ for $H_b = 400$ m (Fig. 10). If (12b). A lthough the general result remains overall results are similar here, we show that the choice of a different base level H_b leads to significant variations in $4\chi \Delta \chi_{av}$ for individual basins. This highlights the main weakness of the $4\chi \Delta \chi_{av}$ metric, which is highly sensitive to the choice of the proper base level H_b . Nevertheless, our results confirm the findings of Willett et al. (2014), suggesting that a significant part of the data variance

- 400 observed in the Matmon et al. (2003a, b) can be explained by divide migration (Fig. 1012), raising this possible explanation for the variability of most natural data-sets. One may note that the Southern Appalachians exhibit migrating knickpoints that can locally affect denudation rates (Gallen et al., 2011-; Gallen et al., 2013). This last point can also explain part of the observed variability in this dataset but this specific impact is beyond the scope of the present <u>manuscriptstudy</u>.
- Based on both our simulations and this field dataset, we propose to favor the use of the Gilbert metric ΔG based on the crossdivide contrast in channel head local slope (Forte $\Delta \chi_{av}$ and Whipple, 2018). ΔH_{av} . Among the tested metrics, $\Delta H \Delta G_{av}$ appears the least sensitive to disequilibrium, and $\Delta \chi$ requires better constraining and defining objective criteria for H_{π} -excepted in active mountain belts with rock uplift U > 1 mm.yr⁻¹.

4.3.2 Assessment of topographic disequilibrium

410 Topographic steady-state is a very convenient assumption and concept to deduce the uplift pattern in mountains ranges from denudation rates, and thus to obtain significant information on the geometry of active structures and on orogen dynamics (Lavé and Avouac, 2001, Godard et al., 2014; Scherler et al., 2014; Le Roux-Mallouf et al., 2015). However, this assumption is seldom verified at the scale of sampled watersheds.

On the basis of our modeling, we show that the competition between low-order basins has a significant impact on basin-wide denudation rates. The proposed approach provides a new tool to assess the potential maximum deviation from topographic steady state based on aggressivity metrics and drainage area, which can both be inferred from a simple DEM. The: the closer to zero the aggressivity metrics and the lower the standard deviation of cross-divide metrics, the more representative toof

uplift rate are the measured denudation rates. However it still remains dispersion of denudation rates values, especially for smaller catchments.
 Based on our sensitivity tests for moderately active orogens (with uplift rates between 0.5 and 2 mm/yr), the empirical

relationship between ΔG and E/U obtained from the reference model (Fig. 6) can be used to assess the topographic disequilibrium of basins. Especially for victim basins ($\Delta G < 0$), this relationship exhibits a linear relationship:

 $\frac{-4}{2} = 0.03\Delta G$ for basin area > 50 km²-

-((

Mis	en forme: Anglais(ÉtatsUnis)
Mis	en forme: Anglais(ÉtatsUnis)
Mis	enforme: Anglais(ÉtatsUnis)

For the sake of simplicity our models involve spatially homogenous and time invariant parameters. Additional simulations are now needed to test this approach in more complex settings, including spatial and temporal variability in climate and tectonic forcing or internal landscape parameters like erodibility.

4.4.3 Improvement of sampling strategy

- 430 Basin-wide denudation rates obtained from CRN concentration measurements, suspended sediments or detrital thermochronology depend on many parameters including lithology, ice cover, rainfall, landslide activity or tectonic uplift (Vance et al., 2003; Bierman and Nichols, 2004; Wittmann et al., 2007; Yanites et al., 2009; Norton et al., 2010; Godard et al., 2012; Whipp and Elhers, 2019). Hence, to unravel the influence of tectonics from other processes, a specific sampling strategy is usually recommended: (1) to sample catchments with homogeneous lithologies to limit the effect of spatial
- 435 variations in the abundance of target minerals in bedrock formations; (2) to select catchments with no ice cover (past or present) because the input of glacier-derived sediments can significantly complicate the interpretation of CRN concentrations; (3) to choose areas with spatially uniform rainfall distribution; and (4) to consider watersheds where the relative contribution of landslides to long-term landscape evolution is low. Unfortunately, these different criteria imply to select watersheds with variable sizes. The first three criteria favor the sampling of small catchments, whereas the last one
- 440 requires basins large enough to be less affected by landslides.-Indeed, Niemi et al. (2005) proposed that mixing effects efficiently dampen the stochastic nature of hillslope sediment delivery by landsliding above a critical catchment area. Considering an uplift rate of 0.5-2 mm.yr⁴, the recommended minimum area needed to mitigate these biases associated with the stochastic input from landslides is of 50 to 200 km².

Our approach suggests the need to pre-assess targeted basins for their potential divide mobility before sampling for CRN

445 concentration measurements. If the objective is to quantify the background uplift rate, one should sample basins that satisfy the conditions we previously enounced in the current section and also display an aggressivity close to zero and with the smallest associated standard deviation. Conversely, to quantify the specific denudation rate associated with the migration of drainage divides, small aggressor or victim basins should be favored.

Based on our simulations, a relationship between <u>the</u> maximum of erosion variability (0.5 and 99.5 percentiles, respectively) due to divide mobility $\left[(E - U)/U \right]_{max}$ and the catchment size A can be derived (Fig. <u>113</u>). Our results suggest a logarithm dependence between these two parameters, regardless of the assumed U, K, D and $S_c \rightarrow 0$.

$$\left[(E - U)/U \right]_{max} = c_1 \log(A) + c_2 \text{ for } -1 \text{ km}^2 < A < 100 \text{ km}^2 ,$$

(7)

455 For victim basins ($\Delta G < 0$), $c_1 \simeq 0.05$ and $c_2 \simeq -0.5$, whereas for aggressor basins $c_1 \simeq -0.14$ and $c_2 \simeq 1$. This provides a new additional guideline for the design of sampling strategies in terms of basin size. For instance, considering a quasisteady state mountain belt with an uplift rate of 1 mm.yr⁴, the minimum basin area required for an erosion rate variability

 Mis en forme: Anglais(ÉtatsUnis)

 Mis en forme: Anglais(ÉtatsUnis)

lower than 0.5 mm.yr^{\pm} is ca. 1 km² for victim basins and 30-40 km² for aggressor basins.with c_1 and c_2 two parameters that depend on balance between erosion processes, uplift rate and state of evolution of the landscape.

460 5 Conclusions

465

Calculations from a Landscape Evolution Model assuming spatially uniform uplift, rock strength and rainfall confirm that the concept of topographic steady state is relevant at the scale of entire mountain belts, but represents an oversimplification at the scale of individual watersheds. Our simulations underline the role of divide mobility on deviations from equilibrium, which can lead to significant differences between tectonic uplift rate and basin-wide denudation rates even if an overall topographic steady state is achieved at large scale.

To better assess these deviations, we propose new basin-averaged aggressivity metrics $-\Delta \chi_{av} \Delta G_{av}$ and ΔH_{av} based on the approach of Willett et al. (2014) and Forte and Whipple (2018). They include mean cross-divide contrasts in channel-heads χ_{a} local slope gradient and elevation at the scale of entire river basins-height. From our calculations, contrasts in channel head elevation appear $\Delta \chi_{av}$ is the most reliable aggressivity metric to be weakly sensitive to assess local disequilibrium, whereas,

- 470 but is highly depend on the basin denudation to chosen base level, which remains hard to constrain. Gilbert's metrics ΔG_{av} and ΔH_{av} are more suitable for relatively high uplift ratio E/U-exhibits a nonlinear relationship with $\Delta \chi$ and ΔG . Together, these two rate (i.e. ≥ 1 mm.yr-1). Altogether, our metrics reveal that E/U deviation of denudation rates from uplift rate related to divide migrations depends on both basin aggressivity and basin area. This last parameter has a key control on the dispersion in E/U, which can reach a factor of two, regardless of the imposed uplift rate (here 0.5-2 mm.01-1 mm.yr⁻¹).
- 475 erodibility (here $5.10^{6} 5.10^{5}$ m^(1-2m), diffusivity (here $10^{-3} 10^{-1}$ m².yr⁻¹) and hillslope gradient (here $20^{\circ} 40^{\circ}$). By comparing our results to CRN measurements from the Great Smoky Mountains (Matmon, 2003a, b), we show that this approach can be used to improve field sampling strategies and provides a new tool to derive a minimal uncertainty in basin-wide denudation rates due to topographic disequilibrium.

For the sake of simplicity our models involve spatially homogenous and time invariant parameters. Additional simulations
 are now needed to test this approach in more complex settings, including spatial and temporal variability in climate and tectonic forcing or parameters like stream power equation exponents *n* and *m*.

Ack nowledgments

485

We are greatly indebted to referees Fiona Clubb and Adam Forte for providing constructive reviews that significantly improve the quality of the manuscript. We thank Wolfgang Schwanghart and Benjamin Campforts for providing the TopoToolBox and TTLEM codes to analyze and to simulate landscape evolution. Timothée Sassolas-Serrayet's Ph.D. is supported by a fellowship from the French Ministry for Higher Education. We thank Wolfgang Schwanghart and Benjamin Campforts for providing the programme TopoToolBox and TTLEM to analyze and to simulate landscape evolution. We

Mis en forme: Anglais(Roy aume-Un) acknowledge funding of the French Agence National de la Recherche, grant ANR-18-CE01-0017 (Topo-Extreme).
ThisThrough the work of Martine Simoes, this study contributes to the IdEx Université de Paris ANR-18-IDEX-0001. This, and is IPGP contribution # XXX².

References

	Binnie, S. A., Phillips, W. M., Summerfield, M. A., and Fifield, L. K.: Tectonic uplift, threshold hillslopes, and denudation rates in a developing mountain range. Geology 35(8), 743-746, 2007.	 Mis enforme: Police :+Titres(Times NewRoman)
	Bierman, P., and Nichols, K.K.: Rock to sediment slope to sea with 10 Be rates of landscape change. Annual Review of	
495	Earth and Planetary Sciences 32, 215-255, 2004.	
	Blöthe, J. H., Korup, O., and Schwanghart, W.: Large landslides lie low: Excess topography in the Himalaya-Karakoram	
	ranges, Geology, 43, 523–526, doi:10.1130/G36527.1, 2015.	
	Burbank, D. W., Leland, J., Fielding, E., Anderson, R. S., Brozovic, N., Reid, M. R., and Duncan, C.: Bedrock incision, rock	 Mis enforme: Police :+Titres(Times
	uplift and threshold hillslopes in the northwestern Himalayas. Nature 379(6565), 505, 1996.	NewRoman)
500	Champagnac, J.D., Schlunegger, F., Norton, K., von Blanckenburg, F., Abbühl, L.M. and Schwab, M.: Erosion-driven uplift	
	of the modern Central Alps. Tectonophysics 474(1-2), 236-249, 2009. Molnar, P., Sue, C., Herman, F.: Tectonics, climate	
	and mountain topography. Journal of geophysical research, 117, B02403, doi:10.1029/2011JB008348, 2012.	 Mis enforme: Police :+Titres(Times
	Campforts, B., and Govers, G.: Keeping the edge: A numerical method that avoids knickpoint smearing when solving the	NewRoman)
	stream power law. Journal of Geophysical Research: Earth Surface, 120(7), 1189-1205, 2015.	
505	Campforts, B., Schwanghart, W., and Govers, G.: Accurate simulation of transient landscape evolution by eliminating	
	numerical diffusion: the TTLEM 1.0 model. Earth Surface Dynamics 5, 47-66, 2017.	
	Djamour, Y., Vernant, P., Bayer, R., Nankali, H.R., Ritz, J.F., Hinderer, J., Hatam, Y., Luck, B., Le Moigne, N., Sedighi, M.	
	and Khorrami, F.: GPS and gravity constraints on continental deformation in the Alborz mountain range, Iran. Geophysical	
	Journal International 183(3), 1287-1301, 2010.	
510	Clubb, F. J., Mudd S. M., Attal M., Milodowski D. T., and GrieveS. W. D.: The relationship between drainage density,	
	erosion rate, and hilltop curvature: Implications for sediment transport processes, J. Geophys. Res. Earth Surf., 121, 1724-	
	<u>1745. doi:10.1002/2015JF003747. 2016.</u>	
	Culling, W.E.H.: Soil Creep and the Development of HillsideSlopes, J. Geol., 71, 127-161, doi:10.1086/626891, 1963.	
	Dahlquist, M.P., West, A.J., & Li, G.: Landslide-driven drainage divide migration. Geology, 46(5), 403-406.	
515	<u>doi:10.1130/g39916.1, 2018.</u>	
	England, P. and Molnar, P.: Surface uplift, uplift of rocks, and exhumation of rocks. Geology 18, 1173-1177, 1990.	 Mis enforme: Police :+Titres(Times
	Forte, A.M. and Whipple, K.X.: Criteria and tools for determining drainage divide stability. Earth and Planetary Science	NewRoman)
	Letters 493, 102-117, 2018.	

ĺ	Gallet Gallen, S.F., Wegmann, K.W., Frankel, K.L., Hughes, S., Lewis, R.Q., Lyons, N., Paris, P., Ross, K., Bauer, J.B., and		Mis enforme: Police :+Titres(Times
520	Witt, A.C.: Hillslope response to knickpoint migration in the Southern Appalachians: implications for the evolution of post-		NewRoman)
	orogenic landscapes. Earth Surface Processes Landforms 36, 1254-1267, 2011.		Mis enforme: Police :+Titres(Times
	Gallet Gallen, S. F., Wegmann, K. W., and Bohnenstiehl, D.R.: Miocene rejuvenation of topographic relief in the southern		NewRoman)
	Appalachians. GSA Today 23, no.2, 2013.		Mis enforme: Police :+Titres(Times NewRoman)
	Gabet, E. J., Burbank, D. W., Pratt-Sitaula, B., Putkonen, J., and Bookhagen, B.: Modern erosion rates in the High		
525	Himalayas of Nepal. Earth and Planetary Science Letters 267(3-4), 482-494,-https://doi.org/10.1016/j.eps1.2007.11.059,		Mis enforme: Police :+Titres(Times
	2008.		NewRoman)
	Gilbert GK .: Report on the Geology of the Henry Mountains. USGS Report, Government Printing Office, Washington, D.C.,	\sim	' Mis en forme: Police :+Titres(Times NewRoman)
	1877.	\ \	Mis enforme: Police :+Titres(Times
	Godard, V., Lavé, J., and Cattin, R.: Numerical modelling of erosion processes in the Himalayas of Nepal: Effects of spatial		NewRoman)
530	variations of rock strength and precipitation. Geological Society, London, Special Publications 253(1), 341-358, 2006.		
	Godard, V., Burbank, D.W., Bourlès, D.L., Bookhagen, B., Braucher, R. and Fisher, G.B.: Impact of glacial erosion on 10Be		
	concentrations in fluvial sediments of the Marsyandi catchment, central Nepal. Journal of Geophysical Research: Earth		
	Surface 117(F3), 2012.		Mis enforme: Police :+Titres(Times
	Godard, V., Bourlès, D.L., Spinabella, F., Burbank, D.W., Bookhagen, B., Fisher, G.B., Moulin, A. and Léanni, L.:		New Roman), Anglais (États Unis)
535	Dominance of tectonics over climate in Himalayan denudation. Geology 42(3), 243-246, 2014.		Mis enforme: Police :+Titres(Times NewRoman), Anglais(ÉtatsUnis)
	Granger, D.E., Lifton, N.A., and Willenbring, J.K.: A cosmic trip: 25 years of cosmogenic nuclides in geology. GSA		Mis enforme: Police :+Titres(Times
	Bulletin 125(9-10), 1379-1402, 2013.		NewRoman), Anglais(ÉtatsUnis) Mis en forme: Police :+ Titres(Times)
	Harel, MA., Mudd, S.M. and Attal, M.: Global analysis of the stream power law parameters based on worldwide ¹⁰ Be		NewRoman)
	denudation rates. Geomorphology 268, 184-196, https://doi.org/10.1016/j.geomorph.2016.05.035, 2016.	V_{-}	Mis enforme: Police :+Titres(Times NewRoman)
540	Hasbargen, L.E. and Paola, C.: Landscape instability in an experimental drainage basin. Geology 28, 1067-1070, 2000.		Mis enforme: Police :+Titres(Times
	Hasbargen, L.E. and Paola, C.: How predictable is local erosion rates in eroding landscapes? Prediction in Geomorphology		NewRoman)
	135, 231-240, https://doi.org/10.1029/135GM16, 2003.	K	Mis enforme: Police :+Titres(Times NewRoman)
	Huntington, K. W., and Hodges, K. V.: A comparative study of detrital mineral and bedrock age-elevation methods for		Mis enforme: Police :+Titres(Times
	estimating erosion rates. Journal of Geophysical Research: Earth Surface 111(3), 1-11,		NewRoman) Mis en forme: Police :+Titres
545	https://doi.org/10.1029/2005JF000454, 2006.	-	NewRoman)
	Jungers, M.C., Bierman, P.R., Matmon, A., Nichols, K., Larsen, J. and Finkel, R.: Tracing hillslope sediment production and		Mis enforme: Police :+Titres(Times NewRoman)
	transport with in situ and meteoric 10Be. Journal of Geophysical Research: Earth Surface 114, F04020,	N	Mis enforme: Police :+Titres(Times
	https://doi.org/10.1029/2008JF001086, 2009.	1. A.	NewRoman)
	Hack JT.: Interpretation of erosional topography in humid temperate regions. American Journal of Science 258-A, 80-97,		Mis en for me: Police :+Titres(Times NewRoman)
550	1960.		Mis enforme: Police :+Titres(Times
	Kirby, E., and Whipple, K.: Ouantifying differential rock-uplift rates via stream profile analysis, Geology 29(5), 415-418.	- N	NewRoman)

Mis enforme: Police :+Titres(Times

NewRoman)

Kirby, E., and Whipple, K.: Quantifying differential rock-uplift rates via stream profile analysis. Geology 29(5), 415-418, 2001.

Lavé, J., and Avouac, J.P.: Fluvial incision and tectonic uplift across the Himalayas of central Nepal. Journal of Geophysical Research 106(B11), 26,561-26,591, 2001.

Mis en forme: Police :+Titres(Times

555 Le Roux-Mallouf, R., Godard, V., Cattin, R., Ferry, M., Gyeltshen, J., Ritz, J.F., Drupka, D., Guillou, V., Amold, M.,

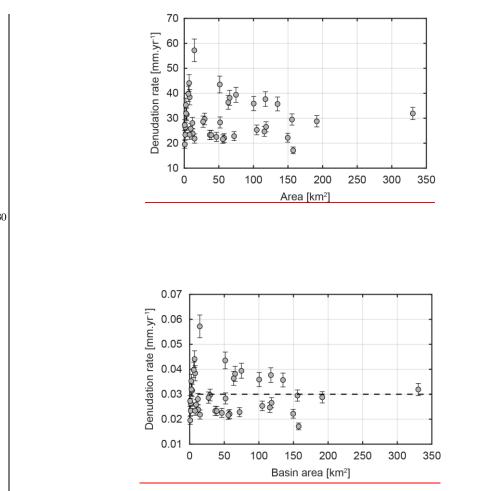
	Aumaître, G. and Bourles, D.L.: Evidence for a wide and gently dipping Main Himalayan Thrust in western Bhutan.		(NewRoman)
	Geophysical Research Letters 42(9), 3257-3265, 2015.		
	Matmon, A., Bierman, P.R., Larsen, J., Southworth, S. Pavich, M., Kinkel, R., and Caffee, M.: Erosion of an ancient		
	mountain range, the Great Smoky Mountains, North Carolina and Tennessee. American Journal of Science 303, 817-855,		
560	2003a.		
500	Matmon, A., Bierman, P.R., Larsen, J., Southworth, S. Pavich, M., Caffee, M.: Temporally and spatially uniform rates of		
	erosion in the southern Appalachian Great Smoky Mountains. Geology 31, 155-158, 2003b.		
	Miller, S.R., Sak, P.B., Kirby, E., and Bierman, P.R.: Neogene rejuvenation of central Appalachian topography: Evidence for differencial rock uplift from stream profiles and erosion rates. Earth and Planetary Science Letters 369-370, 1-12,		
ECE			
565	https://doi.org/10.1016/j.eps1.2013.04.007, 2013.	<	Mis enforme: Police :+Titre≰Times NewRoman)
	Montgomery, D. R., and Foufoula-Georgiou, E.: Channel network source representation using digital elevation models. Water Resources Research 29(12), 3925-3934, 1993.		Mis enforme: Police :+Titres(Times
			NewRoman)
	Montgomery, D.R.: Slope distributions, threshold hillslopes, and steady-state topography. American Journal of science		Mis enforme: Police :+Titres(Times NewRoman)
670	301(4-5), 432-454, 2001.		Mis enforme: Police :+Titres(Times NewRoman)
570	Montgomery, D. R., and Brandon, M. T.: Topographic controls on erosion rates in tectonically active mountain ranges. Earth		(New Rollian)
	and Planetary Science Letters 201(3-4), 481-489, 2002.		
	Niemi, N.A., Oskin, M., Burbank, D.W., Heimsath, A.M. and Cabet, E.J.: Effects of bedrock landslides on cosmogenically		
	determined erosion rates. Earth and Planetary Science Letters 237(3-4), 480-498, 2005.		
	Norton, K.P., von Blanckenburg, F., and Kubik, P.W.: Cosmogenic nuclide-derived rates of diffusive and episodic erosion in		Mis enforme: Police :+Titres(Times NewRoman)
575	the glacially sculpted upper Rhone Valley, Swiss Alps, Earth Surf. Processes Landforms 35(6), 651–662, 2010.		
	Pelletier, J. D.: Persistent drainage migration in a numerical landscape evolution model. Geophysical Research Letters		
	31(20), L20501, 2004.		
	Pelletier, J. D.: Quantitative Modeling of Earth Surface Processes, Cambridge University Press, Cambridge, UK, 2008.		
	Perne, M., Covington, M.D., Thaler, E.A. and Myre, J.M.: Steady state, erosional continuity, and the topography of		Mis enforme: Police :+Titres(Times NewRoman)
580	landscapes developed in layered rocks. Earth Surface Dynamics 5(1), 85-100, 2017.		(NewRollian)
	Perron, J.T.: Numerical methods for nonlinear hillslope transport laws. Journal of Geophysical Research: Earth Surface 116,		
	F02021, https://doi.org/10.1029/2010JF001801, 2011.		Mis enforme: Police :+Titres(Times
	Perron, J. T., and Royden, L.: An integral approach to bedrock river profile analysis. Earth Surface Processes and Landforms	11	NewRoman) Mis enforme: Police :+Titres(Times
	38, 570-576, 2012.		NewRoman)
585	Portenga, E.W. and Bierman, P.R.: Understanding Earth's eroding surface with 10 Be. GSA today 21, 4-10, 2011.	· · · · ·	Mis enforme: Police :+Titres(Times NewRoman)

	Reiners, P. W., Ehlers, T. A., Mitchell, S. G., and Montgomery, D. R.: Coupled spatial variations in precipitation and long- term emotion rates correct the Weschington Coscodes. Nature 426, 645, 647, 2002		
	term erosion rates across the Washington Cascades. Nature 426, 645-647, 2003.		
	Reinhardt, L. and Ellis, M.A.: The emergence of topographic steady state in a perpetually dynamic self-organized critical landscape. Water Resources Research 51, 4986-5003, 2015.		" Mis en forme: Police :+Titre≰Times NewRoman)
590	Richardson, P.W., Perron, J.T., and Schurr, N.D.: Influences of climate and life on hillslope sediment transport. Geology 47,		
	1–4, <u>https://doi.org/10.1130/G45305.1</u> , 2019.		Mis enforme: Police :+TitregTimes
	Roering, J.J., Kirchner, J.W., and Dietrich, W.E.: Evidence for nonlinear, diffusive sediment transport on hillslopes and	$\langle \cdot \rangle$	NewRoman)
	implications for landscape morphology. Water Resources Research 35, 853-870, 1999.	\mathbb{N}	[™] Mis enforme: Police :+Titre≰Times NewRoman)
	Royden, L., and Perron, J.T.: Solutions of the stream power equation and application to the evolution of river longitudinal	Ň	Mis enforme: Police :+Titres(Times
595	profiles. Journal of Geophysical Research: Earth Surface 118(2), 497-518, 2013.		NewRoman)
	Scherler, D., Bookhagen, B. and Strecker, M.R.: Tectonic control on 10Be-derived erosion rates in the Garhwal Himalaya,		Mis enforme: Police :+TitregTimes
	India. Journal of Geophysical Research: Earth Surface 119(2), 83-105, 2014.		NewRoman)
	Schwanghart, W., and Scherler, D.: Short Communication: TopoToolbox 2 - MATLAB-based software for topographic		
	analysis and modeling in Earth surface sciences. Earth Surface Dynamics 2, 1-7, 2014.		
600	Stolar, D.B., Willett, S.D. and Montgomery, D.R.: Characterization of topographic steady state in Taiwan. Earth and		
	Planetary Science Letters 261(3-4), 421-431, 2007.		
	Tucker, G.E., Bras, R.L. Hillslope processes, drainage density, and landscape morphology. Water Resources Reseach		
	<u>34(10), 2751-2764, 1998.</u>		
	Vance, D., Bickle M., Ivy-Ochs S., and Kubik P.W.: Erosion and exhumation in the Himalaya from cosmogenic isotope		Mis enforme: Police :+Titres(Times
605	inventories of river sediment, Earth Planet. Sci. Lett. 206(3-4), 273-288, 2003.		NewRoman)
	Vincent, S.J., Carter, A., Lavrishchev, V.A., Rice, S.P., Barabadze, T.G. and Hovius, N.: The exhumation of the western		Mis enforme: Police :+Titres(Times NewRoman)
	Greater Caucasus: a thermochronometric study. Geological Magazine 148(1), 1-21, 2011.		
	West, N., Kirby, E., Bierman, P., Slingerland, R., Ma, L., Rood, D. and Brantley, S. Regolith production and transport at the		Mis enforme: Police :+Titres(Times
	Susquehanna Shale Hills Critical Zone Observatory, part 2: insights from meteoric 10Be. Journal of Geophysical Research:		NewRoman)
610	Earth Surface 118(3), 1877-1896, 2013.		
	Whipp, D.M. and Elhers T.A.: Quantifying landslide frequency and sediment residence time in Nepal Himalaya. Sci. Adv. 5, eaav3482, 2019.		
	Whipple, K.X.: Fluvial landscape response time: How plausible is steady-state denudation?American Journal of Science		
I	301(4-5), 313-325, 2001.		
615	Whipple, K.X., Forte, A.M., DiBiase, R.A., Gasparini, N.M. and Ouimet, W.B.: 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, 2017.		
	Willett, S.D., Slingerland, R. and Hovius, N.: Uplift, shortening, and steady state topography in active mountain belts.		
	American journal of Science 301(4-5), 455-485, 2001.		
	20		

620 Willett, S.D., McCoy, S.W., Perron, J.T., Goren, L. and Chen, C.Y.: Dynamic reorganization of river basins. Science, Science 343, 1248765, 2014.

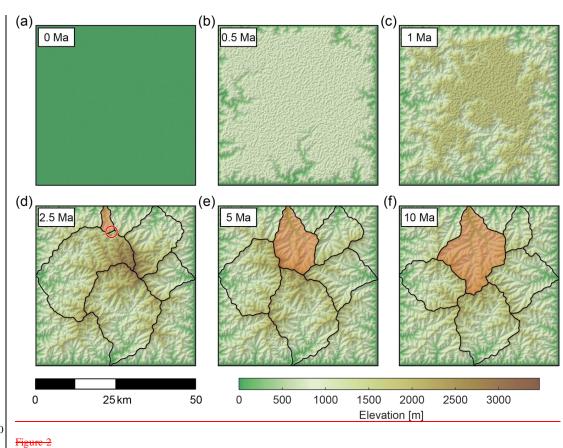
Wittmann, H., von Blanckenburg, F., Kruesmann, T., Norton, K.P. and Kubik, P.W.: Relation between rock uplift and denudation from cosmogenic nuclides in river sediment in the Central Alps of Switzerland. Journal of Geophysical Research: Earth Surface 112, F04010, 2007.

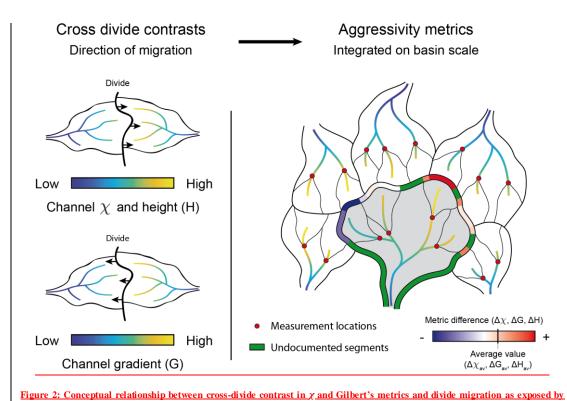
 Yanites, B.J., Tucker, G.E. and Anderson, R.S.: Numerical and analytical models of cosmogenic radionuclide dynamics in landslide-dominated drainage basins. Journal of Geophysical Research: Earth Surface 114, F01007, doi:10.1029/2008JF001088, 2009.



635 Figure 1: Basin-wide denudation rate variability as a function of drainage area in the Great Smoky Mountains. Data originallyOriginal dataset from Matmon et al., (2003a, b), denudation rates are reprocessed by Portenga and Bierman, (2011). Dashed black line show the estimated background uplift rate for the region of 0.03 mm.yr⁴.

Mis en forme: Anglais(Roy aume-Un) Mis en forme: Anglais(Roy aume-Un)





Willett et al. (2014) and Forte and Whipple, (2018). For x and height, divides migrate toward the drainages that present higher values. For channel gradient, divides migrate toward the drainages that present lower values. In our study, channel χ , local gradient and height are measured at the outlet (indicated by red circles) of basins for a reference area (basins bounded with thin black lines). Aggressivity metrics are then calculated for a given basin (represented in grey) by averaging along its perimeter the individual across divide differences in metrics between reference basins. The proportion of perimeter which is not shared by two reference basins is measured to give a confidence index of the calculated aggressivity.

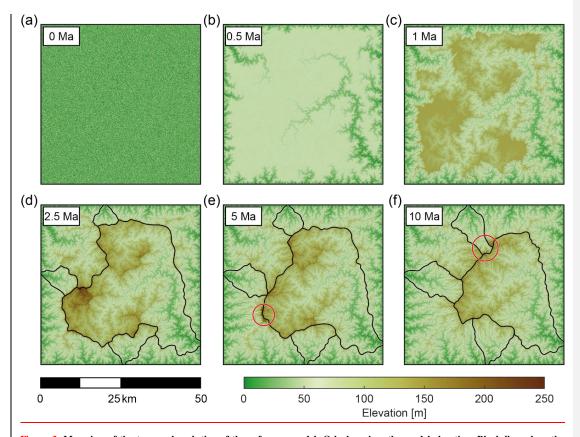
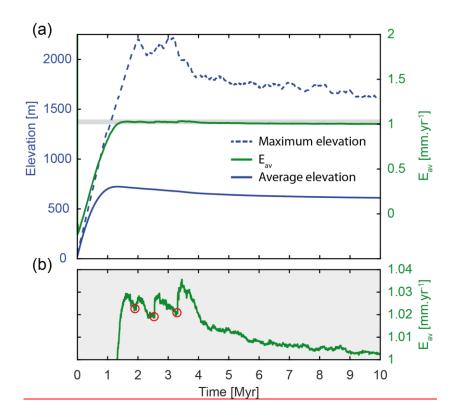


Figure 3: Map view of the temporal evolution of the reference model. Colorbar gives the model elevation. Black lines show the evolution - and the migration - over time of drainage divides for five drainage basins. One basin is colored in orange to underline its expansion. The red circleRed circles in figure 3d shows the location of an imminent3e and 3f show transient topography associated with drainage capture after 2.5 and 10 Myr of simulation, respectively (see Supplementary Video n°1).

Mis enforme: Anglais(Roy aume-Un) Mis enforme: Anglais(Roy aume-Un)



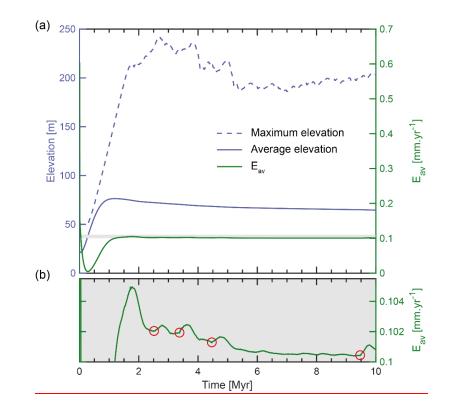
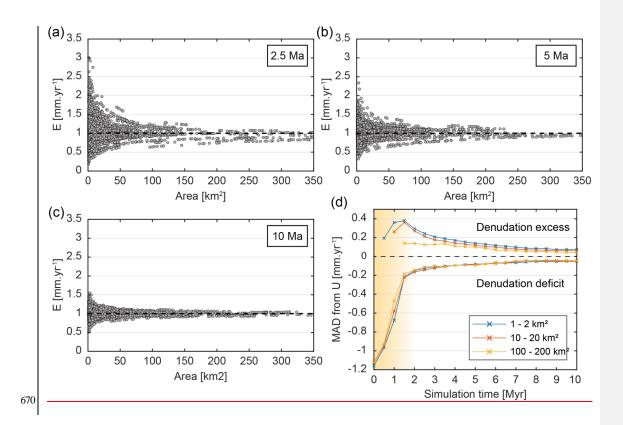


Figure 34; Evolution of the reference model over time. (a) Average elevation (blue solid line), maximum elevation (blue dashed line) and average denudation rate (green solid line) over the whole model. (b) Expanded view of the mean denudation rate of figure 3e4a (in the light grey area). Red circles highlight significant stream captures that lead to an abrupt increase in average denudation rates over a subsequent period of several time steps, (effects associated with stream captures at 4.5 Ma and 9.5 Ma are visible in Fig. 3e and 3f, respectively).

Mis en forme: Anglais(Royaume-Un) Mis en forme: Anglais(Royaume-Un) Mis en forme: Anglais(Royaume-Un) Mis en forme: Anglais(Royaume-Un)



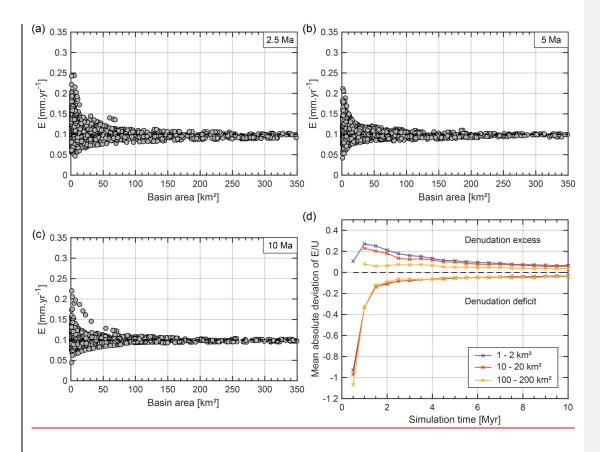
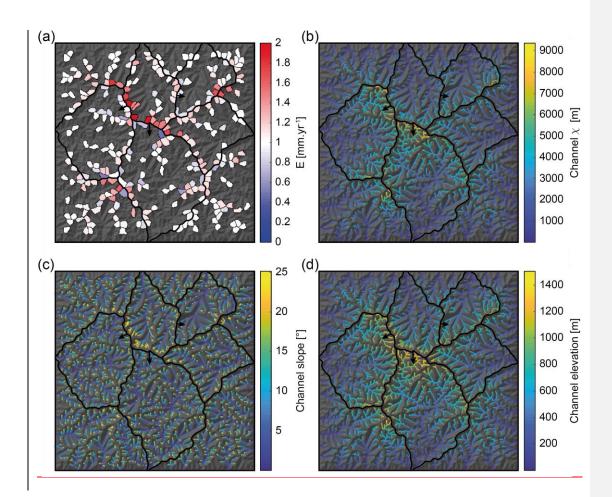
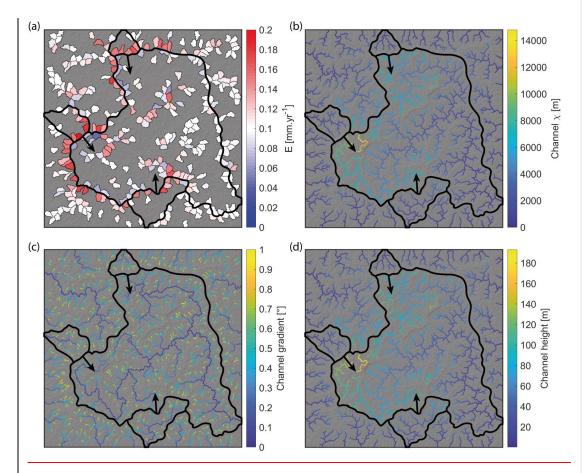


Figure 45: Variability of denudation rates over time for a compilation of five simulations of the reference model, with different initial noised DEM, (a) to (c) Variability of denudation rate as a function of basin area after 2.5, 5 and 10 Myr of simulation, respectively. (d) Mean absolute deviation (MAD) from uplift rate (0,1 mm.yr¹) for three <u>setsclasses</u> of basin sizes: 1-2 km², 10-200 km² and 100-200 km²/₂ between 0.5 and 10 Ma. Negative (positive) deviation is related to basins with a deficit (excess) of denudation. The orange color gradient corresponds to the transient period associated with the rising plateau before the model reaches an average topographic equilibrium.

680

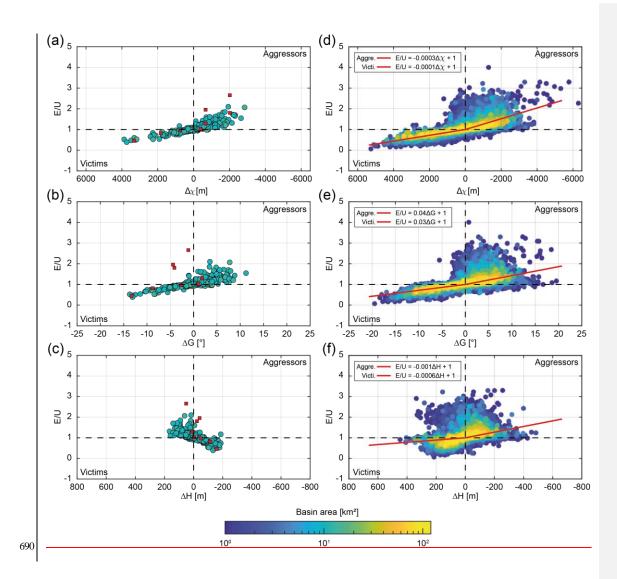
Mis enforme: Anglais(Roy aume-Un) Mis enforme: Anglais(Roy aume-Un)

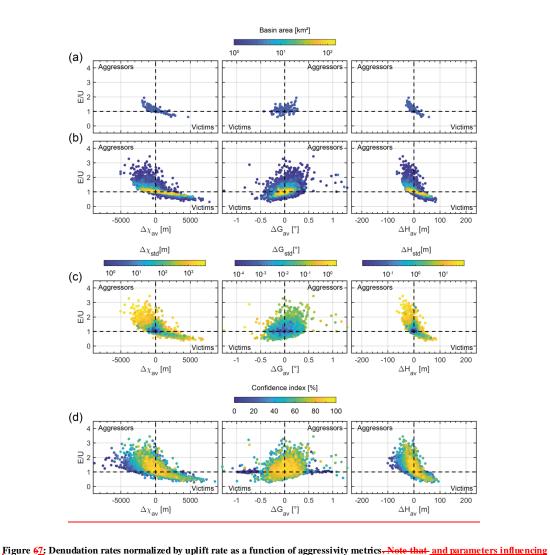




- 685
 - Figure 56: Denudation rates and cross-divide contrast metrics obtained for the reference model after 2.5 Myr- of simulation. Drainage network is extracted from a minimal drainage area of 1 km². (a) Map of denudation rates for basins of 2-4 km². Black thick lines correspond to basin divides in figure 2e2d. Black arrows, show the direction of divide migrations for one-basin.three selected basins. (b) X map. (c) Channel slopegradient map. (d) Channel elevationheight map.

Mis enforme: Anglais(ÉtatsUnis)

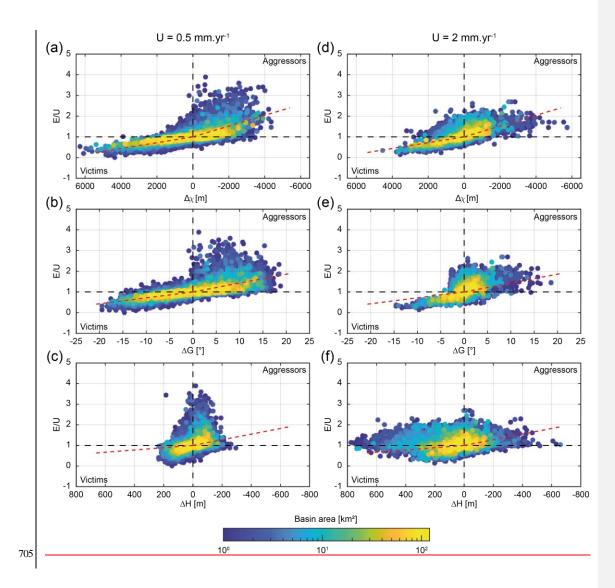




700

Figure 6.2 Definitiation rates nonnanzeous upmer rate as a function or aggressivity mentod state and parameters mindefining data dispersion. (a) Basins of 2-4 km² for the x-axis is reversed for both $A\chi$ and AH-reference model after 2.5 Myr. Red squares correspond to basins that contain at least one knickpoint. (d) to (f)-Basins of 10-20 km² for reference model after 2.5 Myr. Red squares correspond to basins that contain at least one knickpoint. (d) to (f)-Basins with a confidence index lower than 50 % are discarded from the analysis. (b) Basins of variable sizes, dessifiedsorted in seven area classes expanding geometrically with a multiplying factor of 2 from 1-2 to 64-128 km², every 0.5 Myr over the time period 2-10 Myr. Color scale indicates basin area. s^{-10} Myr. Red lines are linear fits for victim and aggressor basins Basins with a confidence index lower than 50 % are discarded from the analysis. (c) Basins of 1-2 km² over the time period 2-10 Myr, color scale indicating the standard deviation of $A\chi$, AG and AH, respectively. Basins with a confidence index lower than 50 % are the time period 2-10 Myr. Color scale indicating the standard deviation of $A\chi$, AG and AH, respectively. Basins with a confidence index lower than 50 % are the time period 2-10 Myr. Color scale indicating the standard deviation of $A\chi$, AG and AH, respectively. Basins with a confidence index lower than 50 % are the time period 2-10 Myr. Color scale indicates confidence index lower the time period 2-10 Myr. Color scale indicates confidence index lower the fide from the analysis. (d) Basin of 1-2 km² over the time period 2-10 Myr. Color scale indicates confidence index lower the fide from the analysis.

Mis en forme: Anglais(Roy aume-Un) Mis en forme: Anglais(Roy aume-Un)



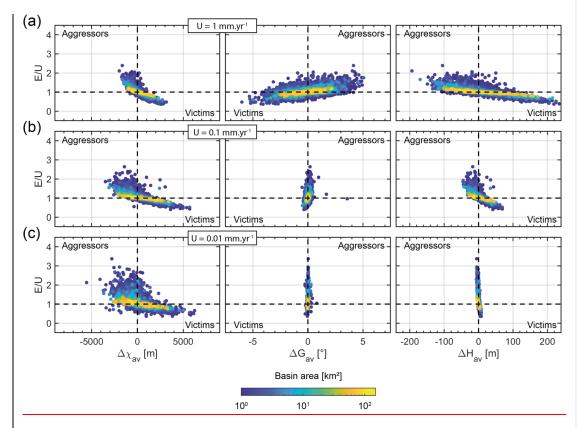


Figure 75: Sensitivity to uplift rates. Note that the horizontal axis is reversed for both $\Delta \chi$ and ΔH . Color scale indicates basin area. Basins of variable sizes, sorted in seven area classes expanding geometrically with a multiplying factor of 2 from 1-2 to 64-128 km², every 0.5 Myr over the time period 5-10 Myr. Basins with a confidence index lower than 50 % are discarded from the analysis. (a) Results Color scale indicates basin area. Red dashed lines correspond to the linear fits obtained for the reference model (see figure 6), (a) to (c) Same as Fig. 6d f, with an uplift rate of 0.51 mm.yr¹, (d) to (f) Same as Fig. 6d f, (b) Results with an uplift rate of 2 mm.yr⁴, 0.1 mm.yr¹ (reference model). (c) Results with an uplift rate of 0.01 mm.yr⁴.

Mis enforme: Anglais(Roy aume-Un) Mis enforme: Anglais(Roy aume-Un)

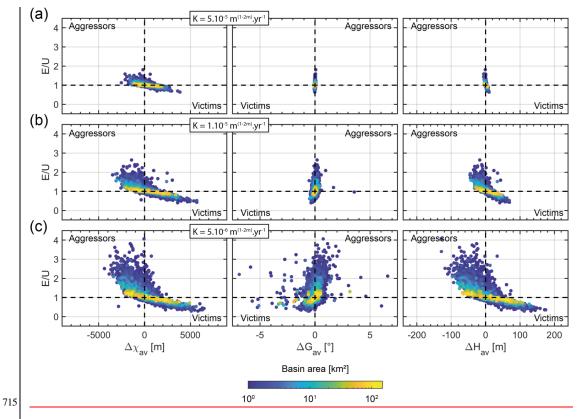


Figure 9: Effect of erodibility, Color scale indicates basin area.

Mis en forme: Anglais(Roy aume-Un

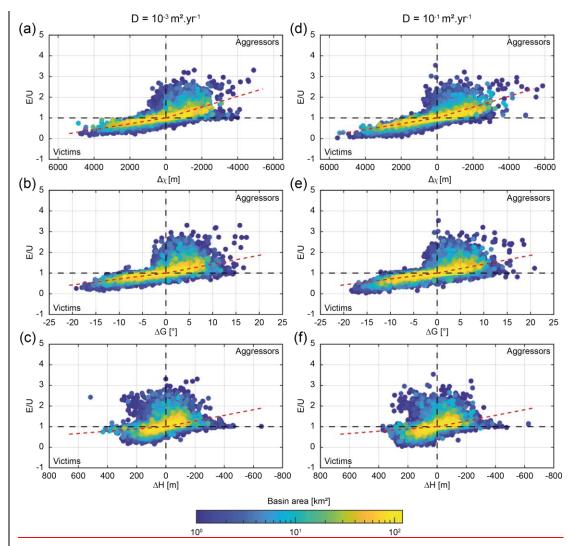
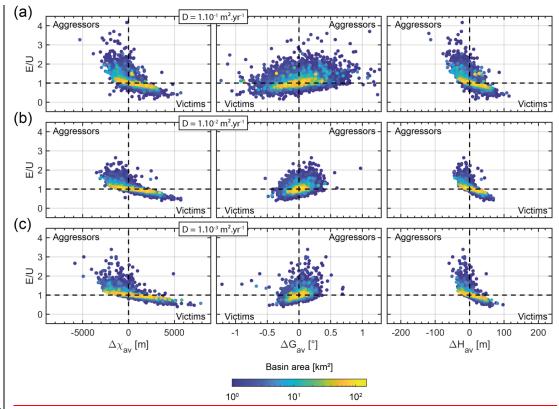


Figure 8:-Basins of variable sizes, sorted in seven area dasses expanding geometrically with a multiplying factor of 2 from 1-2 to 64-128 km², every 0.5 Myr over the time period 5-10 Myr. Basins with a confidence index lower than 50 % are discarded from the analysis. (a) Results with an erodibility coefficient of 5.10^{5} m^(1-2m).yr⁻¹. (b) Results with an erodibility coefficient of 5.10^{6} m^(1-2m).yr⁻¹.

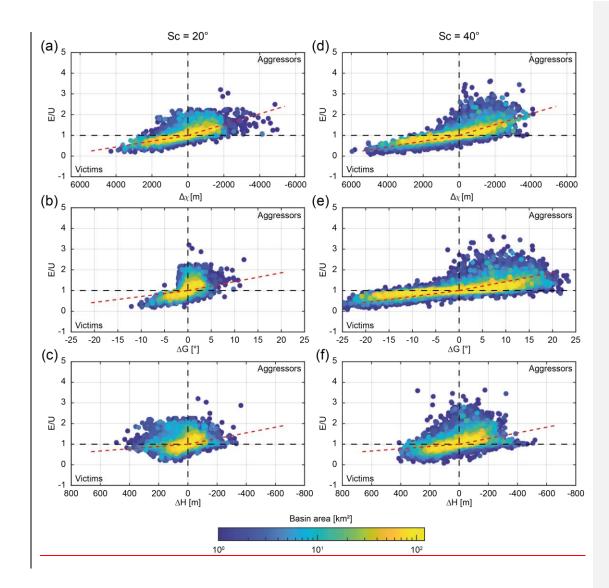


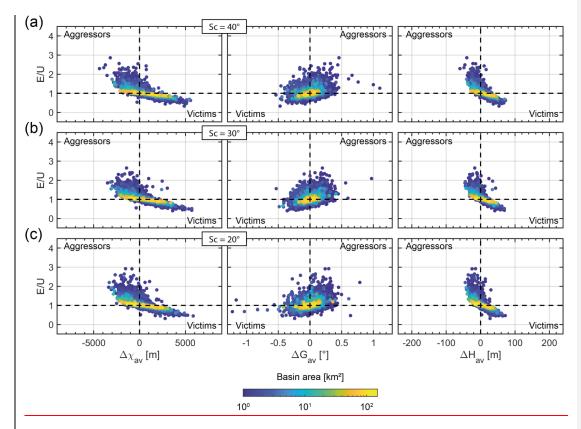
730

Figure 10: Effect of diffusivity. Note that the x axis is reversed for both 4χ and 4HColor scale indicates basin area. Basins of variable sizes, sorted in seven area dasses expanding geometrically with a multiplying factor of 2 from 1-2 to 64-128 km², every 0.5 Myr over the time period 5-10 Myr. Basins with a confidence index lower than 50 % are discarded from the analysis. (a) Results with a diffusivity of 10^{-1} m^2 , yr^{-1} . (b) Results with a diffusivity of 10^{-2} m^2 , yr^{-1} . (c) Results with a diffusivity of 10^{-3} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-3} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-3} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-3} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-3} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-3} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-3} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1} , (c) Results with a diffusivity of 10^{-4} m^2 , yr^{-1}

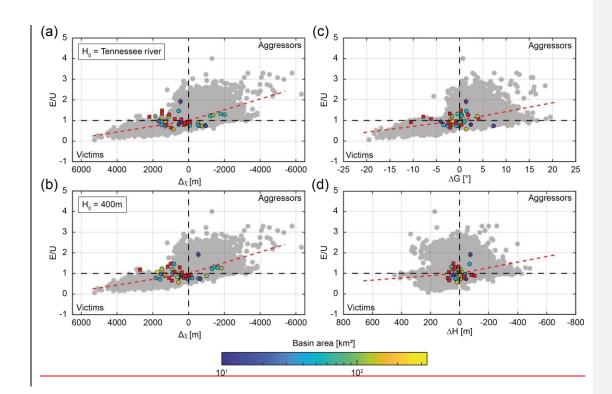
Mis en forme: Anglais(Roy aume-Unj

Mis en forme: Anglais(Royaume-Un Mis en forme: Anglais(Royaume-Un Mis en forme: Anglais(Royaume-Un Mis en forme: Anglais(Royaume-Un)





- Figure 911: Effect of critical slope. Note that the x axis is reversed for both 4x and 4H. Color scale indicates basin area. Red dashed lines correspond to Basins of variable sizes, sorted in seven area classes expanding geometrically with a multiplying factor of 2 from 1-2 to 64-128 km², every 0.5 Myr over the time period 5-10 Myr. Basins with a confidence index lower than 50 % are discarded from the linear fits obtained for the analysis. (a) Results with a critical slope of 40°. (b) Results with a critical slope of 30° (reference model (see figure 6), (a) to), (c) Same as Fig. 6d f. Results with a critical slope of 20°. (d) to (f) Same as Fig. 6d f.
 with a critical slope of 40°.
- Mis enforme: Anglais(Roy aume-Un) Mis enforme: Anglais(Roy aume-Un)



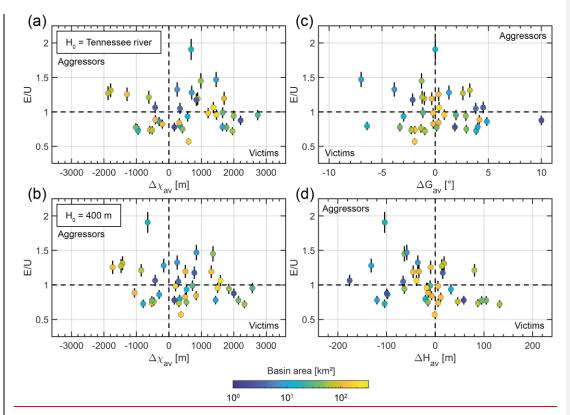
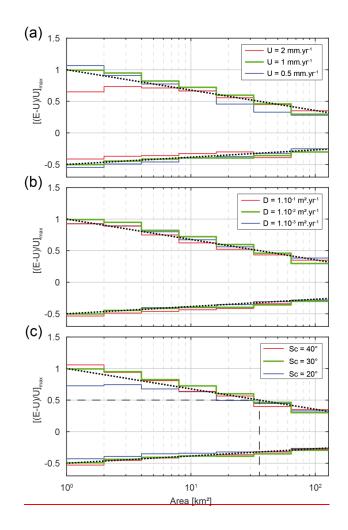
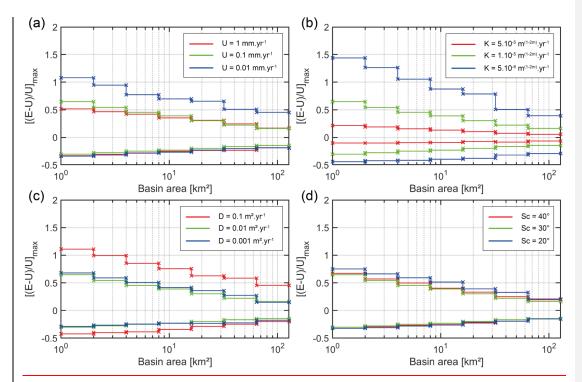


Figure 10: Denudation 12: Normalized denudation rates in the Great Smoky Mountains as a function of aggressivity metrics. Note that the x axis is reversed for both $\Delta \chi$ and ΔH . Color circles are reported according to the *E/U* ratio for basins sampled by Original dataset from Matmon et al., (2003b), with *E* recalculated by Portenga et al., 2011, error, (2011). Denudation rates and uncertainties are normalised by the estimated background uplift rate in this region of 0.03 mm.vr⁻¹. Error, bars are represented with thin black line-but are almost always smaller than symbols. Color scale of these circles is related to indicates basin size. Red squares are basins with an area less than 10 km². Red dashed lines correspond to linear fits obtained for the reference model. Grey circles in the background correspond to the reference model (see figure 6), (a) and (b) Relationship between denudation rates and $\Delta \neq \Delta \chi_{av_0}$ with a base level corresponding to the Tennessee river and at a fixed elevation of 400 m, respectively. (c) and (d) Relationship between denudation rates and GilbertGilbert's aggressivity metrics, AGA *G*_{ave} and AH, *A H*_{ave}, respectively.

Mis en forme: Anglais(Roy aume-Un) Mis en forme: Anglais(Roy aume-Un)







760 basin area for victim and aggressor watersheds (Eq.7).indicate the reference model. Seven sets of basin size are considered: 1-2, 2-4, 4-8, 8-16, 16-32, 32-64 and 64-128 km² every 0.5 Myr between 2,5 and 10 Myr. (a) Effect of uplift rate. (b) Effect of diffusivity erodibility. (c) Effect of diffusivity. (d) Effect of critical slope. Thin dashed black lines show the minimal basin area required for a maximal denudation rate deviation from uplift rate equal to 0.5.

Mis enforme: Anglais(Roy aume-Un) Mis enforme: Légende

