



1 **Effect of changing vegetation on denudation (part 2): Landscape**  
2 **response to transient climate and vegetation cover**

3 Manuel Schmid <sup>(1)</sup>, Todd A. Ehlers<sup>(1)\*</sup>, Christian Werner.<sup>(2)</sup>, Thomas Hickler<sup>(2)(3)</sup>, Juan-Pablo  
4 Fuentes-Espoz<sup>(4)</sup>

5 <sup>(1)</sup>University of Tuebingen, Department of Geosciences; Wilhelmstrasse 56,  
6 72074 Tuebingen, Germany (Manuel.Schmid@Uni-Tuebingen.de, Todd.Ehlers@Uni-Tuebingen.de)

7 <sup>(2)</sup>Senckenberg Biodiversity and Climate Research Center (BiK-F), Senckenberganlage 25,  
8 60325 Frankfurt/Main, Germany (Christian.Werner@Senckenberg.de)

9 <sup>(3)</sup>Department of Physical Geography, Geosciences, Goethe-University, Frankfurt,  
10 Altenhoferalle 1, 60438 Frankfurt/Main, Germany (Thomas.Hickler@Senckenberg.de)

11 <sup>(4)</sup>University of Chile, Department of Silviculture and Nature Conservation, Av. Santa Rosa 11315, La Pintana, Santiago  
12 RM, Chile (jfuente@uchile.cl)

13 \* Corresponding author: (Todd.Ehlers@Uni-Tuebingen.de)

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15 **Abstract** We present a numerical modelling investigation into the interactions between transient climate and vegetation  
16 cover with hillslope and fluvial processes. Model simulations were designed to investigate the effects of climate change  
17 and associated changes in surface vegetation cover on topographic basin metrics such as: slope, relief and channel  
18 steepness. The Landlab surface process model was used to evaluate the effects of temporal variations in vegetation cover  
19 on hillslope diffusion and detachment limited fluvial erosion. A suite of simulations were conducted to represent present-  
20 day climatic conditions and satellite-derived vegetation cover at the four EarthShape study areas as well hypothetical  
21 transient long term changes. Two different transient variations in climate and vegetation cover include a step change in  
22 climate or vegetation, as well as 100kyr oscillations over 5Myr. Results indicate that the coupled influence of surface  
23 vegetation cover and mean annual precipitation shifts basin landforms towards a new steady state, with the magnitude of  
24 change highly sensitive to the initial vegetation and climate conditions of the basin. Dry, non-vegetated basins show  
25 higher magnitudes of adjustment than basins that are situated in wetter conditions with higher vegetation cover. For  
26 coupled conditions when surface vegetation cover and mean annual precipitation change simultaneously, the landscape  
27 response tends to be weaker. When vegetation cover and mean annual precipitation change independently from each  
28 other, higher magnitude shifts in topographic metrics are simulated. Changes in vegetation cover show a higher impact  
29 on topography for low initial surface cover values whereas for areas with high initial surface cover, the effect of changes  
30 in precipitation dominate the formation of landscapes. This study demonstrates a sensitivity of catchment characteristics  
31 to different transient forcings in vegetation cover and mean annual precipitation, with a crucial role for initial vegetation  
32 and climate conditions. Ongoing research is developing fully-coupled landscape evolution and dynamic vegetation model  
33 (see companion paper) forced with predicted paleoclimate histories from an atmospheric general circulation model.

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## 35 1. Introduction

36 Plants cover most of Earth's surface and interact chemically and physically with the atmosphere, lithosphere and  
37 hydrosphere. The abundance and distribution of plants throughout Earth's history is a function, amongst other things, of  
38 changing climate conditions that can impact the temporal distribution of plant functional types and vegetation cover  
39 present in an area (Hughes, 2000; Muhs et al., 2001; Walther et al. 2002). The physical feedbacks of vegetation on the  
40 Earth's near surface manifest themselves mainly through an influence of plants on weathering, erosion, transport and the  
41 deposition of sediments (Marston, 2010; Amundson et al., 2015). Although the effects of biota on surface processes has  
42 been recognized for over a 100 years (e.g., Gilbert, 1877; Langbein and Schumm, 1958), early studies focused mainly on  
43 qualitative descriptions of the underlying processes. With the rise of new techniques to quantify mass transport from the  
44 plot- to catchment-scale, and the emergence of improved computing techniques and landscape evolution models, research  
45 shifted more towards building a quantitative understanding of how biota influence both hillslope and fluvial processes  
46 (Stephan and Gutknecht, 2002; Roering et al., 2002; Marston, 2010; Curran and Hession, 2013). The previous studies  
47 motivate the companion papers presented here. In part 1 (Werner et al. 2018-this volume) a dynamic vegetation model is  
48 used to evaluate the magnitude of past (Last Glacial Maximum to present) vegetation change along the climate and  
49 ecological gradient in the Coastal Cordillera of Chile. Part 2 (this study) presents a sensitivity analysis of how transient  
50 climate and vegetation impact catchment denudation. This component is evaluated through implementation of transient  
51 vegetation effects for hillslopes and rivers in a landscape evolution model. Together, these two components provide a  
52 conceptual basis for understanding how transient climate and vegetation impact catchment denudation.

53 Previous research in agricultural engineering has focused on plot-scale models to predict total soil loss in response to  
54 land-use change (Zhou et al. 2006; Feng et al. 2010) or general changes in plant surface cover (Gyssels et al., 2005), but  
55 do not draw conclusions about large-scale geomorphic feedbacks active over longer (millennial) timescales and larger  
56 spatial scales. However, a better understanding of how vegetation influences the large scale topographic features (e.g.  
57 relief, hillslope angles, catchment denudation) is crucial to understanding the evolution of modern landscapes. At the  
58 catchment scale, observational studies have found a correlation between higher values of mean vegetation cover and basin  
59 wide denudation rates or topographic metrics (Jeffery et al., 2014; Sangireddy et al., 2016; Acosta et al. 2015). Parallel  
60 to the previous observational studies, numerical modelling experiments of the interactions between landscape erosion and  
61 surface vegetation cover have also made progress. For example, Collins et al. (2004) were one of the first who attempted  
62 to couple vegetation dynamics with a landscape evolution model and found that the introduction of plants to their model  
63 resulted in steeper equilibrium landscapes with a higher variability in magnitude of erosional events. Following this,  
64 subsequent modelling studies built upon the previous findings with more sophisticated formulations of vegetation-erosion  
65 interactions (Istanbulluoglu and Bras, 2005) including the influence of root strength on hillslopes (Vergani et al., 2017).  
66 These studies found that not only is there a positive relationship between vegetation cover and mean catchment slope and  
67 elevation but there also exists an inverse relationship between vegetation cover and drainage density, due to the plants  
68 ability to hinder fluvial erosion and channel initiation.

69 The advances of the previous studies are limited mainly by their consideration of static vegetation cover or very simple  
70 formulations of dynamic vegetation cover. The exception to this is Istanbulluoglu and Bras (2005) who also considered  
71 the lag time for vegetation regrowth on hillslopes after a mass wasting event. However, numerous studies (Ledru et al.,  
72 1997; Allen and Breshears, 1998; Bachelet et al., 2003) document that vegetation cover changes in tandem with climate  
73 change over a range of timescales (decadal to million year). Missing from previous landscape evolution studies, is  
74 consideration of not only how transient vegetation cover influences catchment denudation, but also how coeval changes



75 in precipitation influence denudation. While the effects of climate change over geologic timescales on denudation rates  
76 and sediment transport dynamics have been investigated by others (e.g., Schaller et al., 2002; Dos seto et al., 2010;  
77 McPhillips et al., 2013), the combined effects of vegetation and climate change on catchment denudation have not. Thus,  
78 over longer (geologic) timescales, we are left with a complicated situation of both vegetation and climate changes, and  
79 the individual contributions of these changes to catchment scale denudation are difficult to disentangle.  
80 In this study, we compliment previous works by investigating both the temporal and spatial sensitivities of landscapes to  
81 the coupled vegetation-climate system. By focusing on simplified transient forcings such as a step change, or 100 kyr  
82 oscillations in climate and vegetation cover we present a sensitivity analysis of the landscape response to each of these  
83 changes, including a better understanding of the direction, magnitude and rates of landscape change. Our model setup is  
84 motivated by four study-areas along the climate and vegetation gradient in Chile (Fig. 1a) and illuminates the transient  
85 catchment response to biotic vs. climate changes. While we present results representative of four locations in the Coastal  
86 Cordillera, Chile, it is beyond the scope of this study to provide a detailed calibration to this area, and save that as a focus  
87 of future (ongoing) work as numerous new data sets emerge from the Coastal Cordillera as part of the German EarthShape  
88 priority program ([www.earthshape.net](http://www.earthshape.net)).

## 89 2. Background to model setup

90 Model setup and the range of initial conditions chosen for models were based upon four study-areas which are located in  
91 the Coastal Cordillera of Chile with a latitudinal range from 26°S to 38°S. The focus areas shown in Fig. 1a are part of  
92 the German EarthShape priority research program ([www.earthshape.net](http://www.earthshape.net)) and were chosen because of their similar granitic  
93 lithology and geologic and tectonic history (Andriessen and Reutter, 1994; McInnes et al., 1999; Juez-Larré et al., 2010;  
94 Makshev and Zentilli, 1999; Avdievitch et al., 2017), and the large gradient in climate and vegetation cover over the  
95 region (Fig. 1b,c). These study areas include (from north to south): Parque Nacional Pan de Azúcar; Reserva Santa Gracia;  
96 Parque Nacional La Campana, and Parque Nacional Nahuelbuta. Although this study does not explicitly present landscape  
97 evolution model results ‘calibrated’ to these specific areas, we loosely tuned the model input (e.g. precipitation, initial  
98 vegetation cover, rate of tectonic rock uplift) to these areas to provide simulation results that would have some relationship  
99 to changing vegetation and climate conditions observed on Earth.

100 Topographic metrics such as mean basin slope, total basin relief, mean basin channel steepness, and mean surface  
101 vegetation cover and mean annual precipitation were extracted for the main catchments and a subset of adjacent  
102 catchments (Fig. 1a, 1b; Fig. 2). Topographic metrics were extracted from 30m resolution digital elevation model from  
103 the NASA shuttle radar topography mission (SRTM), and vegetation related datasets from the moderate resolution  
104 imaging spectroradiometer (MODIS) satellite data ([https://landcover.usgs.gov/green\\_veg.php](https://landcover.usgs.gov/green_veg.php)).

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### 113 3. Methods

#### 114 3.1. Model Description and governing equations

115 For this study, we use the open-source model framework Landlab (Hobley et al., 2017), which provides easily accessible  
116 methods for building a landscape evolution model. Landlab provides the computational environment to build an  
117 experimental set-up to test hypotheses and conduct sensitivity analyses of topography to different surface process  
118 parameterizations.

119 For our study, we chose a model-domain representative of each of the four Chilean areas using an area of 100km<sup>2</sup> which  
120 is implemented as a rectangular grid, divided into 0.01km<sup>2</sup> spaced grid cells. For simplification in the presentation of  
121 results, we present our results for the driest, northern most area (Parque Nacional Pan de Azúcar) and for the Parque  
122 Nacional La Campana which is situated at 32°S Latitude (Fig.1) and shows the highest values in analyzed basin metrics  
123 (Fig. 2), although the general behavior and results presented here are representative of the other two areas not shown. The  
124 topographic evolution of the landscape is a result of tectonic uplift and surface processes, incorporating detachment  
125 limited fluvial erosion and diffusive transport of sediment across hillslopes (Fig.3). These processes are linked to, and  
126 vary in their effectiveness due to surface vegetation density. Details of the implementation of these processes into Landlab  
127 are explained in the following subsections.

128 This model setup is simplified in regards to hydrological parameters like, for example, soil moisture and groundwater and  
129 unsaturated zone flow. Also, the erosion and transport of material due to mass-wasting processes such as rockfalls and  
130 landslides are not considered. We argue that those processes do not play a major role in the basins we used for model-  
131 calibration and that the processes acting continuously along hillslopes and channels have the largest impact on shaping  
132 our reference landscapes. Additional caveats and limitations of the modeling approach used are discussed in Section 5.4.  
133 Main model parameters used in the model (and described below) are provided in Table 1.

#### 134 3.2. Boundary and Initial Conditions, and Model Free Parameters

135 In an effort to keep simulations for the different EarthShape areas comparable, we minimized the differences in parameters  
136 between simulations. The exceptions to this include the surface vegetation cover and mean annual precipitation, which  
137 were varied between simulations. One of the main controls on topography is the rock uplift rate. We kept the rock uplift  
138 rate temporally and spatially uniform across the domain and at 0.2 mm/yr (Table 1). Studies of the exhumation and rock  
139 uplift history of the Coastal Cordillera, Chile, are sparse at the latitudes investigated here, but existing and in progress  
140 studies further to the north are broadly consistent with the rock uplift rate used here (Juez-Larré et al., 2010; Avdievitch  
141 et al., 2017). Furthermore, the thermochronometer cooling ages in the northern Coastal Cordillera suggest constant  
142 Cenozoic exhumation over >50 Ma at this rate. Thus, despite being located on an active plate boundary, existing  
143 observations suggest relatively slow, and temporally constant rock uplift of this region.

144 The initial topography used in our simulations was a random white-noise topography with <1 m relief. To avoid unwanted  
145 transients related to the formation of this initial topography we conduct equilibrium simulations for each set of the  
146 different climate and vegetation scenarios (see below), and run the model for 15 Myr until a topographic steady-state is  
147 reached. The equilibrium topography after 15 Myr was used as the input topography for subsequent experiments that  
148 impose transient forcings in climate, vegetation, or both (Fig. 4). The model simulation time shown in subsequent plots  
149 is the time since completion of this initial 15 Myr steady-state topography development. In the results section, we present  
150 these results starting with differences in the initial steady-state topographies (prior to imposing transient forcings) and



151 then add different levels of complexity by imposing either: (1) a single transient step-change for the vegetation cover  
152 (Fig. 4b); (2) a step change in the mean annual precipitation (Fig. 4d); (3) 100 kyr oscillations in the vegetation cover  
153 (Fig. 4a); (4) 100 kyr oscillations in the mean annual precipitation (Fig. 4c); or (5) 100 kyr oscillation in both the  
154 vegetation cover and mean annual precipitation (both Fig. 4a and 4c). This approach was used to produce a stepwise  
155 increase in model complexity for evaluating the individual, and then combined, effects of fluvial and hillslope processes  
156 to different forcings.

157 The magnitude of induced rainfall transient forcings were based upon the present-day conditions along the Coastal  
158 Cordillera study areas (Fig. 1b, c). The step change and oscillations in vegetation cover and mean annual precipitation  
159 imposed on the experiments were designed to investigate vegetation and precipitation change effects on topography over  
160 the last ~0.9 Ma, the period during which a 100 kyr orbital forcing is dominant in Earth's climate. Given this timescale  
161 of interest, we impose a 10% magnitude change in the step-increase or decrease, or the amplitude change in oscillations  
162 for the vegetation cover. This magnitude of vegetation cover change is supported by dynamic vegetation modeling of  
163 vegetation changes over glacial-interglacial cycles in Chile (see companion paper by Werner et al. 2018-this journal) and  
164 to some degree elsewhere in the world (Allen et al., 2010; Prentice et al., 2011; Huntley et al., 2013). We assume that the  
165 present-day conditions of combined vegetation cover and mean annual precipitation along the north-south gradient of the  
166 coastal cordillera are directly linked (Fig. 1b, c), and therefore follow an empirical approach based on the present day  
167 mean annual precipitation vs. vegetation cover relationship in Chile (Fig. 5). We do this by associating each 10% change  
168 in vegetation cover ( $dV$ ) with a corresponding change of mean annual precipitation ( $dP$ , Fig. 5) present in the study areas  
169 considered.

170 The boundary conditions used in the model were the same for all simulations explained above (Fig. 3). One boundary  
171 was held at a fixed elevation and open to flow outside the domain. The other three were allowed to increase in elevation  
172 and had a zero-flux condition. This design for boundary conditions is similar to previous landscape evolution modeling  
173 studies (Istanbulluoglu and Bras, 2005) and provides a means for analyzing the effects of different vegetation cover and  
174 precipitation forcings on the individual catchment and subcatchment scale.

### 175 3.3. Vegetation Cover Dependent Geomorphic Transport Laws

176 The governing equation used for simulating topographic change in our experiments follows the continuity of mass.  
177 Changes in elevation at different points of the model domain over time  $dz(x,y,t)$  depend on

$$178 \frac{\delta z(x,y,t)}{\delta t} = U - k_d \nabla z - D_c \quad (1)$$

179 where  $z$  is elevation,  $x$ ,  $y$  are lateral distance,  $t$  is time,  $U$  is the rock uplift rate,  $k_d \nabla z$  the linear diffusive flux of sediment  
180 along hillslopes, and  $D_c$  the detachment capacity of channels (Tucker et al., 2001a). Our implementation of vegetation  
181 cover effects for the last two parameters in the Landlab model are described in more detail below.

#### 182 3.3.1 Vegetation Cover Influenced Diffusive Hillslope Transport

183 The change of topographic elevation in a landscape over time which is directly caused by hillslope-dependent diffusion  
184 can be characterized as:

$$185 \frac{\delta z}{\delta t} = -\nabla q_{sd} \quad (2)$$



186 Landscape evolution models characterize the flux of sediment  $q_s$  either as a linear or non-linear function of surface slope  
187  $S$  (Culling, 1960; Fernandez and Dietrich, 1997). In order to keep the number of free parameters for the simulation to a  
188 minimum, we used the linear description of hillslope diffusion:

$$189 \quad q_{sd} = K_d S \quad (3)$$

190 Following the approach of (Istanbulluoglu and Bras, 2005), we assign the linear diffusion coefficient  $K_d$  as a function of  
191 surface vegetation density  $V$ , an exponential coefficient  $\alpha$ , and a baseline diffusivity  $K_b$ , such that:

$$192 \quad K_d = K_b e^{-(\alpha V)} \quad (4)$$

### 193 3.3.2 Vegetation Cover Influence on Overland Flow and Fluvial Erosion

194 The erosion and transport of material due to river incision is represented as a detachment-limited process

$$195 \quad \frac{\delta z}{\delta t} = K_v A^m S^n - E_{threshold} \quad (5)$$

196 where the amount of lowering of elevation ( $z$ ) over time depends on the fluvial erodibility  $K_v$ , [ $L^{1-2m} t^{-1}$ ] (a function of  
197 soil type, climate, vegetation and the scaling of runoff to drainage area). The contributing drainage area to each node is  
198 represented by  $A$  [ $L^2$ ],  $S$  is the corresponding channelslope, and  $m$ , and  $n$  are empirically derived constants (Howard and  
199 Kerby, 1983; Whipple and Tucker, 1999). The detachment-limited erosion law can be re-written in a form that directly  
200 links excess bed shear stress to detachment capacity of a river where:

$$201 \quad D_c = k_e (\tau_b - \tau_c)^p \quad (6)$$

$$202 \quad \tau_b = \rho_w g R S \quad (7)$$

203 where  $k_e$  represents the erodibility of the substrate,  $\tau_b$  represents the bed shear stress acting on the bed surface,  $\tau_c$  is the  
204 critical shear stress needed to erode the substrate,  $\rho_w$  is the density of water,  $g$  is gravitational acceleration,  $R$  is hydraulic  
205 radius and  $S$  is local channel slope. By combining the shear-stress formulation with Mannings equation and introducing  
206 two Mannings surface roughness values ( $n_v$  for the influence of vegetation on surface roughness and  $n_s$  for the influence  
207 of the soil surface to surface roughness, e.g., Istanbulluoglu and Bras, 2005) we can reformulate the hydraulic radius  $R$   
208 (Wilgoose et al., 1991; Istanbulluoglu et al., 2004) and write the shear-stress formulation as a function of the Mannings  
209 coefficients:

$$210 \quad \tau_f = \rho_w g (n_s + n_v)^{\frac{6}{10}} q^m S^n \quad (8)$$

211 By writing  $n_v$  as a function of a basic Mannings coefficient for a reference vegetation cover  $n_{vr}$  and  $V_r$ , the vegetation  
212 cover at each individual node  $V$  and an empirical scaling parameter  $W$ , we arrive at:

$$213 \quad n_v = n_{vr} \left( \frac{V}{V_r} \right)^W \quad (9)$$

214 Combining equation 8 and 9 results in an equation for vegetation cover dependent shear stress at each node. The effects  
215 of surface vegetation cover on both diffusivity and fluvial erodibility are shown in Figure 6.

### 216 3.4 Model Evaluation

217 Model performance was evaluated using the above equations and different initial vegetation covers and mean annual  
218 precipitation based on the steady-state predicted topography. Our focus in this study is on the general surface process  
219 response to different transient vegetation and climate conditions, rather than a calibrated modeling study of the Chilean  
220 study areas. Nevertheless, topographic metrics of relief, mean slope, and normalized steepness index ( $K_{sn}$ ) were computed  
221 from the model results and compared to observed values from the 30 m SRTM DEM for each of the four areas (Fig. 2).



222 This was done to evaluate if our implementation of the governing equations in Section 3.4 produced topographies within  
223 reason of present day topographies in the four Chilean areas. A more detailed model calibration is beyond the scope of  
224 this study, and not meaningful without additional observational constraints on key parameters such as latitudinal variations  
225 in the rock uplift rate and erosivity.

#### 226 4. Results

227 Our presentation of results is structured around three groups of simulations. These include: 1. steady-state simulations  
228 where equilibrium topographies are calculated for different magnitudes of vegetation cover and identical precipitation  
229 forcing. A second set of steady-state simulations with the same magnitudes vegetation cover as 1. but with different  
230 precipitation forcings corresponding to each vegetation cover (Section 4.1). 2. Simulations with a transient step-change  
231 in either surface vegetation density or precipitation (Section 4.2) that is initiated after the landscape has reached steady  
232 state, and 3. simulations with a transient 100 kyr oscillating time series of changing vegetation or precipitation that occurs  
233 after the landscape has reached steady state (Section 4.3). For each group of transient simulations, we show the  
234 topographic evolution with help of standard topographic metrics and the corresponding erosion rates after the induced  
235 change.

##### 236 4.1 Equilibrium Topographic Metrics

237 Topographic metrics from each of the four Chilean focus areas (Fig. 1a) were extracted for comparison to equilibrium  
238 topographies predicted after 15 Myr of model simulation time. This comparison was done to document the model response  
239 to changing vegetation cover (with climate held constant) and changing vegetation cover and precipitation, and also to  
240 demonstrate the modeling approach employed throughout the rest of this study captures the general characteristics of  
241 different topographic metrics along the Chilean Coastal Cordillera. We refrain from conducting a more detailed model-  
242 observation comparison for reasons previously mentioned.

243 Analysis of the digital elevation model for each of our four Chilean focus areas illustrates observed changes in catchment  
244 relief, slope, and channel steepness ( $K_{sn}$ ) in relation to the surface vegetation (Fig. 7, red points) and latitude (Fig. 2). The  
245 general trend in the observed metrics shows a non-linear increase in each metric until a maximum is reached for regions  
246 with 70% vegetation cover. Following this, all observed metrics show a decline towards the area with 100% vegetation  
247 cover.

248 The model predicted equilibrium topographies (Fig. 7a,b,c) from four different steady-state simulations with variable  
249 vegetation cover but a constant mean annual precipitation (900 mm/yr) show a nearly linear increase in all observed basin  
250 metrics with increasing vegetation cover and therefore do not reflect the overall trend observed in the DEM from the  
251 study areas. For example, basin relief and slope are both under predicted for simulations with  $V < 100\%$  (Fig. 7a,b), and  
252 only the predicted maximum relief for a fully-vegetated simulation resembles the DEM maximum value. For the  
253 normalized channel steepness, only two observed mean values (for  $V = 10\%$  and  $70\%$ ) lie within the range of mean to  
254 maximum predicted  $K_{sn}$  values (Fig. 7c).

255 The resulting equilibrium topographies from simulations with both variable mean annual precipitation and vegetation  
256 cover (Fig. 7d,e,f) show an improved representation of the general trend of the DEM data. The vegetation cover and  
257 precipitation values used in these simulations come from the values observed in the Chilean areas (Fig. 1b, c; Fig. 5). In  
258 these simulations, the maximum in the observed basin metrics is situated at values of  $V = 30\%$  with a following slight



259 decrease in the metric for  $V = 30\%$  to  $V = 70\%$ , followed by a steeper decrease in metrics from  $V = 70\%$  to  $V = 100\%$ .  
260 Generally the model-based results tend to underestimate the basin relief and overestimate the basin channel steepness  
261 (Fig.7d,f). Variations in basin slope are captured for all but the non-vegetated state (Fig.7e).  
262 Although the above comparison between the models and observations demonstrates a range of misfits between the two,  
263 there are several key points worth noting. First, the model results shown are highly simplified in their setup (e.g. assuming  
264 similar rock uplift rate, identical lithology and constants), and assume the present day topography is in steady state for  
265 the comparison. Second, despite the previous simplifying assumptions, the degree of misfit between the observations and  
266 model are surprisingly small when both variable vegetation and variable precipitation, are considered (Fig. 7d,e,f). Finally  
267 (third), the general ‘humped’ shape curve observed in the Chilean areas is captured in the model predictions (Fig. 7d,e,f),  
268 with the notable exception that the maximum in observed values occurs at a higher vegetation cover ( $V = 70\%$ ) than the  
269 model predictions ( $V = 30\%$ ). Explanations for the possible source of these differences are revisited in the discussion  
270 section. Without additional observations from the Chilean areas, reduction of the misfit between the observations and  
271 models is not tractable.

## 272 4.2 Transient Topography – Step Change

273 The evolution of our model topographies after a induced instantaneous disturbance (Fig. 4) of either only the surface  
274 vegetation cover (Fig. 8, green lines) or a step change in only the mean annual precipitation (Fig. 8, blue lines) is analyzed  
275 for changes in topographic metrics for either a positive disturbance (Fig. 8a,b,c) or a negative disturbance (Fig.8d,e,f).  
276 This scenario with changes in only vegetation or only precipitation was chosen to analyze sensitivity of drainage basins  
277 to changes in vegetation cover and precipitation and isolate the effects of these specific transient forcings. Decoupling of  
278 vegetation and climate changes can occur via sudden disturbances, such as wildfires and lagged vegetation responses to  
279 climatic changes. Mean catchment erosion rates are also analyzed for their evolution after the disturbance (Fig.9). For  
280 simplicity in presentation, results are shown for only two of the initial vegetation ( $V$ ) and precipitation ( $P$ ) values for  
281 vegetation covers of 10 and 70%, and precipitation rates that correspond to these vegetation covers (i.e.  $P(V=10\%)$  or  
282  $P(V=70\%)$ ) in the Chilean areas (Fig. 5). The results described below show a general positive correlation between all  
283 observed topographic metrics and surface vegetation cover and a negative correlation between observed topographic  
284 metrics and mean annual precipitation and therefore supports the data from our equilibrium topography simulations. The  
285 adjustment time until the system again reaches a new steady state varies between different simulations.

### 286 4.2.1 Positive Step Change in Vegetation Cover or Precipitation

#### 287 Topographic Analysis

288 A positive step change in vegetation cover ( $V$ ) from  $V = 10\%$  to  $V = 20\%$  (solid green line Fig. 8a,b,c) leads to an increase  
289 of mean basin relief from 270m to 520m, mean basin slope from  $11.2^\circ$  to  $15.9^\circ$ , and mean basin channel steepness from  
290  $108\text{m}^{-0.9}$  to  $222\text{m}^{-0.9}$ , which corresponds to a factor 1.9, 1.42, and 2.1 change, respectively. The adjustment time until a  
291 new steady state in each metric is reached is 3.1Ma. The corresponding positive change in mean annual precipitation  
292 (solid blue lines, Fig. 8a,b,c) leads to a decrease of mean basin relief to 176m, mean basin slope to  $8.6^\circ$  and mean basin  
293 channel steepness to  $67\text{m}^{-0.9}$ . This corresponds to a decrease by factors of 1.5, 1.2 and 1.6, respectively. The adjustment  
294 time to new steady state conditions in this case are shorter and 1.1Ma (Fig.8a,b,c). A second feature of these results is the  
295 brief increase and then decrease in basin average slope angles following the step change (Fig. 8b).



296 For simulations with  $V = 70\%$  initial surface vegetation cover, a positive increase to  $V = 80\%$  leads to an increase of  
297 mean basin relief from 418m to 474m, mean basin slope from  $15.5^\circ$  to  $16.8^\circ$  and mean basin channel steepness from  
298  $172\text{m}^{0.9}$  to  $199\text{m}^{0.9}$ . This causes an increase in each metric by factors of 1.1, 1.1 and 1.2, respectively. The adjustment  
299 time to steady-state conditions is 1.9Ma (dotted green lines, Fig. 8a,b,c). The corresponding positive change in mean  
300 annual precipitation leads to a decrease of relief to 268m, decrease in slope to  $11.9^\circ$  and decrease of channel steepness to  
301  $105\text{m}^{0.9}$ . This resembles a decrease by factors 1.5, 1.3, 1.6, respectively. Adjustment time in this case is 1.7Ma (dotted  
302 blue lines, Fig. 8a,b,c). The basin slope data shows similar behavior as the  $V_{\text{ini}} = 10\%$  simulations with an initial decrease  
303 and then increase for a vegetation cover step change and an initial increase and then decrease for a step change in mean  
304 annual precipitation. Comparison of the change in the topographic metrics for the low ( $V=10\%$ ) and high ( $V=70\%$ ) initial  
305 vegetation covers, the magnitude of change in each metric is larger when the step change occurs on a low, rather than  
306 higher, initial vegetation cover topography.

### 307 **Erosion Rate Changes**

308 Erosion rates show instantaneous reactions to positive disturbances in vegetation or precipitation. Generally, the model  
309 results show a negative relationship between increases in vegetation cover and erosion rates and a positive relationship  
310 between increases in precipitation and erosion rates. Although the reaction between the disturbances and changes in  
311 erosion rates are instantaneous, the specific maximum or minimum is reached after some lag time and the magnitude and  
312 duration of non-equilibrium erosion rates varies between different simulation setups.

313 For initial vegetation cover of  $V = 10\%$ , a change in vegetation cover ( $dV$ ) of  $+10\%$  leads to a decrease in erosion rates  
314 from 0.2 to 0.03mm/yr (factor of 5.7 decrease, Fig. 9a green line). The minimum erosion rate is reached 43.5kyrs after  
315 the step change occurs. Following this minimum in erosion rates, the rates increase until the steady-state erosion rate is  
316 reached after the adjustment time. An increase in mean annual precipitation corresponding to a vegetation cover of 10%  
317 (i.e.  $P(V=10\%)$  to  $P(V=20\%)$ ; Fig. 5) leads to an increase in erosion rates to a maximum of 0.44mm/yr after 74.8kyrs  
318 (factor of 2.2 increase, Fig.9a, blue line). For initial vegetation of  $V = 70\%$  a vegetation increase of  $dV = +10\%$  results in  
319 minimum erosion rates of 0.14mm/yr after 117.7kyrs (factor of 1.4 decrease, Fig. 9b, green line). A corresponding  
320 increase in precipitation for these same vegetation conditions leads to maximum erosion rates of 0.44mm/yr after  
321 107.5kyrs which is an increase by a factor of 2.2 (Fig.9b, blue line). The previous results for a positive step change in  
322 vegetation or precipitation demonstrate that the magnitude of change in erosion rates is large for changes in precipitation  
323 rate than for vegetation cover changes, and in low initial vegetation cover settings ( $V=10\%$ ) that magnitude of change in  
324 erosion rates for changing vegetation is larger (compare green lines Fig. 9a with 9b).

325

### 326 **4.2.2 Negative Step Change in Vegetation Cover or Precipitation**

#### 327 **Topographic Analysis**

328 For negative step-changes in vegetation (green curves, Fig. 8d,e,f), the results show a sharp decrease in topographic  
329 metrics associated with shorter adjustment times compared to the positive step change experiments (compare Fig. 8d,e,f  
330 with a,b,c). For step changes in precipitation (blue curves, Fig. 8d,e,f), the increase of topographic metrics happens slower  
331 and therefore with longer adjustment times. A negative step change in vegetation cover from  $V = 10\%$  by  $dV = -10\%$   
332 leads to a decrease of mean basin relief from 269m to 35m, mean basin slope from  $11.2^\circ$  to  $2.3^\circ$  and mean basin channel  
333 steepness from  $108\text{m}^{0.9}$  to  $11\text{m}^{0.9}$  which resembles decreases by factors of 7.8, 3.8 and 9.6. The adjustment time until a  
334 new steady-state is reached is 0.26Ma (solid green lines, Fig. 8d,e,f). The corresponding negative change in precipitation



335 leads to an increase in mean basin relief to 512m, mean basin slope to 15.8° and mean basin channel steepness to 223m<sup>-0.9</sup>.  
336 These increases reflect changes by factors of 1.9, 1.4 and 2.1 with an adjustment time of 4.9Ma (dotted green lines,  
337 Fig.8d,e,f). Mean basin slope results (Fig. 8e) for a step change in vegetation illustrate a pulse-like feature of initially  
338 increasing slope values, followed by a decrease to lower slope values. In contrast, a negative step change in precipitation  
339 induce an initial decrease in slope, followed by a gradual increase in slope to a value higher than was initially observed  
340 before the change.

341 Simulations with initial vegetation cover  $V = 70\%$  and  $dV = -10\%$  show a decrease in mean basin relief from 418m to  
342 356m, mean basin slope from 15.5° to 13.6° and mean basin channel steepness from 172m<sup>-0.9</sup> to 144m<sup>-0.9</sup> which resembles  
343 changes by factors of 1.2, 1.1 and 1.2 and an adjustment time of 2.1Ma (dotted green lines, Fig.8d,e,f). Corresponding  
344 negative changes in precipitation lead to increase of basin relief of 465m, basin slope to 16.4° and channel steepness to  
345 195m<sup>-0.9</sup> which resembles changes by factors of 1.1 for all three values. Adjustment time in this case is 2.2Ma (dotted  
346 blue lines, Fig.8d,e,f). Behavior of mean basin slope after the step-change follows the  $V = 10\%$  simulations but shows  
347 lower amplitudes of basin slope for both step-changes in vegetation and precipitation.

#### 348 **Erosion Rates**

349 The positive step-change results (Fig. 9a, b) indicated that erosion rates reach their minimum or maximum after a lag time  
350 after the change, and show significant differences in the magnitude and duration of non-equilibrium conditions depending  
351 on if vegetation or precipitation were changing. Simulations with a decrease from  $V = 10\%$  to  $V = 0\%$  (Fig. 9c) show a  
352 sudden increase in erosion rates to a maximum value of 3.5mm/yr which is an increase from steady state conditions by a  
353 factor of 17.7 which is reached after 19.5kyrs (green line, Fig.9c). A step decrease in precipitation for this corresponding  
354 vegetation difference (i.e.  $P(V=10\%)$  to  $P(V=0\%)$ ) leads to a smaller, and protracted (longer adjustment time) decrease  
355 in erosion rates to 0.03mm/yr after 50.1kyrs. These conditions cause a factor of 5.6 decrease (blue line, Fig. 9c).  
356 Simulations of  $V = 70\%$  with a vegetation change of  $dV = -10\%$  show an increase in erosion rates to 0.27mm/yr which is  
357 a factor of 1.4 increase after 126.3kyrs (Fig. 9d). For the corresponding decrease in precipitation the data show a decrease  
358 in erosion rates to 0.15mm/yr after 124.5kyrs. This resembles change by factor of 1.2 (blue line, Fig. 9d).

#### 359 **4.3 Transient Topography – Oscillating**

360 In addition to simulations where a transient step-change in either surface vegetation density or mean annual precipitation  
361 was conducted, we set up two distinct sets of simulations with an oscillating transient signal with a period of 100kyrs.  
362 This period resembles the eccentricity driven part of the Milankovitch cycle that is dominant in Earth's climate over the  
363 last 0.9Ma.

#### 364 **4.3.1 Oscillating Surface Vegetation Cover, Constant Precipitation**

##### 365 **Topographic Analysis**

366 The topographic evolution in simulations with a constant precipitation (10 and 360mm/yr for  $V=10\%$ , and  $V=70\%$ ,  
367 respectively) and oscillating vegetation cover show a different response than the previous step change experiments. The  
368 differences depend on the initial steady-state vegetation cover prior to the onset of 100kyr oscillations. All observed basin  
369 metrics (Fig. 10) show an initial oscillating decrease in values until a new dynamic steady-state is reached where the  
370 amplitude in oscillations is less than in the preceding initial adjustment period. Simulations with  $V = 10\%$  (Fig. 10a) show  
371 a decline in the basin relief from 269m to 107m which resembles a decrease of the mean elevation of factor 2.5. The  
372 positive amplitude of oscillation is 9m, the negative amplitude is 8.3m but time intervals of negative amplitudes are longer



373 compared to positive amplitudes. For simulations with  $V = 70\%$  the reaction and adjustment to the new dynamic steady-  
374 state is less pronounced with a decline in relief from 410m to 407m (Factor 1.01) with positive and negative amplitudes  
375 in dynamic steady-state of 1.6m. Analysis of mean basin slope for the model topographies with low ( $V=10\%$ ) vegetation  
376 shows a similar behavior with a decrease of the mean slope from  $11.2^\circ$  (prior to the onset of oscillations) to  $6.0^\circ$  (Factor  
377 1.6, Fig. 10b). However, before this new equilibrium is reached, the slopes show an increase in mean slope for the first  
378 two periods of vegetation oscillation which then declines towards the new long-term stable dynamic equilibrium which  
379 is reached after approximately 500kyrs. Local maxima of mean basin slope coincide with local minima in basin relief.  
380 For the  $V = 70\%$  simulations, the reaction is significantly smaller with no change in mean slope for the new dynamic  
381 equilibrium and amplitudes of both positive and negative of  $0.16^\circ$ . Mean basin channel steepness (Fig. 10c) reflects the  
382 behavior of mean basin elevation. For  $V = 10\%$  simulations the mean channel steepness decreases from values of  $108\text{m}^{-0.9}$   
383  $^{0.9}$  to  $40\text{m}^{-0.9}$  (factor 2.7 change) with a positive amplitude of  $3.7\text{m}^{-0.9}$  and a negative amplitude of  $6.1\text{m}^{-0.9}$ . For  $V = 70\%$   
384 simulations the response is again only minor, compared to the lower initial vegetation cover simulations with a change of  
385 mean channel steepness from  $186\text{m}^{-0.9}$  to  $167\text{m}^{-0.9}$  and positive amplitudes of  $1.1\text{m}^{-0.9}$  and negative amplitudes of  $0.9\text{m}^{-0.9}$ .  
386  $^{0.9}$ . Like the elevation data, the steepness data shows a distinct oscillating pattern with slow increases to local maxima and  
387 rapid decreases to local minima which coincide with maxima/minima of elevation data. Taken together, the previous  
388 observations demonstrate a larger change in topography for oscillations in poorly vegetated areas compared to those with  
389 higher vegetation cover. Furthermore, the magnitude of topographic change that oscillations in vegetation impose on  
390 topography are largest in the first ~500 kyr after the onset of an oscillation, and diminish thereafter.

#### 391 **Erosion Rates**

392 The erosion history for simulations with oscillating vegetation cover (Fig. 11) demonstrate large variations in the erosion  
393 rate that depend on the average vegetation cover of the oscillation. Furthermore, pronounced differences in the amplitude  
394 of erosion occur if the vegetation cover is above or below the mean of the oscillation (Fig. 4a). More specifically,  
395 simulations with  $V = 10\%$  show a pattern of a small decrease in erosion rates (from 0.2 to 0.03mm/yr) when vegetation  
396 cover increases above the mean cover, in contrast to a large increase in erosion rates up to 3.3mm/yr when vegetation  
397 cover decreases below the mean of the oscillation (Fig. 2, Fig. 11). Maximum erosion rates decline over multiple periods  
398 of oscillation until they reach a dynamic steady-state with maximum rates of 1.2mm/yr at 760kyrs after the onset. Time  
399 periods of higher erosion rates ( $>0.2\text{mm/yr}$ ) have a mean duration of 28kyrs, whereas periods of lower erosion rates  
400 ( $<0.2\text{mm/yr}$ ) have a mean duration 72kyrs. For simulations with high vegetation cover ( $V = 70\%$ ) the maximum and  
401 minimum erosion rates are 0.28 and 0.15mm/yr, respectively. The magnitude of maximum and minimum erosion rate are  
402 not significantly time-dependent and are reached at each local vegetation cover minimum. The mean duration of periods  
403 with higher erosion rates ( $> 0.2\text{mm/yr}$ ) is 55kyrs whereas the duration for periods with lower rates ( $< 0.2\text{mm/yr}$ ) is 45kyrs.  
404 These results demonstrate that areas with low vegetation cover experience not only larger amplitudes of change in erosion  
405 rates, but also an asymmetric change whereby decreases in erosion rates are lower magnitude than the increases in erosion  
406 rates.

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411 **4.3.2 Oscillating precipitation, Constant Vegetation**

412 **Topographic Analysis**

413 The evolution of topographic parameters for simulations with oscillating mean annual precipitation and constant surface  
414 vegetation cover ( $V=10$  or  $70\%$ , Fig. 12) show a less extreme and smaller temporal change in erosion rate to variations  
415 in precipitation compared to the previous discussed effects of oscillating vegetation cover (Fig. 11). In Fig. 12a, the mean  
416 basin relief results for  $V = 10\%$  and oscillating precipitation show small variations (+4.9m to -3.8m) in relief around a  
417 mean of 269m, which is similar to the mean relief prior to the onset of oscillations at 5000kyrs. For simulations with  $V =$   
418  $70\%$  the change in relief is slightly more pronounced with adjustment to a new mean of 380m from 418m in steady-state  
419 conditions. This change in mean relief equates to factor of 1.1 change. The evolution of topographic slope (Fig. 12b) for  
420  $V = 10\%$  simulations shows an adjustment to a new dynamic equilibrium from  $11.2^\circ$  to  $10.6^\circ$  (Factor 1.05) with a negative  
421 amplitude of  $1.2^\circ$  and positive amplitude of  $0.9^\circ$ . For  $V = 70\%$  the mean slope values do not significantly change from  
422 steady-state to transient conditions and the amplitudes of oscillation are  $0.6^\circ$  (positive) and  $0.7^\circ$  (negative). Mean channel  
423 steepness (Fig. 12c) for  $V=10\%$  shows an adjustment from  $108\text{m}^{-0.9}$  to  $110\text{m}^{-0.9}$  (Factor 1.01). The amplitude of oscillation  
424 is  $4\text{m}^{-0.9}$  for both negative and positive amplitudes. For  $V_{ini} = 70\%$  simulations the topography adjusts from  $171\text{m}^{-0.9}$  to  
425  $152\text{m}^{-0.9}$  (Factor 1.1) with amplitudes of  $4.5\text{m}^{-0.9}$  for both positive and negative changes. Thus, although Figure 12  
426 illustrates changes in topographic metrics that result from oscillations in precipitation occurring around vegetation covers  
427 of 10 and 70%, these changes are significantly smaller than those predicted for constant precipitation, but oscillating  
428 vegetation conditions (Fig. 10).

429

430 **Erosion Rates**

431 Predicted erosion rates from simulations with constant surface vegetation cover and oscillating mean annual precipitation  
432 indicate different amplitudes of change around the mean erosion rate depending on the vegetation cover. For simulations  
433 with  $V = 10\%$  (Fig. 13, blue solid line) erosion rates oscillate symmetrically around the steady-state erosion rate of  
434  $0.2\text{mm/yr}$ . The maximum and minimum erosion rates of  $0.42$  and  $0.01\text{mm/yr}$ , respectively, result in no change in the  
435 mean rate. In contrast, predicted rates with a higher vegetation cover of  $V = 70\%$  (Fig. 13, blue dotted line) demonstrate  
436 an asymmetric oscillation in rates around the mean, whereby the maximum in rates ( $0.43\text{mm/yr}$ ) has a larger difference  
437 above the mean rate, then do the minimums in the oscillation ( $0.12\text{mm/yr}$ ). For this higher vegetation cover scenario, a  
438 gradual increase in the mean erosion rate from  $0.2$  to  $0.25\text{mm/yr}$  as time progresses is evident. Furthermore, the maximum  
439 and minimum erosion rates decline over several oscillation periods to values of  $0.38$  and  $0.15\text{mm/yr}$ , respectively. Taken  
440 together, these results indicate that oscillations in precipitation impact erosion with different magnitude depending on the  
441 amount of vegetation cover. Areas with low vegetation cover demonstrate the highest and symmetric oscillation of erosion  
442 rates due to changes in precipitation whereas in areas with high vegetation cover the effect of negative changes in  
443 precipitation is dampened by vegetation.

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## 447 5. Discussion

448 The previous results highlight predicted topographies with different sensitivities to changes in either surface vegetation  
449 cover or mean annual precipitation. The previous simulations, were conducted to isolate the magnitude of effect each  
450 parameter has on topography and erosion. In the following, we synthesize the previous results and then build upon them  
451 to discuss the, over longer time-scales, more common scenario of synchronous variation in both precipitation and  
452 vegetation cover.

### 453 5.1 Interpretation of Steady-State Simulations

454 Landscapes in a topographic steady-state show distinct features in topographic metrics that are widely used to estimate  
455 catchment-averaged erosion rates and therefore the leading processes of erosion within a landscape (DiBiase et al., 2010).  
456 In most studies, focusing on comparing *in-situ* measured  $^{10}\text{Be}$  erosion rates with topographic metrics, this is done in  
457 catchments with low variations in precipitation to focus on distinct topographic controls on soil erosion and transport  
458 processes. By conducting simulations with equal soil properties and assuming that the basic processes of sediment erosion  
459 and transport do not change between different climate-settings we can reproduce (Fig. 7) variations in topographic metrics  
460 over different climates seen in other studies (Langbein and Schumm, 1958; Walling and Webb, 1983) in steady-state  
461 landscapes with homogeneous erosion rates. Comparison of simulations with homogeneous precipitation and changing  
462 values of vegetation cover (Fig. 7a,b,c) to simulations with both changing precipitation and vegetation cover (Fig. 7d,e,f)  
463 we see that we are only able to reproduce a similar trend with a distinct peak in topographic metrics when both variable  
464 precipitation and vegetation cover are considered. From this, we conclude that modern model-based landscape evolution  
465 studies that aim to compare areas with different climates should incorporate vegetation dynamics in their simulations.  
466 Misfits between the predicted and Chilean observed topographic metrics (Fig. 7d,e,f) present when the vegetation and  
467 precipitation both vary likely stem from the simplicity of the model setup used and the likelihood of differences of the  
468 rock uplift rate and lithology's present in these areas.

### 469 5.2 Interpretation of Step-Change Experiments

470 Our analysis shows that changes in vegetation-cover typically have a higher magnitude of impact on topographies for  
471 lower values of initial vegetation cover, compared to simulations with high initial vegetation cover (Fig. 8, 9). In those  
472 settings the influence of vegetation cover outweighs the influence of precipitation in cases of negative and positive  
473 directions of the step change. The reasons for this is due to a higher impact of changes in vegetation on erosivity and  
474 diffusivity (parameter  $K_v$ ,  $K_d$ ; equation 4, 8) then changes in precipitation and therefore changes in runoff have on overall  
475 erosion values.

476 Furthermore, a negative step change in vegetation cover impacts the topographic metrics a factor of two more than do  
477 positive step change changes (Fig. 8d,e,f). This response is interpreted to be due to the non-linear reaction of diffusivity  
478 and fluvial erodibility to changes in vegetation cover (See Fig.6). Negative changes in vegetation cover lead to a higher  
479 overall change in diffusivity and erodibility which leads to a higher sensitivity of equations 4 and 8 to negative step-  
480 changes compared to positive step-changes.

481 Model results for the topographic metrics and erosion rates also indicate a difference in the adjustment times of the system  
482 until a new steady state is reached when either precipitation or vegetation cover changes (Figs. 8, 9). For simulations with  
483 positive step-changes (Fig. 8a,b,c) the adjustment time for changes in vegetation cover to reach a new equilibrium in



484 topographic metrics or erosion rates is three times higher than the adjustment time for changes in precipitation.  
485 Simulations with a negative step-changes in vegetation cover show an adjustment time which is lower by a factor of 18  
486 compared to negative changes in precipitation. This difference in adjustment time again is a result of the non-linear  
487 behavior of erosion parameters  $K_d$  and  $K_v$  which influence how effective a signal of increasing or decreasing erosion can  
488 travel through a river basin (Perron et al., 2012). High values of  $K_d$  and  $K_v$  are associated with lower adjustment times  
489 and are a result of negative changes in vegetation cover. The influence of changing precipitation on adjustment time  
490 behaves in a more linear fashion and therefore mostly depends on the overall magnitude of change. Therefore, positive  
491 step-changes in vegetation cover decrease  $K_d$  and  $K_v$  which leads to higher adjustment times than the corresponding  
492 changes in precipitation.

493 An increase and then decrease, or decrease and then increase, in predicted slope and erosion rates is observed for both the  
494 positive and negative step changes experiments (Fig. 8b,e; and Fig. 9). This non-linear response in both positive and  
495 negative step changes in precipitation and vegetation cover is also manifested in the subsequent oscillation experiments,  
496 but most clearly identifiable in the step change experiments. The explanation for this behavior is as follows. A positive  
497 step change in vegetation cover (Fig. 8b) leads to a decrease in fluvial capacity because increased vegetation cover  
498 increases the Manning's roughness (parameter  $n_v$ , equation 8). The effect of changing the Manning's roughness varies  
499 with the location in the catchment and influences which processes (fluvial or hillslope) most strongly influence slopes  
500 and erosion rates. In the upper part of catchments where contributing areas (and discharge) are low, this increase in  
501 Manning's roughness causes many areas to be below threshold conditions such that fluvial erosion is less efficient, and  
502 hillslope diffusion increases in importance's and lowers slopes. In the lower part of catchments, where contributing area  
503 and discharge are higher, changes in the Manning's roughness are not large enough to impact fluvial erosion because  
504 these areas remain at, or above, threshold conditions for erosion. With time, the lower regions of the catchments that are  
505 at or above threshold conditions propagate a wave of erosion up to the higher regions that are below threshold conditions.  
506 The propagating wave of erosion eventually leads to increase in slope angles, essential due to the response time of the  
507 fluvial network to adjust to new Manning's roughness conditions.

508 In contrast, a positive step change in mean annual precipitation leads to an initial increase in fluvial shear stress which  
509 initially causes headward incision of river channels and leads to wave of erosion that propagates upstream and increases  
510 channel slope values (Fig. 8b, see also e.g. Bonnet and Crave, 2003). The increase in channel slopes leads to an increase  
511 in the hillslope diffusive flux adjacent to the channels that then propagates upslope. Eventually, this increase in hillslope  
512 flux leads to a decrease in hillslope angles, and an overall reductions in mean catchment slopes after the systems reaches  
513 equilibrium.

514 Negative step-changes in vegetation cover or precipitation (Fig. 8e, green curves) shows the opposite behavior of the  
515 previous positive step change description. A negative step change in vegetation cover leads to an initial increase of fluvial  
516 erosion everywhere in the catchment because the Manning's roughness decreases everywhere. This catchment wide  
517 decrease in Manning's roughness leads to fluvial incision everywhere in the catchment and an increase in mean slope.  
518 However, eventually hillslope processes catch up with increased slopes near the channels and with time an overall  
519 reduction of slope occurs. Negative changes in precipitation (Fig. 8e, blue curves) lead to an initial decrease in fluvial  
520 erosion which leads to an increase in the significance of hillslope processes such that slope angles between channel and  
521 ridge decrease as hillslope processes fill in channels. With time, the fluvial network equilibrates to lower precipitation  
522 conditions by increasing slopes maintain equilibrium between erosion and rock uplift rates.



523 Thus, the contrasting behavior of either initially increasing or decreasing slopes and erosion rates, followed by a change  
524 in the opposite direction of this initial change highlight a complicated vegetation-climate induced response to changes in  
525 either parameter. This non-linear behavior, and the timescales over which these changes occur, suggest that modern-  
526 systems that experienced past changes in climate and vegetation will likely be in a state of transience and the concept of  
527 a dynamic equilibrium in hillslope angles and erosion rates may be difficult to achieve in these natural systems.  
528 Previous studies have inferred relationships between mean catchment erosion rates derived from cosmogenic  
529 radionuclides and topographic metrics (e.g., DiBiase et al., 2010; DiBiase and Whipple, 2011). However, the previous  
530 discussion of how topographic metrics change in response to variable precipitation and vegetation suggest that empirical  
531 relationships between erosion rates and topographic metrics contain a signal of climate and vegetation cover in the  
532 catchment. We illustrate the effect of step changes in climate and vegetation on the new steady-state of topographic  
533 metrics in Figure 14. In this example, the new steady state conditions in basin relief and mean slope after a modest (+/-  
534 10%) change in vegetation or precipitation (triangles) differ from the initial steady-state condition (circles). These  
535 changes in topographic metrics when the new steady-state is achieved occur despite the rock uplift rate remaining  
536 constant. Thus, differences in mean slope and relief can occur solely due to changes in climate or vegetation and are not  
537 necessarily linked to variations in erosion rate. The change in relief or slope is most pronounced for catchments with  
538 initially low (e.g. 10%) vegetation cover.

### 539 **5.3 Interpretation of Oscillation Experiments**

540 The results from the 100kyr oscillating vegetation and precipitation conditions shows that oscillating vegetation cover  
541 without the corresponding oscillations in precipitation leads to adjustments of topographic features, to a new dynamic  
542 equilibrium after approximately 1.5Ma (Figs. 10, 11). The results indicate that the magnitude of adjustment depends on  
543 the initial vegetation cover, whereby simulations with 10% initial vegetation cover (solid lines, Fig. 10) show the largest  
544 changes from the initial (pre-oscillation) conditions to the new dynamic steady-state. Simulations with 70% initial  
545 vegetation cover (dashed lines, Fig. 10) show only minor adjustment to a new dynamic steady-state and lower amplitudes  
546 of oscillation. This is also represented by the mean basin erosion rates which show a significant peak for the first period  
547 of oscillating vegetation cover with erosion rates being 16 times higher than steady-state erosion rates for simulations  
548 with 10% initial vegetation cover whereas the peak erosion rate for 70% vegetation cover simulations is only higher by a  
549 factor of 1.4 in the first period of oscillation. The previously described response of topographic metrics and erosion rates  
550 to oscillating vegetation are due to processes described in the previous step change experiments. For example, the  
551 asymmetric oscillations in topographic metrics for  $V=10\%$  (Fig. 10) are due to the superposition of positive, then negative  
552 changes described in section 5.2. Variations in the imposed Manning's roughness, and relative strengths of fluvial vs  
553 hillslope processes in different parts of the catchments at different times causes the topographic metrics and erosion rates  
554 to have a variable amplitude and shape of response from the symmetric oscillations imposed on the topography (Fig. 4a).  
555 Simulations with oscillating precipitation and constant vegetation cover however show a less pronounced shift to new  
556 equilibrium conditions and in general lower amplitudes of oscillation in both topographic metrics and erosion rate (Figs.  
557 12, 13). This difference in the response of the topographic metrics and erosion rates in Figures 12 and 13, compared to  
558 the oscillating vegetation cover experiments (Figs 10, 11) is due to a generally higher impact of changes in vegetation  
559 cover on parameters which guide erosion rates and therefore adjustment to topographic metrics, compared to the  
560 calibrated, corresponding changes in precipitation in our model domains. Especially for simulations with low initial  
561 vegetation cover the effect of changing vegetation shows larger magnitude effects because of the non-linear response of



562 diffusivity and fluvial erodibility to changes in vegetation cover compared to the linear response to changes in  
563 precipitation.

#### 564 **5.4 Coupled Oscillations in Both Vegetation and Precipitation**

565 The previous sections present a sensitivity analysis of how step changes or oscillations in either vegetation cover or  
566 precipitation influence topography. Here we present a step-wise increase towards reality by investigating the topographic  
567 response to changes in both precipitation and climate at the same time. The amplitude of change prescribed for both  
568 precipitation and vegetation is based upon the present empirical relationship observed in the Chilean study areas for initial  
569 vegetation covers of 10 and 70%, and mean annual precipitations for 10 and 360mm/yr (Fig. 5). As with the previous  
570 experiments, oscillations in parameters were imposed upon steady-state topography that developed with the previous  
571 values, and a rock uplift rate of 0.2mm/yr.

572 Figure 15 shows the evolution of topographic metrics for simulations with combined oscillations in precipitation and  
573 vegetation. The variation in topographic metrics resembles those described for simulations with constant vegetation cover  
574 and oscillating climate by showing little to no significant adjustment towards new dynamic steady-state conditions. The  
575 amplitudes of oscillation are dampened from those of previous results because of the opposing effects of changes in  
576 precipitation and vegetation cover (e.g. compare blue and green curves in Figs. 8 and 9).

577 However, inspection of the predicted erosion rates (Fig. 16) for the combined oscillations indicates a significant ( $\sim 0.1$ ; -  
578  $\sim 0.15$ mm/yr), and highly non-linear response. The response between the 70% and 10% vegetation cover scenarios are  
579 very different such that for heavily vegetated areas ( $P(V=70\%)$ ) erosion rates typically increase during an oscillation,  
580 whereas for the low vegetation cover conditions ( $P(V=10\%)$ ) erosion rates initially show a decrease, and then an increase  
581 and decrease at a higher frequency.

582 To better understand this contrast in the response to combined precipitation and vegetation changes, the first cycle of the  
583 imposed oscillation is shown in Figure 17. After an oscillation starts, the 10% initial vegetation cover simulations show  
584 a decline in erosion rates with the minimum erosion rate correlated with highest values of both vegetation cover and mean  
585 annual precipitation (compare top and bottom panels). This part of the response is interpreted as resulting from the  
586 hindering effect of increased vegetation on erosion rates outweighing the impact of higher values of precipitation on  
587 erosion rates (Fig. 17). After values of vegetation cover and precipitation start to decline, erosion rates show a very rapid  
588 increase to values of  $\sim 0.3$ mm/yr. This increase in erosion rates is due to an increase in both  $K_v$  and  $K_d$  (Fig. 3b, equations  
589 4, 5) which outcompetes the effect of precipitation decrease.

590 Following this, a sudden drop in erosion rates to 0mm/yr occurs and lasts for 3kyrs due to the onset of hyper arid  
591 conditions at minimum precipitation. After this low in erosion rates, they increase again to 0.3mm/yr as precipitation and  
592 vegetation cover increase while the effect of increased precipitation outweighs the effect of the non-linear decrease in  $K_v$   
593 and  $K_d$  (Fig. 3b, c; equations 4, 5). Finally, at the end of this complex cycle a decrease in erosion rates occurs (Fig. 17b)  
594 while vegetation and precipitation are increasing (upper panel) because the effect of vegetation increases  $K_v/K_d$  and  
595 outweighs the effect of increasing precipitation.

596 Lastly, a clearly different behavior in erosion rates occurs for settings with higher vegetation cover (e.g.  $P(V=70\%)$ )  
597 compared to the previous lower vegetation cover scenarios. As the vegetation cover and precipitation increase (Fig. 17a)  
598 in the first half of the 100kyr cycle, the erosion rates increase to values of approximately 0.35mm/yr. This is due the  
599 increase in precipitation which outcompetes the decline in erosivity/diffusivity parameters  $K_d$  and  $K_v$ . Following this,  
600 when vegetation cover and precipitation decrease in the second half of the cycle, little to no change occurs in the erosion



601 rates. This near static behavior in erosion rates while precipitation and vegetation cover decrease is due to an equilibrium  
602 between the negative effect on erosion rates for decreasing precipitation and the positive effect on erosion rates for  
603 decreasing vegetation cover.

604 In summary, the non-linear shape of the vegetation dependent erosivity ( $K_v$ ) and hillslope diffusivity ( $K_d$ ) in combination  
605 with linear effects of mean annual precipitation on erosion rates, exert a primary control on the direction and magnitude  
606 of change in catchment average erosion rates. Despite a simple oscillating behavior in precipitation and vegetation cover,  
607 a complex and non-linear response in erosion rates occurs. The implications of this are large for observational studies of  
608 catchment average erosion rates and suggest that the direction and magnitude of response observed in a setting is highly  
609 dependent on the mean vegetation and precipitation conditions of the catchment, as well as what time the observations  
610 are made within the cycle of the varying vegetation and precipitation. Furthermore, these results highlight the need for  
611 future modeling studies (and motivation for our ongoing work), to investigate the response of catchment topography and  
612 erosion rates to more realistic climate and vegetation change scenarios, as well as a broader range of initial vegetation  
613 covers and precipitation rates than those explored here such that the threshold in behavior between the two curves shown  
614 in figure 17b can be understood.

615 This could be achieved by using simulation results from state-of-art dynamic vegetation models (e.g. Smith et al., 2014;  
616 see also companion paper by Werner et al., 2018) as inputs into the landscape evolution model or by full coupling between  
617 both model types. Vegetation simulations could, e.g. benefit from simulated changes in soil depth, which can crucially  
618 determine plant water stress, provided by the landscape evolution model. Coupling between Landlab and the dynamic  
619 regional to global vegetation model LPJ-GUESS (Smith et al., 2014; Werner et al., 2018) is envisioned.

## 620 **5.5 Model Restrictions and Caveats**

621 Like other previous work on this topic (Collins et al., 2004; Istanbuloglu and Bras 2005) the model setup used in this  
622 study was intentionally simplified to document how different vegetation and climate related factors impact topography  
623 over long (geologically relevant) timescales. We acknowledge that future model-studies should address some of the  
624 restrictions imposed by our approach to evaluate their significance for the results presented here. Future work should  
625 consider a transport-limited fluvial model or a fully-coupled alluvial sedimentation and transport model. Addition of this  
626 could bring new understanding in to how vegetation not only influences detachment limited systems, but also influences  
627 sedimentation and entrainment of material. This added level of complexity could however limit (due to computational  
628 concerns) the temporal scales over the investigation can be conducted. Future studies could improve upon this work by  
629 considering a more in-depth parameterization of how vegetation related processes (e.g. root depth and density, plant  
630 functional type) influence topographic metrics and erosion rates. Due to the long-timescales considered here, mean annual  
631 precipitation rather than a stochastic distribution of precipitation was implemented. Future work should evaluate how  
632 stochastic distributions in precipitation and extreme events in arid, poorly vegetation settings, impact these results.

633 Regarding the vegetation and water-budget, a more sophisticated model of evapotranspiration and infiltration as a function  
634 to surface plant cover and plant functional traits such as rooting depth would improve model predictions and is a priority  
635 goal of future research within this project. Improvements will come from planned coupling of surface process model with  
636 the dynamic vegetation model LPJ-GUESS (Werner et al., 2018, this issue).

637 We also acknowledge here that the transient forcings we have chosen for driving our model are simplistic and could be  
638 improved by a higher-fidelity time-series of climate over the last millennia. We choose a 100kyr, eccentricity driven,  
639 periodicity because it is widely recognized that the eccentricity cycles are a main control in driving Earths glacial cycle



640 over the last 0.9Ma. While this approach is reasonable for a sensitivity analysis such as we've conducted, it prohibits a  
641 detailed comparison to observations without additional refinement. Our results suggest that a shorter periodicity, which  
642 would resemble other periodicities in the Milankovitch cycle (e.g., 41kyrs, 23kyrs) or shorter time scale climate variations,  
643 such as Heinrich events (see Huntley et al., 2013) would lead to smaller magnitudes of adjustment to new dynamic  
644 equilibria, because of short timespans in high-/low-erosive climate conditions within one period.

## 645 6. Conclusions

646 The results from our model-based experiments in comparison to observations from topographic analysis from four  
647 different areas in the Coastal Cordillera in Chile show that the interactions of vegetation cover and mean annual  
648 precipitation on the evolution of landscapes is a complex system with competing effects. Main conclusion which emerge  
649 from this study are:

650 (I) vegetation cover in general has a hindering effect on eroding surface processes but the magnitude on which changes  
651 in vegetation cover affect these processes is a function of the initial state of the system. Changes in systems with higher  
652 initial values of vegetation cover have a less pronounced effect than changes in systems with lower initial vegetation  
653 cover.

654 (II) In comparison to the Coastal Cordilleras of Chile, the relationship between precipitation and surface vegetation cover  
655 shows a distinct shape: For a 10% increase in surface vegetation cover, the corresponding increase in mean annual  
656 precipitation is smaller in areas of lower vegetation cover and increases for areas with higher vegetation cover. This has  
657 an effect on transient topographies by shifting the equilibrium of vegetation and precipitation effects on erosion rates.

658 (III) Following our step-change simulations, our model results show different behaviors for changes in vegetation-cover  
659 and mean annual precipitation. While increases in mean annual precipitation have an increasing effect on erosion rates  
660 and therefore a long-term negative effect on topographic metrics, an increase in vegetation cover hinders erosion, and  
661 leads to higher topographic metrics. The magnitude of these changes is highly dependent on the initial vegetation cover  
662 and precipitation before the step-change.

663 (IV) Simulations with either oscillating vegetation cover or oscillating precipitation show adjustments to new dynamic  
664 mean values around which the basin metrics oscillate. The magnitude of adjustment is highly sensitive to initial vegetation  
665 cover, where simulations with 10% initial cover show higher magnitudes than simulations with 70% cover, for oscillating  
666 vegetation. Oscillating precipitation leads to lower-/no adjustments but an oscillation of basin metrics around the initial  
667 mean values with generally lower amplitudes compared to simulations with oscillating vegetation cover.

668 (V) Simulations with coupled oscillations of both vegetation cover and precipitation show only small magnitudes of  
669 adjustments of topography metrics to new dynamic equilibria similar to simulations with a oscillation in only  
670 precipitation. However corresponding erosion rates show a complex pattern of rapid increases and decreases which results  
671 from an interplay of competing effects of hindering of erosion by vegetation and aiding of erosion by precipitation.

672 Taken together, the above findings from this study highlight a non-linear and highly variable behavior in how variations  
673 in vegetation cover impact erosion and topographic properties. The complexity in how vegetation cover and precipitation  
674 changes influence topography highlights the need for future work to consider both of these factors in tandem, rather  
675 singling out either parameter (vegetation cover or precipitation) to understand potential transients in topography.

676

677



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684



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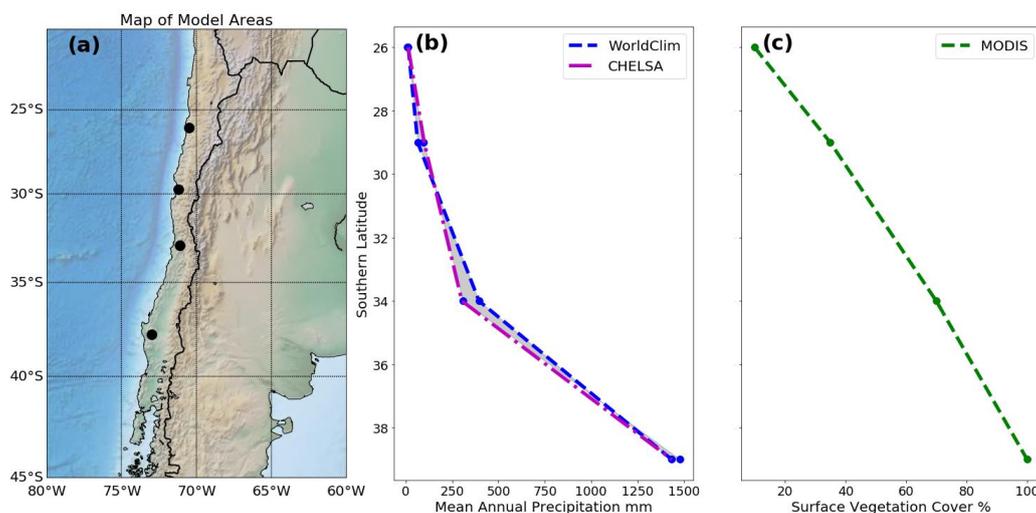
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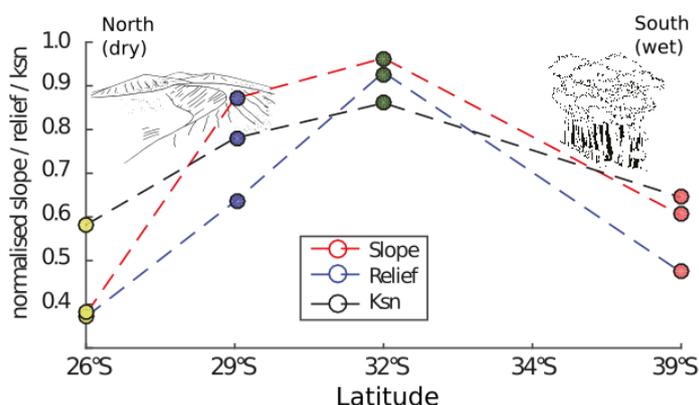


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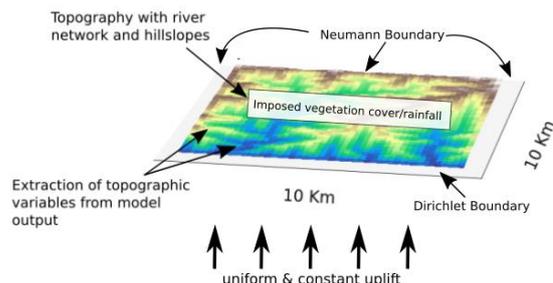
868 **Figure 1** Overview of the geographic location, precipitation, and vegetation cover of the Coastal Cordillera, Chile studies areas  
 869 used for model setup. A) Digital topography of the areas considered and corresponding to the EarthShape  
 870 ([www.earthshape.net](http://www.earthshape.net)) focus areas where ongoing related research is located. B) Observed present day mean annual  
 871 precipitation from the WorldClim and CHELSA datasets used as model input. B) Present day maximum surface vegetation  
 872 cover from MODIS data.



873

874 **Figure 2** Normalized basin metrics for study-areas derived from 90 m SRTM digital topography from the study areas shown  
 875 in Figure 1a Dots represent cumulative mean values calculated for all locations using 5-8 representative catchments in each  
 876 area. Dotted lines represent linear interpolation for normalized slope, relief and channel steepness ( $k_{sn}$ ) values. Note the gradual  
 877 increase, then decrease in all metrics around study area at  $\sim 32^\circ\text{S}$ .

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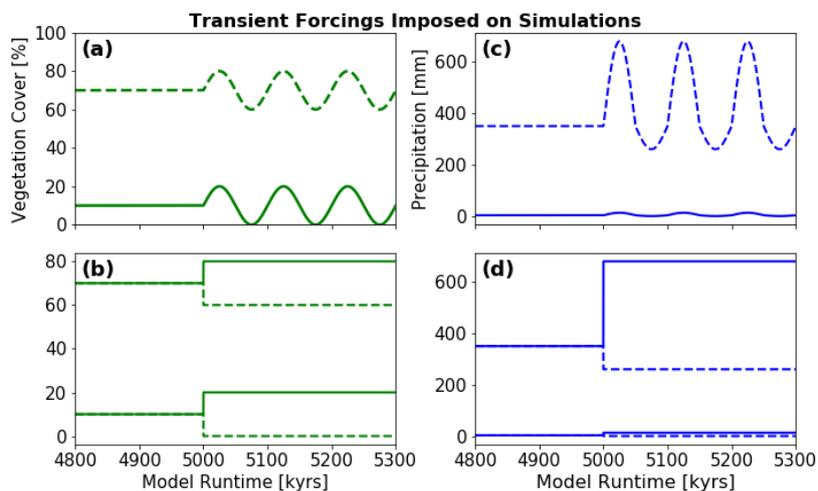
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880 **Figure 3** Example model setup used in simulations in this study. Boundary conditions and parameterizations used in the models  
 881 are labeled. Blue colors represent low elevations, brown colors represent higher elevations. Additional details of parameters  
 882 used are specified in Table 1.

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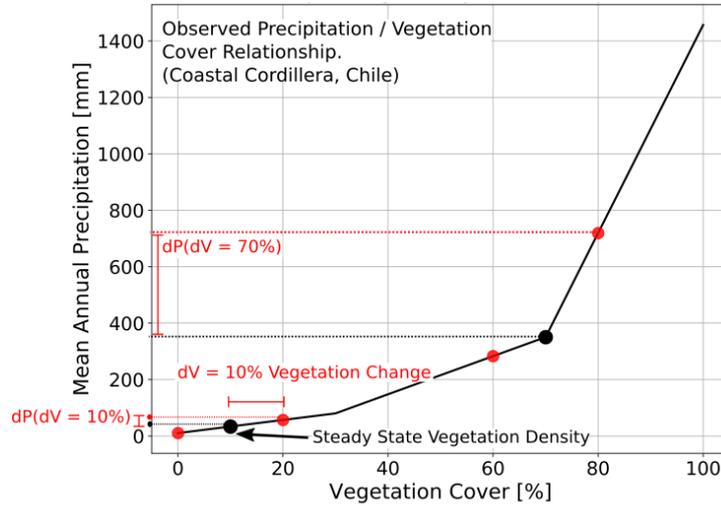
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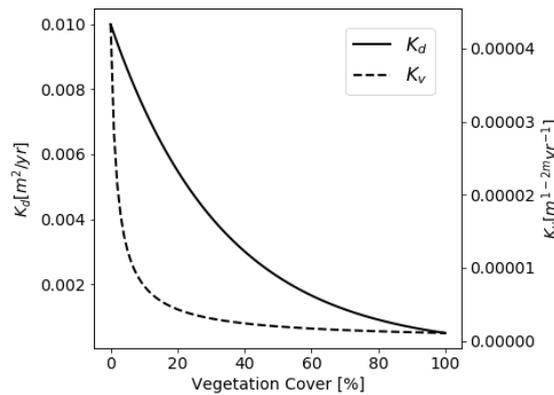
887 **Figure 4** Transient forcings in vegetation (a,b) and precipitation (c,d) considered in model experiments. Simulations were run  
 888 for 15 Myr prior to the runtime show in the figure. All transients imposed started a runtime of 5 Myr. A) Variations in  
 889 vegetation cover imposed in the oscillating experiment conditions for initial vegetation cover of 10 and 70%. B) Positive and  
 890 negative step change parameterizations for vegetation cover. C) Variations in mean oscillating annual precipitation. D) Positive  
 891 and negative step changes in mean annual precipitation used in experiments. The initial precipitation amounts prior to the  
 892 transient oscillations or step changes correspond to the observed precipitation corresponding to the vegetation cover in each of  
 893 the observed study areas (Fig. 1). See also Figure 5.

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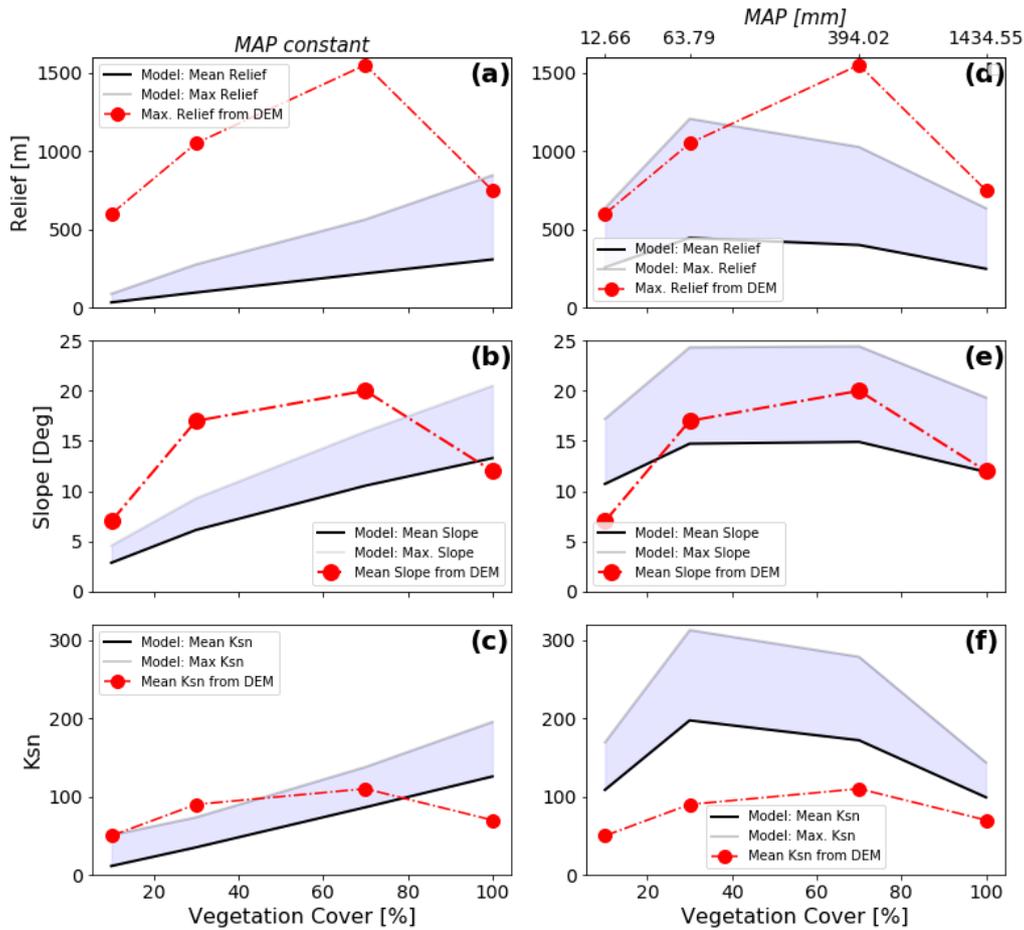
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 896 **Figure 5** Graphical representation of the observed precipitation – vegetation relationship in the focus areas (Fig. 1) and how  
 897 precipitation amounts were selected when perturbations in vegetation cover were imposed. Black dots represent vegetation-  
 898 precipitation values used in the steady-state model conditions and prior to any transients. Red dots show how vegetation cover  
 899 perturbations in +/- 10% in the model simulations were used to select corresponding mean annual precipitation amounts.

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 903 **Figure 6** Predicted values of hillslope diffusivity  $K_d$  (solid line) and fluvial erodibility  $K_v$  (dashed line) as a function of vegetation  
 904 surface cover. Although absolute values can't be compared due to different units, the shape of the curves representing the  
 905 different parameters show different sensitivities to changes in vegetation cover. Fluvial erodibility shows highest magnitude of  
 906 change for vegetation cover values < 25% whereas hillslope diffusivity reacts in a more linearly with highest change below <  
 907 65% vegetation cover.

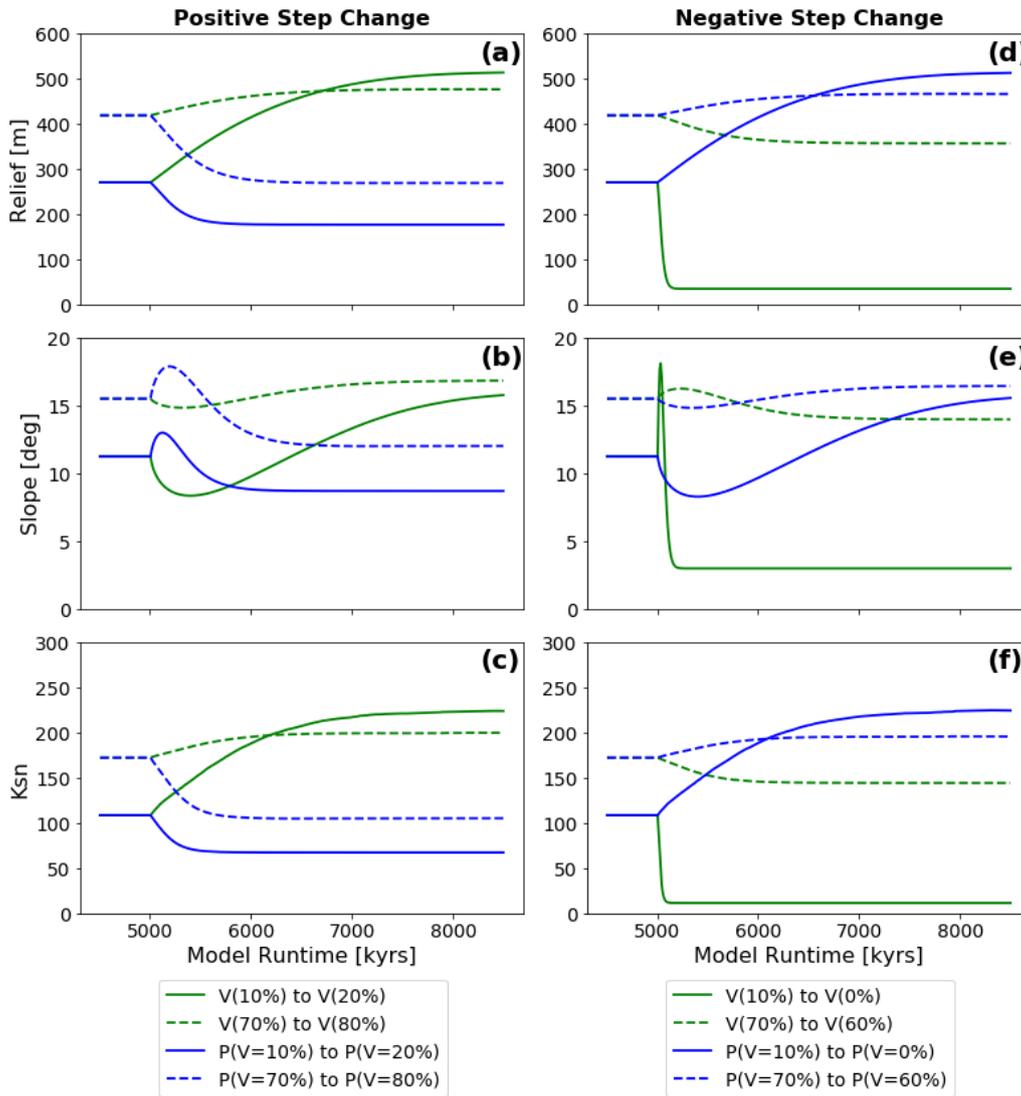
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913 **Figure 7** Steady-state model predicted (shaded regions) and observed (red dots) topographic metrics from the study areas  
 914 shown in Figure 1 for different vegetation cover amounts. Observed topographic metrics were extracted from SRTM 90 m  
 915 DEM. Model predicted values are shown for the cases of constant mean annual precipitation (a,b,c) or variable precipitation  
 916 (D,E,F). Variable precipitation rates and vegetation covers were selected for these simulations using the observed values from  
 917 the focus areas (Fig. 5). Note that for variable precipitation and vegetation cover simulations (d,e,f) the predicted values (similar  
 918 to the observations) develop a humped shape pattern of an increase and then decrease in each parameter suggesting the changes  
 919 in both precipitation and vegetation cover are needed to reproduce the general trend seen in observations. The sources of misfit  
 920 between the predicted and observed values are due to the simplified (and untuned) setup of the simulations and discussed in  
 921 the text.

922

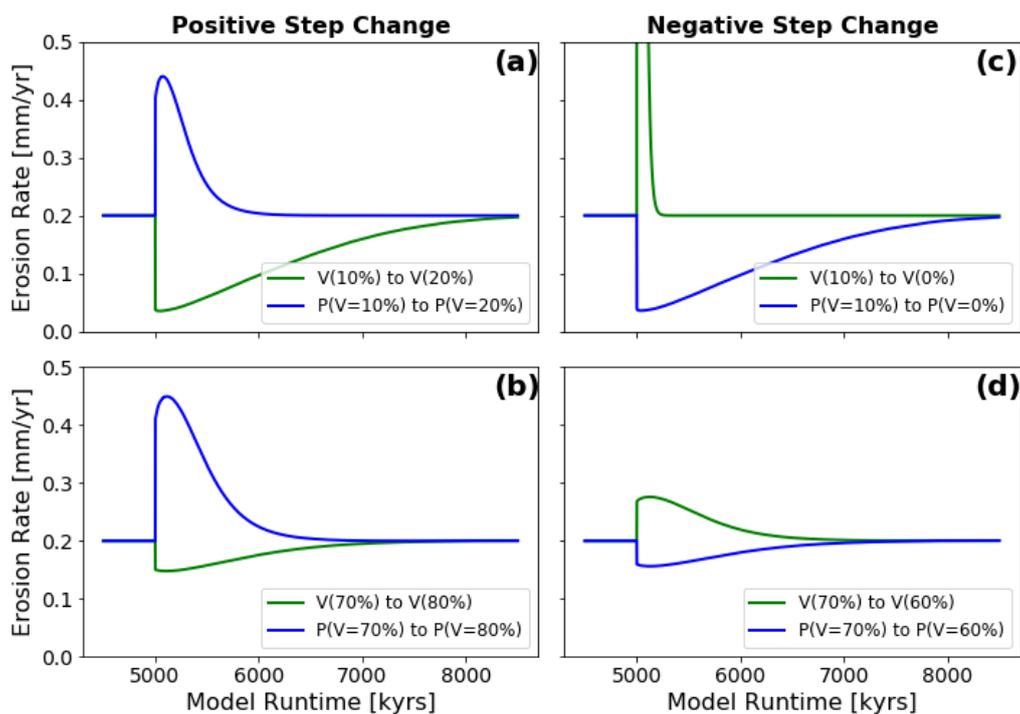


923

924 **Figure 8** Observed evolution of topographic metrics after a step-change in either vegetation (green lines) or mean annual  
 925 precipitation (blue lines). Results are shown for two different initial vegetation cover amounts of V=10 and 70%. Imposed mean  
 926 annual precipitation changes were done by selecting the precipitation amount corresponding to the initial and final vegetation  
 927 amounts used in the simulations for vegetation cover 'only' change. Panels a,b,c show reaction of model topographies to positive  
 928 changes in boundary conditions, panels d,e,f show reaction to negative changes in boundary conditions.

929

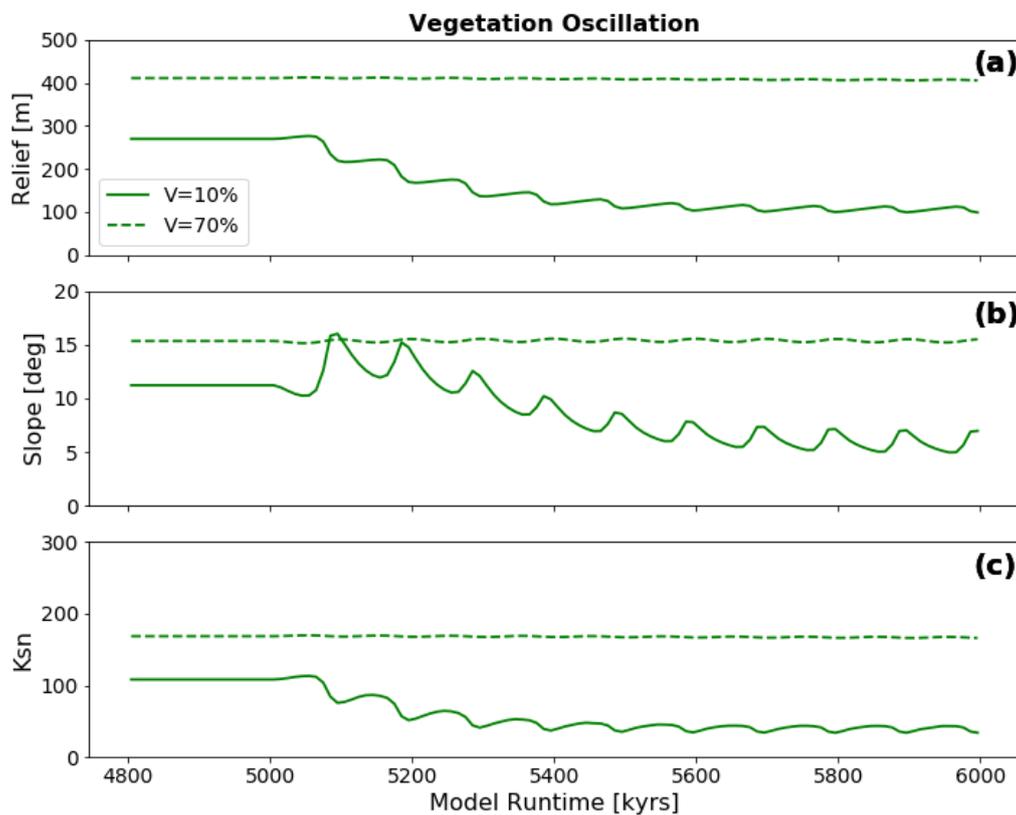
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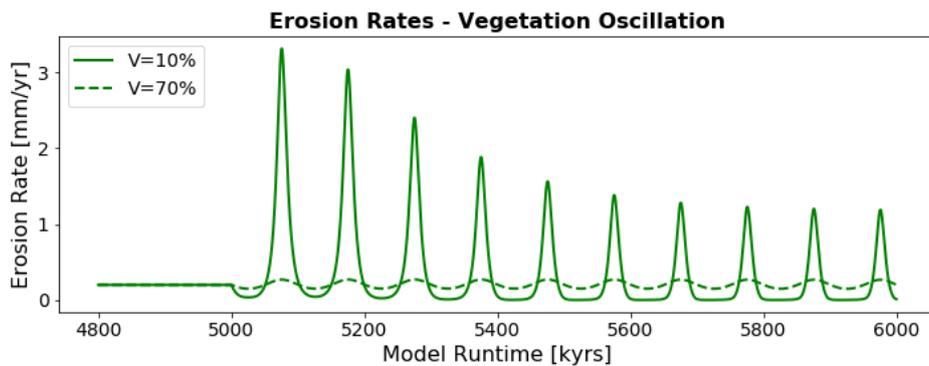
932 **Figure 9** Mean catchment-wide erosion rates after step-change disturbance in model boundary conditions. Blue lines represent  
933 erosion rates for models with changes in only precipitation, green lines represent erosion rates for models with changes in only  
934 vegetation cover. Panels a,b show evolution after positive step-change, panels c,d for models with negative step-change. Note  
935 that the direction of change (positive or negative) from the initial state is in opposite directions for precipitation and vegetation  
936 cover changes. This effect is manifested in the subsequent plots.

937



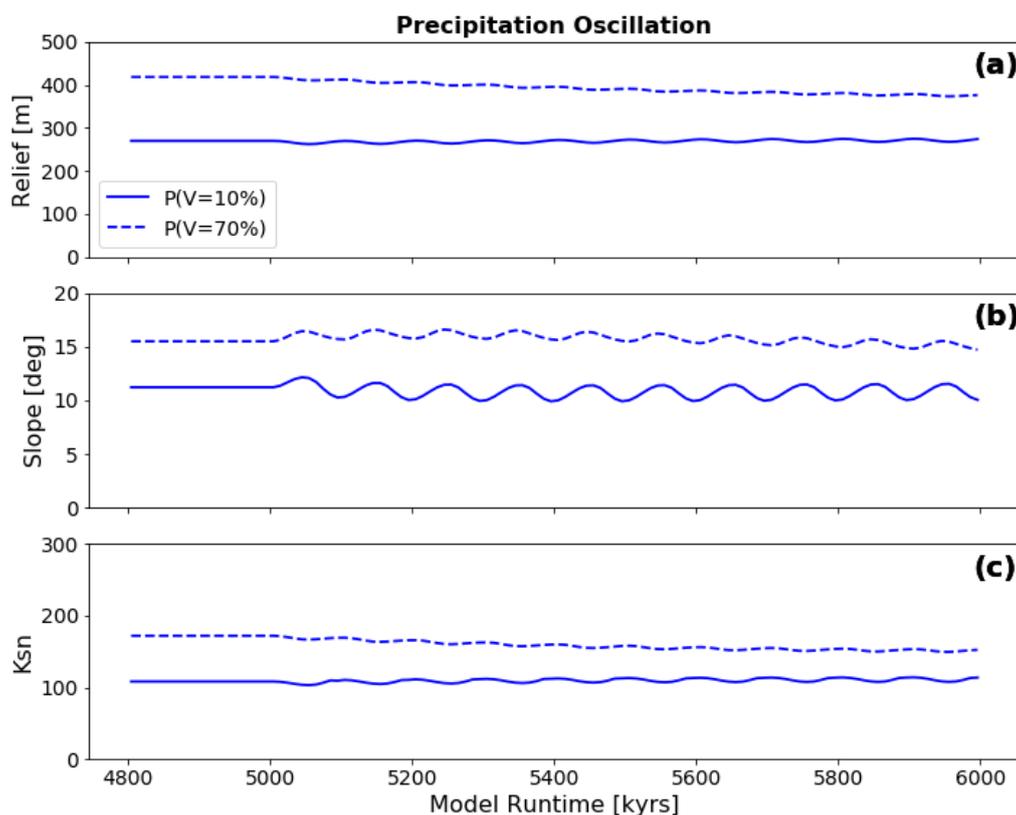
938

939 **Figure 10 Evolution of topographic metrics for simulations with oscillating surface vegetation cover and constant precipitation**  
940 **corresponding to the initial vegetation cover prior to the transient in vegetation cover. Panels a,b,c show mean basin relief,**  
941 **mean basin slope and mean basin channel steepness ( $k_{sn}$ ), respectively.**



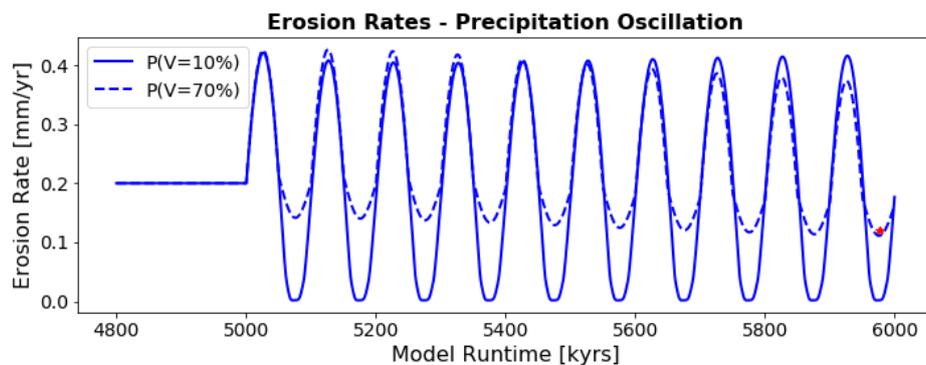
942

943 **Figure 11 Predicted mean catchment erosion rates for simulations with oscillating surface vegetation cover and constant**  
944 **precipitation.**



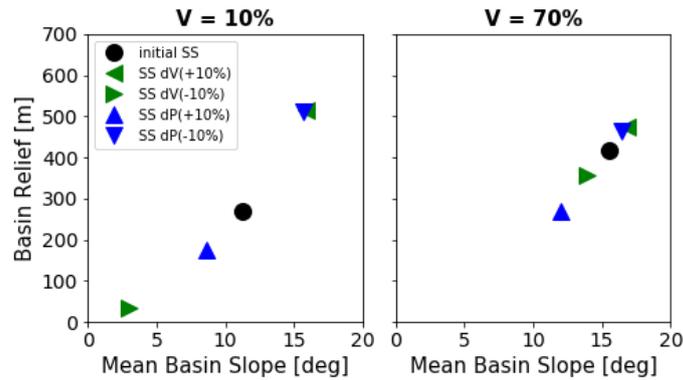
945

946 Figure 12 Evolution of topographic metrics for simulations with oscillating mean annual precipitation and constant vegetation  
947 cover. The vegetation cover was held constant at the value corresponding to the precipitation rate prior to the onset of the  
948 transient at 5000 kyrs. Panels a,b,c show mean basin relief, mean basin slope and mean basin channel steepness ( $k_{sn}$ ),  
949 respectively.



950

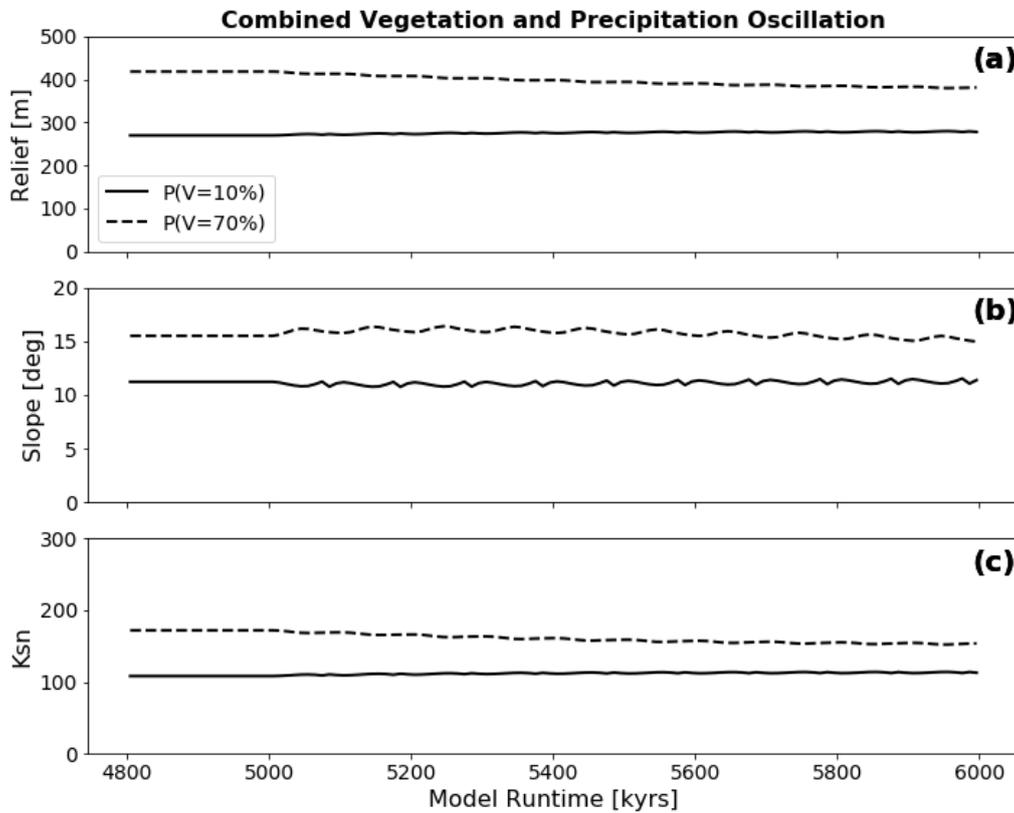
951 Figure 13 Mean catchment erosion rates for simulations with oscillating mean annual precipitation and constant surface  
952 vegetation cover.



953

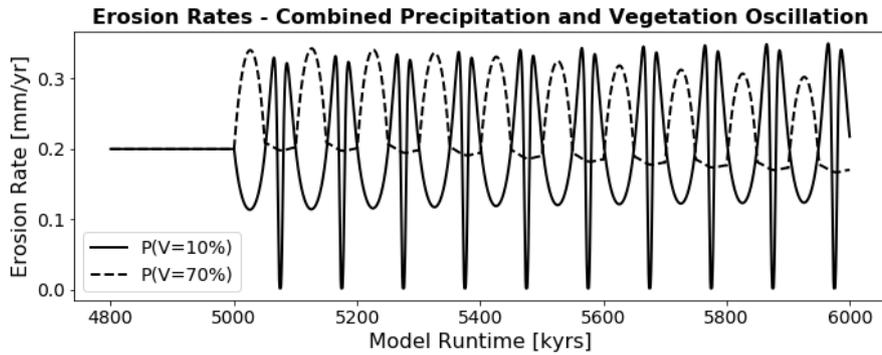
954 **Figure 14** Shifts in mean basin slope/mean basin relief relationship for simulations with positive and negative step-changes in  
 955 either vegetation cover (green triangles) or mean annual precipitation (blue triangles). Black dots represent initial steady-state  
 956 conditions prior to any imposed transient in vegetation cover or mean annual precipitation.

957



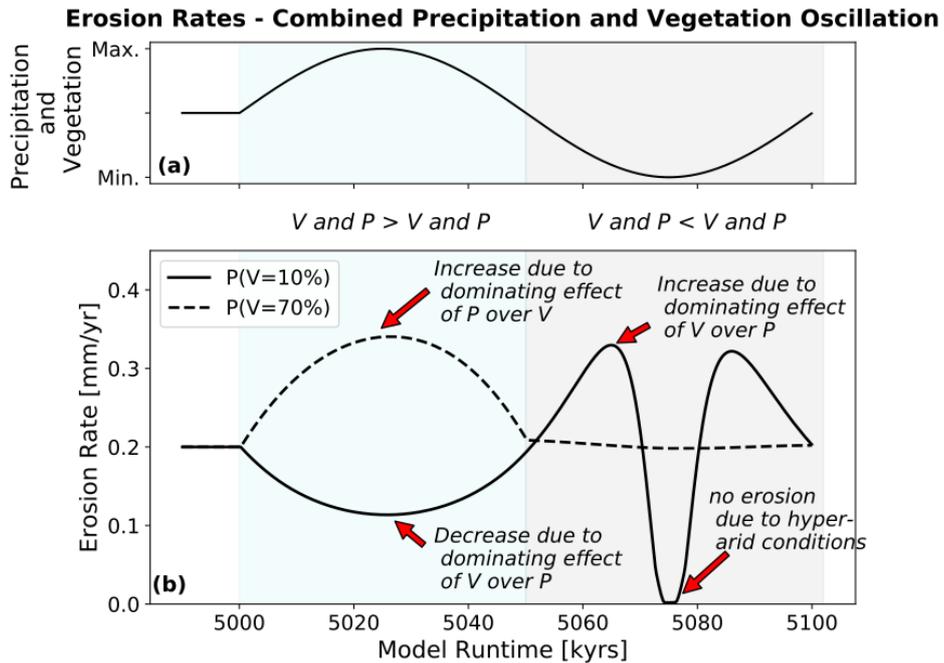
958

959 **Figure 15** Evolution of topographic metrics for coupled simulations where both changes in surface vegetation cover and a  
 960 corresponding change (Fig. 5) in mean annual precipitation are simultaneously imposed. Panels a,b,c show evolution of mean  
 961 basin relief, mean basin slope and mean basin channel steepness ( $k_{sn}$ ) after start of oscillation at 5Ma. Note the muted/damped  
 962 response relative to previous simulations of oscillating vegetation cover or precipitation conditions.



963

964 **Figure 16** Mean catchment erosion rates for coupled simulations with changes in surface vegetation cover and mean annual  
 965 precipitation. The first cycle in the time series is expanded in Figure 17.



966

967 **Figure 17** Mean catchment erosion rates for coupled simulations for one period of oscillation after the start of transient  
 968 conditions (see also Fig. 15). Upper subplot shows conceptualized transient forcing in vegetation cover and mean annual  
 969 precipitation, lower subplot shows erosion rate for simulations with low (black line) and high (dotted line) in initial vegetation  
 970 cover and precipitation values.

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978 **Table 1 Model parameters used for conduction simulation experiments with Landlab model environment.**

Model Parameter	Value
Uplift (U), mm/yr	0.2
Base Fluvial erodibility, m/yr (kg m-1 s-2)-p	7.00E-06
Fluvial Threshold Erosion Rate, L/T	4.00E-04
Base Diffusivity Constant, m <sup>2</sup> /yr	0.02
Vegetation Oscillation Amplitude, %	+ - 10%
Mannings Number for fully vegetated conditions , -	0.01
w	1
alpha	0.3
p	1

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980