Influence of lithology on hillslope morphology and response to tectonic forcing in the northern Sierra Nevada of California

Martin D. Hurst,1,2 Simon M. Mudd,1,3 Kyungsoo Yoo,4 Mikael Attal,1 and Rachel Walcott1,5

Received 6 February 2012; revised 13 February 2013; accepted 15 February 2013.

1Many geomorphic studies assume that bedrock geology is not a first-order control on landscape form in order to isolate drivers of geomorphic change (e.g., climate or tectonics). Yet underlying geology may influence the efficacy of soil production and sediment transport on hillslopes. We performed quantitative analysis of LiDAR digital terrain models to examine the topographic form of hillslopes in two distinct lithologies in the Feather River catchment in northern California, a granodiorite pluton and metamorphosed volcanics. The two sites, separated by <2 km and spanning similar elevations, were assumed to have similar climatic histories and are experiencing a transience in landscape evolution characterized by a propagating incision wave in response to accelerated surface uplift c. 5 Ma. Responding to increased incision rates, hillslopes in granodiorite tend to have morphology similar to model predictions for steady state hillslopes, suggesting that they adjust rapidly to keep pace with the incision wave. By contrast, hillslopes in metavolcanics exhibit high gradients but lower hilltop curvature indicative of ongoing transient adjustment to incision. We used existing erosion rate data and the curvature of hillslopes proximal to the main channels (where hillslopes have most likely adjusted to accelerated erosion rates) to demonstrate that the sediment transport coefficient is higher in granodiorite (8.8 m² ka⁻¹) than in metavolcanics (4.8 m² ka⁻¹). Hillslopes in both lithologies get shorter (i.e., drainage density increases) with increasing erosion rates.

Citation: Hurst, M. D., S. M. Mudd, K. Yoo, M. Attal, and R. Walcott (2013), Influence of lithology on hillslope morphology and response to tectonic forcing in the northern Sierra Nevada of California, J. Geophys. Res. Earth Surf., 118, doi:10.1002/jgrf.20049.

1. Introduction

2Climate and tectons act in concert to control the morphology of the Earth’s surface. The ability to quantify relationships between topography and climatic or tectonic driving processes is dependent on understanding how efficiently, and by which processes, sediment is generated and transported on hillslopes and in valleys [e.g., Ahnert, 1970; Dietrich et al., 2003]. Such knowledge is vital for ongoing modeling efforts which help link empirical observations to theoretical predictions [e.g., Tucker and Hancock, 2010]. Hillslope processes control the flux and caliber of sediment supplied to streams [e.g., Whittaker et al., 2010], which subsequently influence fluvial incision rates [e.g., Sklar and Dietrich, 2004] and the rate at which sediment is delivered to basins [e.g., Duller et al., 2010; Armitage et al., 2011].

3Tectonic processes redistribute rock mass within the lithosphere and control the type and flux of rock material exhumed to the surface. This material may be weakened at depth via mechanical fracturing due to tectonic processes [e.g., Molnar et al., 2007] and later disrupted at/near the surface by physical and chemical weathering processes (e.g., penetration and growth of tree roots [Roering et al., 2010; Gabet and Mudd, 2010], frost wedging [e.g., Small et al., 1999], and chemical alteration and weakening [e.g., Burke et al., 2007; Dixon et al., 2009]). These processes generate soil/regolith, here used synonymously and defined as material at or near the Earth’s surface that is being physically disturbed (equivalent to the physically disturbed zone as defined by Yoo and Mudd [2008]) and which can be subsequently transported away. The type and physical properties (composition, rock mass strength, and degree of fracturing) of bedrock will influence the physical properties of the resulting soil (e.g., composition, grain size distribution, degree of weathering, porosity, and cohesion [Yoo et al., 2005]). The physical characteristics of soil are in turn expected to influence the efficacy at which sediment...
transport occurs on hillslopes [Furbish et al., 2009]. Thus, there is the potential for bedrock lithology to influence topography even in soil-mantled landscapes.

Several studies have attempted to quantify relationships between rock strength and topography, or have considered the role that spatially variable rock type may have in controlling processes which generate and redistribute sediment at the Earth’s surface. Schmidt and Montgomery [1995] demonstrated that hillslope relief is limited by the material strength of bedrock. Similarly, Burbank et al. [1996] suggested that hillslope gradients were limited at their internal friction angle, despite variation of over an order of magnitude in denudation rates in the northwest Himalayas (1–12 mm a⁻¹). Hillslopes were thus interpreted to evolve in response to variable erosion rates by adjusting the frequency of landslides rather than by steepening. The distribution of slope angles may serve as a proxy for rock mass strength in landslide-dominated terrain as demonstrated by Korup [2008] and Korup and Schlunegger [2009]. Clarke and Burbank [2010] however were not able to distinguish rock mass strength from hillslope gradients in distinct lithologies in Southern New Zealand. They attributed the similarities between low-grade metamorphics of the Southern Alps and high-grade and igneous units of the Fiordland to the nature of bedrock fracturing. Both sites are susceptible to landsliding, but different styles of fracturing were interpreted to control the type of mass wasting process operating. In Fiordland, fracturing occurs primarily due to near surface processes and thus drives frequent shallow landslides, whereas pervasive tectonic fracturing in the Southern Alps facilitates larger, deeper landsides [Clarke and Burbank, 2010; 2011]. Although links between bedrock lithology and topography have been explored in bedrock landscapes, no studies have explored the role that lithology may play in controlling topography in soil-mantled landscapes.

Lithology may play an important role in controlling the efficiency of sediment transport on hillslopes. McKean et al. [1993] showed that the sediment transport coefficient $D$, which relates hillslope gradient to sediment flux, is an order of magnitude larger in weak clay-rich soils than in strong, granular soils, presumably due to variation in the efficiency of shrink-swell cycles as a transport process. This is at least partially controlled by the parent lithology through the nature of jointing and susceptibility to weathering. Furbish et al. [2009] described the sediment transport coefficient $D$ as a function of the material properties of soil including thickness, grain size distribution, and cohesion which may directly influence the efficiency of sediment transport. The presence of coarse material in the soil may result in a boulder lag which armours underlying soil from erosion [e.g., Granger et al., 2001]. Owen et al., 2010 demonstrated that hillslope erosion rates across a climate gradient in Chile were sensitive to precipitation, which influences transport processes, with more rapid erosion rate attributed to wetter climate and biologically driven sediment transport. To assess whether there is any existing evidence that $D$ might be influenced by lithology, we compiled published values of $D$ to search for global trends with lithology (Figure 1a and Table 1). Simplifying to cohesionless, clastic, volcanic, and crystalline lithologic groups, we were not able to observe any trends between lithology and the sediment transport coefficient (Figure 1b). However, isolating for values derived from cohesionless substrate (i.e., alluvium; $n = 24$), we observe that $D$ increases with precipitation and

![Figure 1](https://example.com/figure1.png)
| Study | Site | MAT (°C) | MAP (mm) | Vegetation | Brief Description | Climatea | \(D \text{ (m}^2 \text{ka}^{-1})^b\) |
|-------|------|----------|----------|------------|------------------|---------|------------------|
| Almond et al. [2008] | Charwell Basin, New Zealand | 10.6 ± 3.6 | 1159 ± 35 | Grassland/shrubland | Fluvial gravel terraces | 3.0 ± 1.0 |
| Almond et al. [2008] | Charwell Basin, New Zealand | 11.8 ± 3.9 | 662 ± 22 | Grassland/shrubland | Thick loess deposits (underlain by altered basalt) | 3.1 ± 0.4 |
| Arrowsmith et al. [1998] | Carrizo Plain, CA, USA | 14.5 ± 6.2 | 475 ± 73 | Grasses and shrubs | Fault scarps in alluvial gravel | 8.6 ± 0.8 |
| Avouac et al. [1993] | Tien Shan, China | 9.5 ± 14.2 | 138 ± 13 | Gravels and shrubs | Fault scarps in alluvial gravel | 5.4 ± 0.2 |
| Begin [1992] | Northen Negev, Israel | 19.6 ± 5.0 | 193 ± 36 | Not vegetated | Fluvial gravel terraces | 0.4 ± 0.3 |
| Bowman and Gerson [1986] | Lake Lisan, Dead Sea | 24.3 ± 6.5 | 142 ± 27 | Not vegetated | Lake terraces | 0.4 |
| Bowman and Gross [1989] | Northern Arava, Israel | 18.8 ± 5.5 | 198 ± 37 | Not vegetated | Fault scarps in alluvial gravel | <0.4 |
| Carreter et al. [2002] | Gurvan Bugd fault system, Mongolia | 0.3 ± 12.2 | 160 ± 33 | Not vegetated | | Fault scarps in alluvial gravel | 3.3 ± 1.7 |
| Colman and Watson [1983] | Lake Bonneville, UT, USA | 9.5 ± 9.2 | 456 ± 19 | Grasses and shrubs | Alluvial shoreline scarps | 0.9 |
| Constantine et al. [2012] | Simulated Douglas Fir Forest | N/A | N/A | Douglas Fir | N/A | 0.1–3.5 |
| Gabet [2000] | Transverse Ranges, CA, USA | 15.0 ± 4.0 | 498 ± 83 | Coastal Sage | Pistio-Pleistocene fanglomerates, Process specific: gopher bioturbation | 7.4 |
| Gabet [2003] | Transverse Ranges, CA, USA | 15.0 ± 4.0 | 498 ± 83 | Coastal Sage | Pistio-Pleistocene fanglomerates. | 0.17 |
| Hanks et al. [1984] | Lake Bonneville, UT, USA | 9.5 ± 9.2 | 456 ± 19 | Grasses and shrubs | Alluvial shoreline scarps | 1.1 |
| Hanks et al. [1984] | Santa Cruz sea cliffs, CA, USA | 13.8 ± 2.8 | 693 ± 111 | Lower terraces are farmed, upper terraces grassland | Quaternary wave-cut terraces cut into Pleistocene mudstone | 11 |
| Hanks et al. [1984] | Raymond Fault Scarp, LA, CA, USA | 18.1 ± 4.2 | 450 ± 76 | Not reported | Fault scarps in alluvial gravel | 16 |
| Hanks et al. [1984] | Drum Mtns., UT, USA | 10.1 ± 9.9 | 192 ± 10 | Low shrubs (sagebrush and shadscale) | Fault scarps in alluvial gravel | 1.1 |
| Hanks and Wallace [1985] | Lake Lahonta, NV, USA | 10.0 ± 8.5 | 202 ± 9 | Not reported | Alluvial shoreline scarps | 1.1 |
| Hanks [2000] | Lost River, ID, USA | 3.3 ± 9.1 | 270 ± 16 | Not reported | Fault scarps in alluvial gravel | 0.9–1.0 |
| Heimsath et al. [2000] | Nunnock River, SE Australia | 12.8 ± 4.4 | 827 ± 32 | Schlerophyll forest | Soil-mantled granite | 4.0 |
| Heimsath et al. [2005] | Nunnock River, SE Australia | 12.8 ± 4.4 | 827 ± 32 | Schlerophyll forest | Soil-mantled granite | 4.0 |
| Hanks et al. [1984] | Point Reyes, CA, USA | 12.9 ± 3.9 | 977 ± 153 | Bishop pine forest | Soil-mantled granite | 4.0–5.5 m ka⁻¹ |
| Hughes et al. [2009] | Charwell Basin, New Zealand | 10.6 ± 3.6 | 1159 ± 35 | Shrubland/grassland (late Pliocene) | Fluvial gravel terraces (underlain by greywacke) | 4.7 ± 2.0 |
| Hurst et al. [2012] | Feather River, CA, USA | 13.2 ± 6.5 | 1508 ± 217 | Mixed conifer forest | Soil-mantled granitoids | 8.0 |
| Jungers et al. [2009] | Great Smokey Mountains, NC, USA | 8.4 ± 7.3 | 1855 ± 24.1 | Deciduous forest | Soil-mantled quartzite | 6.5–10 |
| Martin and Church [1997] | Various | N/A | N/A | Various | From field measurements of volumetric creep rates | 0.2 |
| Mattson and Bruhn [2001] | Lake Bonneville, UT, USA | 9.5 ± 9.2 | 456 ± 19 | Grasses and shrubs | Alluvial shoreline scarps | 1.2 ± 0.3 |
| Mattson and Bruhn [2001] | Wasatch Fault Zone, UT, USA | 9.5 ± 9.3 | 420 ± 17 | Not reported | Fault scarps in alluvial gravel | 2.8 ± 1.1 |
| McKeen et al. [1993] | East Bay Regional Park, CA, USA | 14.9 ± 5.4 | 522 ± 79 | Grasses, clay-rich soil | Soil-mantled Eocene marine shale | 36 ± 5 |
| Nash [1980a] | Emmet County, MI, USA | 5.7 ± 10.1 | 825 ± 35 | Mixed pine, oak, beech forest | Wave-cut terraces in moraine | 12 |
| Nash [1980b] | Drum Mtns., UT, USA | 10.1 ± 9.9 | 192 ± 22 | Low shrubs (sagebrush and shadscale) | Fault scarps in alluvial gravel | 0.4 |
| Nash [1984] | Hebgen Lake, MT, USA | 1.9 ± 9.6 | 615 ± 22 | Not reported | Fault scarps in alluvial gravel and fluvial gravel terraces | 2.0 ± 0.4 |
| Nivière and Marquis [2000] | Upper Rhine Graben, Germany | 10.2 ± 6.6 | 707 ± 32 | Forested | Fluvial gravel terraces | 1.4 |
| Pelletier et al. [2006] | Lake Bonneville, UT, USA | 9.5 ± 9.2 | 456 ± 19 | Grasses and shrubs | Alluvial shoreline scarps | 1.0 |
Table 1. (continued)

| Study                  | Site                      | Climate\(^a\) | Vegetation Brief Description | Brief Description                                      | \(D (\text{m}^2 \text{ka}^{-1})\)\(^b\) |
|------------------------|---------------------------|----------------|-------------------------------|--------------------------------------------------------|----------------------------------------|
| Pelletier and Cline [2007] | Lathrop Wells, NV, USA    | 17.6 ± 8.9     | Little vegetation             | Loose vesicular scoria lapilli                          | 3.9                                    |
| Pelletier et al. [2011] | Banco Bonito lava flow, Valles | 4.9 ± 8.4     | Pine, oak, and mixed conifer forest | Soil-mantled tholothic lava flow                        | \(D = 0.5 \pm 0.2\) \(D_a = 0.55 \pm 0.35\) |
| Calera, NM, USA        |                           | 4.9 ± 8.4      |                               |                                                        |                                        |
| Petit et al. [2009]    | Wasatch Mtns., UT, USA    | 8.4 ± 9.3      | Not reported                  | Soil-mantled gneiss (Percambrian)                      | 120 ± 10                               |
| Pierce and Colman [1986] | Big Lost River Valley, ID, USA | 5.2 ± 10.0    | Not reported                  | Alluvial fan scarps                                    | \(D = 0.2 \rightarrow 7.0\)           |
| Renoir [1988] reported in Heimsath et al. [2005] | Tennessee Valley | 13.8 ± 2.8     | Coastal grassland and scrub   | Soil-mantled deep marine metasedimentary                | 5.0                                    |
| Renoir et al. [1989]   | Point Reyes              | 12.9 ± 2.9     | Bishop pine forest            | Soil-mantled granite                                   | 3.0                                    |
|                        | Clearwater River, WA, USA | 9.0 ± 4.5      | Western hemlock and Pacific silver fir forest | Soil-mantled deformed tertiary silts, sandstones, and conglomerates | 4.7 ± 2.5                             |
| Riggins et al. [2011]  | Bodmin Moor, Cornwall, UK | 9.2 ± 4.5      | Grasses, (previously hazel, and oak woodland) | Soil-mantled coarse-grained granite (Permian)         | 46 ± 16                                |
| Roering et al. [1999]  | Sullivan Creek, OR, USA   | 10.8 ± 3.5     | Douglas fir, mixed conifer forest | Soil-mantled turbidite beds                            | 3.0                                    |
| Roering et al. [2001b] | Experimental             | N/A            | N/A                           | Sand pike disturbed by acoustics                        | 0.27 ± 0.02                            |
| Roering et al. [2002]  | Charwell River, South Island, | 10.6 ± 3.6    | Podocarp and beech forest     | Fluvial gravel terraces (underlain by greywacke)      | 12 ± 8                                 |
| Roering et al. [2004]  | New Zealand              | N/A            | N/A                           |                                                        | 16 ± 6                                 |
| Roering and Gerber [2005] | Sullivan Creek, OR, USA   | N/A            | N/A                           | Same site as Roering et al. [1999] but post-wildfire   | 11 ± 3.5                               |
| Roering et al. [2007]  | Gabilan Mesa, CA, USA    | 14.7 ± 5.1     | Oak Savannah                  | Soil-mantled shallow marine and alluvial sediment      | \(D\) in range 38 ± 15 (±40–24)        |
| Rosenbloom and Anderson [1994] | Santa Cruz, CA, USA       | 13.5 ± 2.6     | Lower terraces are farmed, upper terraces grassland | Marine terraces cut into Miocene marine mudstone      | 10                                     |
| Small et al. [1999]    | Wind River Range, WY, USA | −4.4 ± 7.7     | Not vegetated                 | Soil-mantled crystalline bedrock                       | 17.5 ± 2.7                             |
| Spelz et al. [2008]    | Laguna Salada, Baja California, Mexico | 21.6 ± 6.3   | Not vegetated                 | Fault scarp in fluvial gravel terraces                 | 0.051–0.066                           |
| Tapponeer et al. [1990] | Qilian Shan, China        | 5.9 ± 10.8     | Not vegetated                 | Fault scarp in Quaternary fanglomerates                | 3.3 ± 1.7                              |
| This study             | Feather River, CA, USA    | 13.2 ± 6.5     | Mixed conifer forest          | Soil-mantled intermediate metavolcanics                | 4.8 ± 1.8                              |
| Walther et al. [2009]  | Blue Mountains, WA, USA   | −0.7 ± 15.6    | Coniferous forest             | Soil-mantled granodiorite                              | 8.8 ± 3.3                              |

\(^a\)Mean annual temperature (MAT) and mean annual precipitation (MAP) calculated over the period 1950–2000. Data from the WorldClim global climate 30 arc-second dataset (http://www.worldclim.org; accessed 6 August 2011) (Hijmans et al., 2005). Error ranges are 1σ of monthly means as an indicator of seasonality.

\(^b\)We report absolute values, range constraints, and/or error estimates as they appear in the literature.

\(^c\)Pierce and Colman [1986] found that \(D\) ranged over two orders of magnitude depending on the scarp aspect.

\(^d\)Large range due to poor constraints on boundary erosion rate.
seasonality (Figures 1b and 1c; seasonality defined here as the standard deviation about mean annual precipitation), as suggested by Hanks [2000]. Yet it seems likely that both substrate lithology and climate will control D, since lithology will influence the production and material properties of the soil, and climate will control the style and efficiency of processes which mobilize the soil. If lithology can significantly influence hillslope sediment transport, we anticipate differences in landscape morphology for adjacent areas (with similar climate) overlying distinct bedrock types, even if the mechanisms of sediment transport are similar.

In soil-mantled, forested landscapes, the dominant mechanism of sediment flux is often via tree throw, and the growth and decay of tree roots [e.g., Roering et al., 2010; Gabet and Mudd, 2010; Constantine et al., 2012]. The efficiency of sediment transport may therefore be strongly linked to the amount and type of vegetation acting to disturb sediment. Hughes et al. [2009] inferred that sediment transport increased at the start of the Holocene due to colonization by forests, replacing previous grassland in the Charwell Basin, New Zealand (Table 1). Light Detection and Ranging (LiDAR) allows for quantification of metrics for aboveground biomass, such as vegetation density or mean canopy height [e.g., Nilsson, 1996; Naesset, 1997; Lefsky et al., 2002; Holmgren et al., 2003; Donoghue and Watt, 2006], which can be compared to topographic attributes to explore whether D may vary systematically as a function of vegetation [Pelletier et al., 2011].

The morphology of soil-mantled hillslopes reflects the processes which create and redistribute sediment downslope, and the erosion rate in the adjacent channels. Where constraints have been placed on erosion rates within a landscape, one may infer the nature of sediment transport based on the morphological properties of hillslopes such as hilltop curvature $C_{HT}$, mean hillslope gradient $S$, and hillslope length $L_H$ [e.g., Roering et al., 2007; Roering, 2008; Hurst et al., 2012]. In this contribution, we extract these properties from high resolution (1 m grid) topography, derived from airborne LiDAR, to compare the topographic signature of landscapes in two distinct lithologies, the granodiorite of the Cascade pluton and the metavolcanic rocks of the Central Belt in the northern Sierra Nevada of California. In this region, we have constraints on rock uplift rates and associated transient erosion rates [Riebe et al., 2000; Wakabayashi and Sawyer, 2001; Hurst et al., 2012], present day climate, and current vegetation. Hence, there is an opportunity to quantitatively analyze the morphological properties of rapidly denuding, soil-mantled hillslopes in two distinct lithologies in order to investigate the control bedrock type plays in a soil-mantled landscape. We sought to quantify the efficiency of sediment transport processes on hillslopes in two distinct lithologies and identify whether any differences could be attributed to climate, the type and distribution of vegetation, or underlying lithology. We investigated differences in the distribution of hillslope gradients between bedrock types. Finally, we documented differences in the length of hillslopes between the two lithologies, and a tendency for hillslope lengths to shorten with increasing erosion rate, suggesting that drainage density may be controlled by erosion rate.

2. Theory on Hillslope Morphology

2.1. Hillslope Mass Balance and Sediment Transport Equations

The spatial and temporal evolution of soil-mantled landscapes can be examined using principles of mass conservation [Gilbert, 1909; Culling, 1960; Dietrich et al., 2003] where the surface elevation $z$ [L] changes in time $t$ relative to a moving reference elevation $z_0$ [L] [e.g., Mudd and Furbish, 2005]. The surface elevation evolves according to the following:

$$\frac{dz}{dt} = -\frac{d_0}{dt} \nabla q_s,$$

where $q_s$ [L$^2$ T$^{-1}$] is the volumetric sediment flux per unit contour width. We equate the lowering rate of the reference elevation $z_0$ to the rate of local bedrock lowering (i.e., valley incision) at the base of the hillslope ($E$ [L T$^{-1}$]) such that $dz_0/dt = -(\rho_r/\rho_s)E$ where $\rho_r$ and $\rho_s$ [M L$^{-3}$] are the densities of bedrock and dry soil, respectively. If the entire hillslope lowers at the same rate as the channel, then equation (1) reduces to

$$\frac{\rho_r}{\rho_s} E = \nabla q_s. \tag{2}$$

Equations (1) and (2) assume that all mass transport is the result of physical processes, and mass/volume change due to aeolian processes is negligible. We do not account for volume changes in soils due to chemical denudation. Riebe et al. [2001] demonstrated that chemical denudation scales with total denudation in granitoid portions of our study area and that chemical denudation is small compared to physical denudation and should have minimal impact on hillslope morphology [e.g., Mudd and Furbish, 2004].

Most processes which act to transport sediment down a hillslope are gravity driven and are therefore dependent on hillslope angle; both grain displacements during disturbance (normal to the surface) and subsequent gravitational settling (vertical) increase with steeper slopes [e.g., Roering et al., 1999; Furbish et al., 2009]. On gentle, soil-mantled hillslopes, sediment flux, $q_s$, is often attributed to slope-dependent creep-like processes [Davis, 1892; Gilbert 1909]. However, in landscapes with high relief, hillslopes often become planar away from topographic divides, commonly inferred to be driven by a process transition to landslide-dominated sediment flux [e.g., Howard, 1994; Roering et al., 1999; Binnie et al., 2007] and/or an increase in particle displacement distances [e.g., Tucker and Bradley, 2010; Foufoula-Georgiou et al., 2010]. Roering et al. [1999] formulated a disturbance-driven transport law allowing sediment flux to increase in a non-linear fashion with hillslope gradient to account for this process transition. As local gradient approaches a critical slope $S_C$, which field studies have shown to vary between 0.8 [DiBiase et al., 2010; Hurst et al., 2012] and 1.25 [Roering et al., 1999], sediment flux asymptotically approaches infinity [Andrews and Bucknam, 1987; Anderson, 1994; Roering et al., 1999]:

$$q_s = -D \nabla z \left( 1 - \left( \frac{\nabla z}{S_C} \right)^2 \right)^{-1}, \tag{3}$$

where $D$ [L$^2$ T$^{-1}$] is a transport coefficient. Equation (3) has empirical and experimental support [Gab...
et al. 2001a; Pelletier and Cline, 2007]. We do not consider similar depth-dependent models [e.g., Heimsath et al., 2005] since soil depth does not vary systematically with erosion rate in the soil-mantled portions of our field site [Yoo et al., 2011], and we restrict our analyses to the soil-mantled areas of the field site.

[11] In Equation (3), the combined influences of climate and lithology on a suite of processes are lumped into a single parameter, $D$. These processes include freeze/thaw, wet/dry, and shrink/swell cycles [Gilbert, 1909]; bioturbation due to tree throw [e.g., Roering et al., 2010; Gabet and Mudd, 2010; Constantine et al., 2012] or burrowing organisms [Gabet, 2000; Yoo et al., 2005]; and rainsplash grain displacement [e.g., Dunne et al., 2010]. Hanks [2000] documented that $D$ increases systematically with climate from $D=0.1–0.7$ m$^2$ ka$^{-1}$ in the arid Middle East [Bowman and Gerson, 1986; Bowman and Gross, 1989; Begin 1992] through $0.5–2.0$ m$^2$ ka$^{-1}$ and $3.3–5.5$ m$^2$ ka$^{-1}$ in the semi-arid regions of the western U.S. [Hanks et al., 1984; Hanks and Wallace, 1985; Hanks and Andrews, 1989; Hanks, 2000] and western China [Tapponier et al., 1990; Avena et al., 1993; Avena and Peltzer, 1993], respectively, to $8.5–16$ m$^2$ ka$^{-1}$ in more humid coastal California [Hanks et al., 1984; Arrowsmith et al., 1998] and Michigan [Nash, 1980a]. Several studies have postulated increased hillslope sediment transport rates at the glacial-interglacial transition between the late Pleistocene and early Holocene in New Zealand, attributed to changes in vegetation density and type [Roering et al. 2004; Almond et al., 2008; Hughes et al., 2009; Walther et al., 2009]. Variation in the type and density of vegetation may also occur as soil conditions, and the availability of nutrients changes at lithologic boundaries.

2.2. Hillslope Morphology

[13] When hillslope gradient ($\nabla \zeta$) is small, the bracketed term in equation (3) approaches unity. Substituting equation (3) into equation (2), we can therefore solve for erosion rate where slope angles are low (i.e., on hilltops):

$$E = -\frac{\rho_s}{\rho} DC_{HT},$$

where $C_{HT}$ is the hillslope curvature, i.e., $\nabla^2 \zeta$, at the hilltop, since this is where we expect hilltop gradients to be the gentlest. Equation (4) predicts that the erosion rate on a steadily denuding hillslope should be linearly proportional to hilltop curvature $C_{HT}$ and the sediment transport coefficient $D$. We adopt the sign convention that convex up surfaces (i.e., hilltops) have negative curvature and erosion is a positive quantity (i.e., a positive value of $E$ indicates a lowering of the land surface). Equation (4) predicts erosion rates as a function of hillslope topography as long as $(\nabla^2 \zeta / SC)^2$ in equation (3) is small enough to be negligible. Equation (3) describes hillslope sediment transport; in valleys, fluvial transport and erosion dominate. Thus, equation (3) applies to the convex portions of the landscape [e.g., Roering et al., 2007]. The lowest gradients within the convex portions of the landscape, where equation (3) is most likely to occur, are on hilltops. Critically, however, this relationship only applies when the hilltop has attained topographic steady state; that is to say the rate of denudation should be the same everywhere on the hillslope, matching the rate in the channel at the base of the hillslope. During landscape adjustment to tectonics and/or baselevel change, hilltops are always the last part of the landscape to respond [e.g., Furbish and Fagherazzi, 2001; Mudd and Furbish, 2007].

[13] Hillslope relief has also been used to estimate erosion rates, but once erosion rates exceed $\sim 100–300$ mm ka$^{-1}$, further increases in erosion rates are accommodated by increased landsliding frequency on threshold slopes and hillslope gradients or hillslope relief become poor predictors of erosion rates [e.g., Burbank et al., 1996; Binnie et al., 2007; Quinmet et al., 2009; DiBiase et al., 2010; Larsen and Montgomery, 2012]. Equation (4) provides an alternative approach to estimating erosion rates from topography in landscapes with steep, planar hillslopes [Hurst et al., 2012]. Roering et al. [2007] provided a comprehensive framework for analyzing relationships between denudation and hillslope topography (i.e., relief, topographic slope, curvature, and hillslope length) when equation (3) is combined with the mass balance equation (equation (2))) in 1D form. Non-dimensionalization of erosion rate and relief allows comparisons between landscapes with distinct process rates and morphology. Roering et al. [2007] cast erosion rate and relief in non-dimensional form ($E^*$ and $R^*$, respectively) as functions of topographic parameters $C_{HT}$, $L_{HI}$, and mean hillslope gradient $S$:

$$E^* = \frac{E}{E_R} = \frac{\rho_s 2EL_H}{\rho S DC_{HT}},$$

$$E^* = \frac{2C_{HT}L_H}{SC},$$

$$R^* = \frac{1}{E^*} \left\{ \sqrt{1 + (E^*)^2} - \ln \left[ \frac{1}{2} \left(1 + \sqrt{1 + (E^*)^2}\right) \right] - 1 \right\}.$$

[14] Equation (6a) predicts a non-linear relationship between $E^*$ and $R^*$, which all hillslopes with a morphology that is adjusted to its boundary conditions should obey (provided that equation (3) gives a reasonable approximation of sediment transport processes on the hillslope). Similarly to equation (3), the prediction of equation (6a) only holds when the hillslope is denuding in concert with the adjacent channel. However, equations (5b) and (6b) allow us to calculate $E^*$ and $R^*$ from topographic attributes, even where the steady state assumption is violated, in order to compare to the model predictions for steady state encapsulated in equation (6a). In such a scenario, hillslope morphology is expected to vary from the model prediction in a manner that reflects the style of transience [Hurst et al., 2012]. We therefore developed techniques to quantify the spatial distribution of $C_{HT}$, $L_{HI}$, and $S$ within a landscape from LiDAR-derived topography in order to explore the spatial distribution of $E^*$ and $R^*$ and their relationship to bedrock type.

3. Methods

3.1. Quantifying Hillslope Morphology

[15] In forested landscapes, high resolution LiDAR DEMs commonly exhibit high local variability due to the presence of pits associated with the upheaval or decay of tree root clumps [e.g., Roering et al., 2010] or dense vegetation or
“brush” which has been misclassified as bare earth [Laschermes et al., 2007]. Thus, standard algorithms computing slope and curvature from $3 \times 3$ pixel moving windows produce noisy results. Assuming a diffusion-like model for sediment transport requires slope to be calculated at a larger scale than that at which the disturbance forces operate [Joytsna and Haff, 1997; Furbish et al. 2009]. Laschermes et al. [2007] found that a length scale (where length scale is twice the search radius) of 12 m was appropriate for LiDAR from the South Fork Eel River, CA, whilst Roering et al. [2010] demonstrated a length scale of 15 m in the forested landscape of the Oregon Coast Range. At our field site, the appropriate scaling was 12 m [Hurst et al., 2012]. Here we calculated the slope, curvature, and aspect from a 6-term quadratic surface fitted to a $12 \times 12$ m window in the gridded elevation data, centered on the pixel of interest [see Hurst et al., 2012].

[16] A hillslope can be considered to begin at a topographic divide and extend to a valley bottom, at which a transition from hillslope processes (i.e., diffusive processes and landslides) to valley-forming processes occurs (i.e., debris flow and/or fluval erosion). We extracted hilltops from the LiDAR DEM as the intersecting margins of zero-order and upward drainage basins, where slope ($\nabla z$) $< 0.4$. The valley network was defined using the Geonet algorithm of Passalacqua et al. [2010]. Hilltop curvature $C_{HT}$ was sampled at all pixels within 2 m of these hilltops. Adjacent hillslopes were sampled using an aspect-driven routing algorithm [Lea, 1992] to trace from each hilltop pixel to an adjacent valley bottom. Along the resulting profile, the mean slope ($S$), relief ($R$), and horizontal hillslope length ($L_H$) were recorded. A mean value for each of these metrics was then determined for each hilltop segment [see Hurst et al., 2012; for detailed description of methods].

3.2. Quantifying Vegetation Properties

[17] Sediment transport in forested, soil-mantled landscapes is driven, in part, by tree turnover through growth and decay of tree roots and the upheaval of tree root wads [e.g., Schaetzl and Folling, 1990; Gabet et al., 2003; Gabet and Mudd, 2010]. Therefore, it has been suggested that the sediment transport coefficient ($D$ in equation (3)) may vary with aboveground biomass (AGB) [Roering et al., 2004; Walther et al. 2009]. Airborne-derived LiDAR data collected from forested landscapes contain a wealth of information about vegetation, with last/lowest returns being generally classified as the ground surface and all aboveground returns being reflected from vegetation surfaces (leaves, branches, etc.). Properties of the canopy elevation structure such as the mean and standard deviation of the height of canopy returns ($V_{mean}$ and $V_{sd}$, respectively) can be readily extracted from LiDAR [e.g., Nilsson, 1996; Naesset, 1997; Lefsky et al., 2002; Holmgren et al., 2003; Donoghue and Watt, 2006] and may provide useful indicators of AGB [e.g., Hall et al., 2005; Clark et al. 2011; Pelletier et al., 2011; Saatchi et al., 2011]. Clark et al. [2011] found that LiDAR-derived metrics $V_{mean}$ and $V_{max}$ provided robust indicators of measured AGB. Pelletier et al. [2011] used a 1 m resolution canopy height map derived from LiDAR to demonstrate a negative relationship between mean $C_{HT}$ and $V_{mean}$ suggesting that vegetation cover may be controlling the sediment transport coefficient $D$.

[18] Here we investigated how vegetation cover varies on hillslopes as a function of lithology between the two study areas to assess whether variation in quantifiable vegetation metrics derived from LiDAR could account for differences in the efficacy of sediment transport. The approach is limited by the assumption that modern vegetation accounts for the current shape of hillslopes, yet others have attributed change in biologically driven sediment transport to change in vegetation cover at the end of the last glaciation [e.g., Roering et al., 2004; Walther et al., 2009].

[19] We analyzed point cloud density and height above the ground surface for returns classified as vegetation. Return classification was carried out by the National Center for Airborne Laser Mapping. To estimate canopy height, the heights of aboveground point returns were detrended by subtracting the elevation of the ground surface interpolated to a 1 m grid. The resulting canopy heights were analyzed to compute values for $V_{mean}$ and $V_{sd}$ in each grid cell in 4 m resolution grids (coarsened to avoid data gaps where there were no aboveground LiDAR returns). Although the low-relief portions of the landscape have been heavily logged, there is no evidence of recent logging (i.e., no cut stumps and numerous trees with diameters exceeding 1 m). Canopy height data from the LiDAR appear to qualitatively agree with satellite imagery (i.e., bare patches apparent on satellite images correspond to absent or minimal canopies from LiDAR). We defined a vegetation density ratio $V_{dens}$ as the ratio between the points classified as aboveground and points classified as ground, normalized to the total number of returns within a 4 m resolution grid. A ratio $V_{dens}$ $= 1$ indicates that all points returned were aboveground and the canopy is dense, whilst a ratio $V_{dens}$ $\to 0$ indicates little/no vegetation cover. We consider this a crude approach since the results are limited by the average point spacing of LiDAR returns ($\sim 4$ m$^{-2}$) and may be influenced by any variation in leaf structure and tree spacing.

4. Study Sites

[20] We explored hillslope morphology in the lower reaches of the Middle Fork Feather River, in the northern Sierra Nevada of California (Figure 2). Our study is focused on an area where granitoid plutons are intruded into the Central belt terrain which consists of Upper Triassic-Jurassic ophiolitic, volcanic, and sedimentary units of the Fiddle Creek Complex [Day and Bickford, 2004] (Figure 3).

[21] The landscape comprises a low-relief, relit surface characterized by concave up channel profiles and broad “diffusive” hillslopes which is likely adjusted to some previous erosional regime. This landscape is dissected by the canyons of the Middle Fork Feather River and its tributaries (Figure 2). Canyon incision was initiated by accelerated uplift c. 3.5–5 Ma [Wakabayashi and Sawyer, 2001; Stock et al., 2004], possibly caused by the delamination of an eclogite root beneath the mountain range [Saleeby and Foster, 2004; Jones et al., 2004]. Apatite fission track dates reveal an average erosion rate of 40 mm ka$^{-1}$ for the relict landscape, persisting until at least 32 million years ago (Ma) [Cecil et al., 2006]. Long-term exhumation rates derived from (U-Th)/He ages fail to record a late-Cenozoic acceleration in denudation, implying that less than 3 km of the crust has been exhumed since the acceleration
Canyon erosion rates of ~200 mm ka\(^{-1}\) south of the Feather River, et al. [2012] estimated minimum incision rate of 170 mm ka\(^{-1}\) for the Feather River having an estimated minimum incision rate of late-Cenozoic volcanics capping ridges/divides with an estimated rate of 170 mm ka\(^{-1}\) in the Feather River basin [Wakabayashi and Sawyer, 2001]. Erosion rates measured within the Feather River basin vary by over an order of magnitude from the canyons to the adjacent relict upland, with the black box showing the location of Figure 3. The spatial reference system is UTM Zone 10N with spatial units in meters.

[Cecil et al., 2006] Incision rates for the Feather River canyon have been reconstructed for the last 5 Ma from the presence of late-Cenozoic volcanics capping ridges/divides with the Feather River having an estimated minimum incision rate of 170 mm ka\(^{-1}\) over the last 5 Ma [Wakabayashi and Sawyer, 2001]. Erosion rates measured within the Feather River basin vary by over an order of magnitude from the relict surface (~20 mm ka\(^{-1}\)) to the canyons (~250 mm ka\(^{-1}\)) [Riebe et al., 2000; Wakabayashi and Sawyer, 2001; Hurst et al., 2012] (Figure 2b). For several catchments >200 km south of the Feather River, Stock et al. [2004] established modern mountainous areas with ~200 mm ka\(^{-1}\) between 1.5 and 2.7 Ma, compared with ~30 mm ka\(^{-1}\) since 1.5 Ma, from CRN dating of cave sediments now suspended above the modern climate is semi-arid with a strong precipitation gradient from the dry Central Valley of California to the high elevations of the Sierra Nevada mountains. At our study site, mean annual temperature is 12.5 °C and mean annual precipitation is 1750 mm (data from the PRISM Climate Group, Oregon State University, http://prism.oregonstate.edu (accessed 7 July 2011) [Daly et al., 1997]). The Feather River basin remained largely unglaciated during the Pleistocene, except for its uppermost reaches [Wahrhaftig and Birman, 1965; Clark, 1995].

[Hurst et al., 2012] extended the dataset used by Riebe et al. [2000] and demonstrated that as basin-averaged erosion rates increase, hilltops get sharper (i.e., curvature becomes more negative) in granitoid portions of the landscape (Figure 2c). A linear relationship provides the best fit (R\(^2\) = 0.83), as predicted by equation (4), allowing the sediment transport coefficient to be constrained at 8.6 m\(^2\) ka\(^{-1}\) [see Hurst et al., 2012]. The black box shows the location of Figure 3. The steepest parts of the landscape. This is consistent with a sediment transport law in which flux increases to infinity as slopes approach some limiting angle (approximating the effect of increased landslide frequency), as in equation (3), as previously demonstrated in other landscapes [e.g., Binnie et al., 2007; Ouimet et al., 2009; DiBiase et al., 2010].

We focused our analysis on the north end of the Sierra Nevada mountains. At our study site, mean annual temperature is 12.5 °C and mean annual precipitation is 1750 mm (data from the PRISM Climate Group, Oregon State University, http://prism.oregonstate.edu (accessed 7 July 2011) [Daly et al., 1997]). The Feather River basin remained largely unglaciated during the Pleistocene, except for its uppermost reaches [Wahrhaftig and Birman, 1965; Clark, 1995].

[Hurst et al., 2012] extended the dataset used by Riebe et al. [2000] and demonstrated that as basin-averaged erosion rates increase, hilltops get sharper (i.e., curvature becomes more negative) in granitoid portions of the landscape (Figure 2c). A linear relationship provides the best fit (R\(^2\) = 0.83), as predicted by equation (4), allowing the sediment transport coefficient to be constrained at 8.6 m\(^2\) ka\(^{-1}\) [see Hurst et al., 2012]. The black box shows the location of Figure 3. The steepest parts of the landscape. This is consistent with a sediment transport law in which flux increases to infinity as slopes approach some limiting angle (approximating the effect of increased landslide frequency), as in equation (3), as previously demonstrated in other landscapes [e.g., Binnie et al., 2007; Ouimet et al., 2009; DiBiase et al., 2010].

We focused our analysis on the north end of the Sierra Nevada mountains. At our study site, mean annual temperature is 12.5 °C and mean annual precipitation is 1750 mm (data from the PRISM Climate Group, Oregon State University, http://prism.oregonstate.edu (accessed 7 July 2011) [Daly et al., 1997]). The Feather River basin remained largely unglaciated during the Pleistocene, except for its uppermost reaches [Wahrhaftig and Birman, 1965; Clark, 1995].

[Hurst et al., 2012] extended the dataset used by Riebe et al. [2000] and demonstrated that as basin-averaged erosion rates increase, hilltops get sharper (i.e., curvature becomes more negative) in granitoid portions of the landscape (Figure 2c). A linear relationship provides the best fit (R\(^2\) = 0.83), as predicted by equation (4), allowing the sediment transport coefficient to be constrained at 8.6 m\(^2\) ka\(^{-1}\) [see Hurst et al., 2012]. The black box shows the location of Figure 3. The steepest parts of the landscape. This is consistent with a sediment transport law in which flux increases to infinity as slopes approach some limiting angle (approximating the effect of increased landslide frequency), as in equation (3), as previously demonstrated in other landscapes [e.g., Binnie et al., 2007; Ouimet et al., 2009; DiBiase et al., 2010].

We focused our analysis on the north end of the Sierra Nevada mountains. At our study site, mean annual temperature is 12.5 °C and mean annual precipitation is 1750 mm (data from the PRISM Climate Group, Oregon State University, http://prism.oregonstate.edu (accessed 7 July 2011) [Daly et al., 1997]). The Feather River basin remained largely unglaciated during the Pleistocene, except for its uppermost reaches [Wahrhaftig and Birman, 1965; Clark, 1995].

[Hurst et al., 2012] extended the dataset used by Riebe et al. [2000] and demonstrated that as basin-averaged erosion rates increase, hilltops get sharper (i.e., curvature becomes more negative) in granitoid portions of the landscape (Figure 2c). A linear relationship provides the best fit (R\(^2\) = 0.83), as predicted by equation (4), allowing the sediment transport coefficient to be constrained at 8.6 m\(^2\) ka\(^{-1}\) [see Hurst et al., 2012]. The black box shows the location of Figure 3. The steepest parts of the landscape. This is consistent with a sediment transport law in which flux increases to infinity as slopes approach some limiting angle (approximating the effect of increased landslide frequency), as in equation (3), as previously demonstrated in other landscapes [e.g., Binnie et al., 2007; Ouimet et al., 2009; DiBiase et al., 2010].
complex. With the Feather River downcutting rapidly, the Little North Fork and Cascade rivers are undergoing a transient adjustment to acceleration in baselevel lowering. We studied hilltops and adjacent hillslopes near the confluences where they were most likely to be adjusting/adjusted to baselevel lowering, since adjacent valleys are downstream of major knickpoints (location of a sudden increase in slope downstream; Figure 4). The studied hillslopes, separated by less than 5 km and spanning similar ranges in elevation, can be assumed to have similar climatic histories, and their proximity to the Feather River implies similar denudation history. Based on field observations, a significant driver of sediment transport in this forested landscape is growth/decay of tree roots and the upheaval of root wads and associated soil by tree throw. Thick soils are developed on steep slopes in both areas (Figure 5). The two areas lie in the Plumas National Forest, and vegetation consists of the California mixed conifer forest type, which includes Douglas fir, incense-cedar, and sugar pine [Warbington and Beardsley, 2002].

5. Results

5.1. Morphology of Hillslopes as a Function of Lithology

[25] Figure 6 shows the spatial distribution of hilltops sampled and their hilltop curvature. The highest values of hilltop curvature (i.e., the most convex or sharpest hilltops) occur on hilltops most proximal to the Feather River, the Cascade River, and the Little North Fork River, downstream of the main knickpoints (Figure 4). High values of hilltop curvature are spatially more distributed in the northwest of the study area since the knickpoint has propagated further up the Little North Fork tributary than along Cascade River. In Figure 7, the relationship between hilltop curvature and hillslope gradient is compared for the granodiorite and metavolcanics. We find a non-linear relationship between mean hilltop curvature and mean hillslope gradient. Where there is low hilltop curvature, hillslope gradients are also low. As hilltop curvature increases, so too does hillslope gradient; however, beyond $C_{HT} \approx -0.03 \text{ m}^{-1}$ (i.e., where $C_{HT}$ is more negative), hillslope gradients do not continue to increase as rapidly. This relationship occurs in both lithologies, but the two datasets are offset such that in the metavolcanics, for low values of hilltop curvature (greater than $-0.03 \text{ m}^{-1}$), hillslopes tend to be steeper, and hillslopes in the metavolcanics approach their limiting gradient at lower hilltop curvatures than in the granodiorite. Since hillslope gradients appear to be limited, they will not reflect the erosion rates driving their evolution; however, following equation (4), hilltop curvature may better reflect the distribution of erosion rates if the hillslope has fully responded to the change in boundary conditions [Hurst et al., 2012].

[26] We cast these results in non-dimensional form to compare them to the expected form of model hillslopes governed by equation (3), calculating dimensionless erosion rate as a function of hilltop curvature and hillslope length, and dimensionless relief as a function of mean slope (equations (5b) and (6b)) [Roering et al., 2007]. Despite considerable scatter,
we found that the distribution of binned $R^*$ and $E^*$ is of a similar form to that predicted by equation (6a), as depicted by the dashed line in Figure 8. Note from equations (5b) and (6b) that $S_C$ is required to quantify both $E^*$ and $R^*$ based on measurable topographic properties, and hence, the value of $S_C$ used can alter the position of the data relative to the dashed steady state line (see section 5.2). $E^*$ was calculated as a function of hilltop curvature and hillslope length (equation (5b)). Despite non-dimensionalization, there is still a tendency for hillslopes in the metavolcanics to be steeper when $E^*$ is low. Frequency distributions of hillslope lengths (Figure 9a) reveal that hillslopes are slightly longer in the metavolcanics (peak at 175–200 m, and a larger proportion of long hillslopes) than in the granodiorite (peak at 150–175 m, with a larger number of short hillslopes). We carried out a $t$-test to test the equality of the two sample means and concluded at 99% confidence that the samples were drawn from different populations. This is also shown by plotting $C_{HT}$ (controlled by $E$) versus $L_H$.

Figure 4. Longitudinal profiles for Little North Fork (blue) and Cascade (red) rivers relative to the lower reaches of the Feather River (black) which sets baselevel. Profiles were generated from U.S. Geological Survey National Elevation Dataset 1/3 arc-second (approx. 10 m) DEMs [http://seamless.usgs.gov/; accessed 15/1/2009]. Shaded region indicates area in which the landscape is interpreted to be adjusting or adjusted to the rapid incision rate of the Feather River along both the Little North Fork and Cascade Rivers, downstream of major knickpoints marked with filled circles. Note these knickpoints both fall out with the bounding region of Figure 3. The lower reaches of Cascade and Little North Fork rivers cross the Cascade Pluton (granodiorite) and Central Belt (metavolcanics), respectively.

Figure 5. Example soil pits in (left) granodiorite and (right) metavolcanics. Pit in granodiorite comes from hilltop along cascade ridge (red line in Figure 4) with hilltop curvature $C_{HT} = -0.062$; pit depth is 65 cm. Pit in metavolcanics on steep slope (c. 40°) on the western flank of the Little North Fork River; pit depth is 85 cm.
C. The MLE was calculated as follows, reporting error

\[ MLE = \prod_{i=1}^{n} \exp \left( \frac{(R^*_\text{meas} - R^*_\text{mod})^2}{2\sigma_p^2} \right) \]  

where \( n \) is the number of hillslopes sampled, the subscripts meas and mod refer to measured and modeled values, respectively, and \( \sigma_p \) is the variance in measured \( R^* \) values, which will alter the magnitude of MLE calculated but will not change the most likely value of \( S_C \). The MLE for \( S_C \) was 0.79 \(-0.07/ +0.38 \) for the granodiorite and 0.85 \(-0.08/+0.53 \) for the metavolcanics. This indicates that the maximum attainable gradient on hillslopes in the metavolcanics may be slightly higher. We interpret the large range in error values as due to a significant proportion of hillslope data having low \( E^* \) (<10) and \( R^* \) (<0.8) (Figure 8), at which the model predictions are insensitive to changes in \( S_C \). Having more data points at high \( E^* \) would significantly reduce the error range since it is at high erosion rates that hillslopes become steep and planar and are most likely to reflect \( S_C \). The likelihood that \( S_C \) is 0.85 and 0.79 in the granodiorite and metavolcanics (i.e., that our result is reversed) is over a factor of two less likely. As our result is reversed) is over a factor of two less likely. As

5.2. Constraining \( S_C \)

[27] Critical slope \( S_C \) is used when calculating both \( E^* \) and \( R^* \) from topographic metrics. We estimated \( S_C \) by finding the maximum likelihood estimator (MLE) between the hillslope data and model predictions for \( R^* \) as a function of \( E^* \) (equation (6a)) with varying \( S_C \). It is important to highlight here that the model fitted applies to steady state hillslope morphology, yet the landscape analyzed spans a range of erosion rates (Figure 2) and there are knickpoints in the channel system (Figure 4). Thus, it is likely that some hillslopes may have transient morphology. Nevertheless, the tendency of the asymptote created by equation (6a) (see dashed line in Figure 8) is controlled by \( S_C \), and the highest values of \( R^* \) contained in the datasets should reflect \( S_C \). The MLE was calculated as follows, reporting error range at one standard deviation of the normalized probability distribution:

\[ MLE = \prod_{i=1}^{n} \exp \left( \frac{(R^*_\text{meas} - R^*_\text{mod})^2}{2\sigma_p^2} \right) \]  

\[ \text{MLE} = \prod_{i=1}^{n} \exp \left( \frac{(R^*_\text{meas} - R^*_\text{mod})^2}{2\sigma_p^2} \right) \]  

\[ \text{MLE} = \prod_{i=1}^{n} \exp \left( \frac{(R^*_\text{meas} - R^*_\text{mod})^2}{2\sigma_p^2} \right) \]  

where \( n \) is the number of hillslopes sampled, the subscripts meas and mod refer to measured and modeled values, respectively, and \( \sigma_p \) is the variance in measured \( R^* \) values, which will alter the magnitude of MLE calculated but will not change the most likely value of \( S_C \). The MLE for \( S_C \) was 0.79 \(-0.07/ +0.38 \) for the granodiorite and 0.85 \(-0.08/+0.53 \) for the metavolcanics. This indicates that the maximum attainable gradient on hillslopes in the metavolcanics may be slightly higher. We interpret the large range in error values as due to a significant proportion of hillslope data having low \( E^* \) (<10) and \( R^* \) (<0.8) (Figure 8), at which the model predictions are insensitive to changes in \( S_C \). Having more data points at high \( E^* \) would significantly reduce the error range since it is at high erosion rates that hillslopes become steep and planar and are most likely to reflect \( S_C \). The likelihood that \( S_C \) is 0.85 and 0.79 in the granodiorite and metavolcanics (i.e., that our result is reversed) is over a factor of two less likely. As our result is reversed) is over a factor of two less likely. As

5.2. Constraining \( S_C \)

[27] Critical slope \( S_C \) is used when calculating both \( E^* \) and \( R^* \) from topographic metrics. We estimated \( S_C \) by finding the maximum likelihood estimator (MLE) between the hillslope data and model predictions for \( R^* \) as a function of \( E^* \) (equation (6a)) with varying \( S_C \). It is important to highlight here that the model fitted applies to steady state hillslope morphology, yet the landscape analyzed spans a range of erosion rates (Figure 2) and there are knickpoints in the channel system (Figure 4). Thus, it is likely that some hillslopes may have transient morphology. Nevertheless, the tendency of the asymptote created by equation (6a) (see dashed line in Figure 8) is controlled by \( S_C \), and the highest values of \( R^* \) contained in the datasets should reflect \( S_C \). The MLE was calculated as follows, reporting error range at one standard deviation of the normalized probability distribution:

\[ MLE = \prod_{i=1}^{n} \exp \left( \frac{(R^*_\text{meas} - R^*_\text{mod})^2}{2\sigma_p^2} \right) \]  

where \( n \) is the number of hillslopes sampled, the subscripts meas and mod refer to measured and modeled values, respectively, and \( \sigma_p \) is the variance in measured \( R^* \) values, which will alter the magnitude of MLE calculated but will not change the most likely value of \( S_C \). The MLE for \( S_C \) was 0.79 \(-0.07/ +0.38 \) for the granodiorite and 0.85 \(-0.08/+0.53 \) for the metavolcanics. This indicates that the maximum attainable gradient on hillslopes in the metavolcanics may be slightly higher. We interpret the large range in error values as due to a significant proportion of hillslope data having low \( E^* \) (<10) and \( R^* \) (<0.8) (Figure 8), at which the model predictions are insensitive to changes in \( S_C \). Having more data points at high \( E^* \) would significantly reduce the error range since it is at high erosion rates that hillslopes become steep and planar and are most likely to reflect \( S_C \). The likelihood that \( S_C \) is 0.85 and 0.79 in the granodiorite and metavolcanics (i.e., that our result is reversed) is over a factor of two less likely. As our result is reversed) is over a factor of two less likely. As

5.2. Constraining \( S_C \)

[27] Critical slope \( S_C \) is used when calculating both \( E^* \) and \( R^* \) from topographic metrics. We estimated \( S_C \) by finding the maximum likelihood estimator (MLE) between the hillslope data and model predictions for \( R^* \) as a function of \( E^* \) (equation (6a)) with varying \( S_C \). It is important to highlight here that the model fitted applies to steady state hillslope morphology, yet the landscape analyzed spans a range of erosion rates (Figure 2) and there are knickpoints in the channel system (Figure 4). Thus, it is likely that some hillslopes may have transient morphology. Nevertheless, the tendency of the asymptote created by equation (6a) (see dashed line in Figure 8) is controlled by \( S_C \), and the highest values of \( R^* \) contained in the datasets should reflect \( S_C \). The MLE was calculated as follows, reporting error range at one standard deviation of the normalized probability distribution:

\[ MLE = \prod_{i=1}^{n} \exp \left( \frac{(R^*_\text{meas} - R^*_\text{mod})^2}{2\sigma_p^2} \right) \]  

where \( n \) is the number of hillslopes sampled, the subscripts meas and mod refer to measured and modeled values, respectively, and \( \sigma_p \) is the variance in measured \( R^* \) values, which will alter the magnitude of MLE calculated but will not change the most likely value of \( S_C \). The MLE for \( S_C \) was 0.79 \(-0.07/ +0.38 \) for the granodiorite and 0.85 \(-0.08/+0.53 \) for the metavolcanics. This indicates that the maximum attainable gradient on hillslopes in the metavolcanics may be slightly higher. We interpret the large range in error values as due to a significant proportion of hillslope data having low \( E^* \) (<10) and \( R^* \) (<0.8) (Figure 8), at which the model predictions are insensitive to changes in \( S_C \). Having more data points at high \( E^* \) would significantly reduce the error range since it is at high erosion rates that hillslopes become steep and planar and are most likely to reflect \( S_C \). The likelihood that \( S_C \) is 0.85 and 0.79 in the granodiorite and metavolcanics (i.e., that our result is reversed) is over a factor of two less likely. As our result is reversed) is over a factor of two less likely. As
The two lithologies have similar mean (0.84 and 0.85) and median (0.83 and 0.86) slope values for granodiorite and metavolcanics, respectively.

5.3. Estimates of Aboveground Biomass

In the field, we observed mixed A soil horizons of fairly uniform depth [Yoo et al., 2011] and no evidence of overland flow or ravelling processes, even during the 2009 field season when we visited the site after a fire. The entire area studied is forested, and we observed a number of uprooted trees and associated surface pits. These field observations suggest that slope-dependent sediment transport on hilltops is dominated by vegetation turnover in the Feather River region, although we cannot rule out rheologic creep as a contributing mechanism [e.g., McKean et al., 1993].

Vegetation properties were compared along two prominent ridges in the granodiorite and the metavolcanics to determine whether the differing distributions of hilltop curvature could be explained by vegetation controlling the sediment transport coefficient (Figure 11). These ridgelines were selected as the only hilltops bound on both sides by the main tributary channels. On these ridges, hilltops are sharp (more negative $C_{HT}$ values indicate sharper hilltops and imply more rapid erosion): mean $C_{HT}$ for granodiorite is $-0.067$ m$^{-1}$ and $-0.12$ m$^{-1}$ for metavolcanics (Figure 12), indicating that these sites have likely responded to baselevel lowering (though they may still be adjusting). We find that vegetation on the two ridges has remarkably similar density ratios $V_{dens}$ yet exhibit differences in canopy height (Figure 11). The mean height values from profiles along the length of each ridgeline are similar on both ridges ($V_{mean} = 7.9 \pm 4.9$ m and $8.0 \pm 3.5$ m for ridges in the granodiorite and metavolcanics, respectively).

6. Discussion

6.1. Calibrating the Sediment Transport Coefficient

To compare estimates of the sediment transport coefficient $D$ between lithologies, we sampled hilltop curvature on all ridges within 500 m of reaches of the Feather River, Cascade River, or Little North Fork River that are downstream of knickpoints (Figure 4) where the long-term erosion rate is estimated to be c. 250 mm ka$^{-1}$ [Riebe et al., 2000; Wakabayashi and Sawyer 2001; Hurst et al., 2012]. The results were binned to produce histograms of hilltop canyons. Figure 10 shows slope histograms sampled for both granodiorite and metavolcanics portions of the landscape which are downstream of convexities in the channel profile. The two lithologies have similar mean (0.84 and 0.85) and median (0.83 and 0.86) slope values for granodiorite and metavolcanics, respectively.

$C_{HT}$ is expected to be a good indicator of relative erosion rate, whilst at high erosion rates, $S$ becomes insensitive to baselevel fall. Note we plot bin-mean averaged $C_{HT}$ and $S$. In both lithologies, $C_{HT}$ continues to vary despite $S$ becoming limited. For low $C_{HT}$, hillslopes are steeper in the metavolcanics (at 99% confidence level). $S$ in both lithologies appears to be limited to ~0.85.
The result from the granodiorite is similar to the value of predicting a lower diffusivity of curvature tends to be higher (more negative) (Figure 12), $E = 250 \text{mm ka}^{-1}$. Heimsath et al. [e.g., has been demonstrated for some other granitic lithologies in this.]

6.2. Applicability of Sediment Transport Models

Increased incision. The sediment transport rates reported here are dependent on the assumption that erosion rates are the same in parts field site based on data in Figure 2. The sediment transport rates reported here are dependent on the assumption that erosion rates are the same in parts of the landscape that are most likely to be adjusted to in-

Hillslopes in the metavolcanics tend to be steeper for a given erosion rate despite correcting for hillslope length, suggesting a greater number of hillslopes undergoing transient adjustment were sampled in the metavolcanics (see text for further discussion).

6.2. Applicability of Sediment Transport Models

[32] Much of the topographic analysis above has been carried out assuming that hillslope sediment transport is well approximated as a non-linear function of local slope (equation (3)). This model is assumed to be applicable to the Feather River since hilltop curvature varies linearly with erosion rate [Hurst et al., 2012]. Models similar to equation (3) in which sediment flux is also a product of soil depth (i.e., $D = D_L \times h$, where $D_L \text{[L T}^{-1}]$ is the transport coefficient for depth-dependent transport and $h \text{[L]}$ is soil depth) predict that hilltop curvature will vary non-linearly with erosion rate, becoming extremely sensitive to changes in $E$ when $E$ is high [Roering, 2008]. Field measurements of soil depth were invariant on hillslopes above, at, and just below a prominent break in slope that separates the relict landscape from the steep topography in an area of tonalite ~10 km to the south of the study area [Yoo et al., 2011]. This implies that soil thickness is set primarily by the depth of root action in this forested land-

Figure 8. Non-dimensional erosion rate ($E^*$) and relief ($R^*$) calculated from topographic metrics following equations (5b) and (6b) for hilltops in the granodiorite (red) and metavolcanics (blue), using maximum likelihood estimated values for $S_C$ of 0.79 and 0.85, respectively. Data are binned into regularly spaced bins in $E^*$. Black dashed line shows theoretical relationship predicted by equation (6a) for steadily eroding hillslopes. Non-dimensional analysis effectively normalizes the data from Figure 7 for variation in hillslope length. Hillslopes in the metavolcanics tend to be steeper for a given erosion rate despite correcting for hillslope length, suggesting a greater number of hillslopes undergoing transient adjustment were sampled in the metavolcanics (see text for further discussion).
increasing hilltop curvature (an indicator of erosion rates).

0.66, respectively. Hillslope lengths tend to shorten with however stress that 

orite (red) and metavolcanics (blue). Black lines represent 

the maximum frequency. Maximum frequency occurs at 

granodiorite (red) and metavolcanics, normalized by 

hillslopes in the granodiorite and more long hillslopes in the 

metavolcanics. Additionally, there are more short 

hillslopes proximal to the main stem channels. Equation (3) 

applied only to soil-covered hillslopes, whereas emergent 

bedrock will be capable of maintaining steeper slopes, limited 

by the mechanical strength of the rock face.

6.3. Transient Landscape Response

[34] In Figure 4, it can be observed that the transient erosion rate signal, represented in this case by a distinct convexity in the channel profile, has migrated further along the Little North Fork River than the Cascade River. All else being equal, we would expect knickpoints to propagate faster into weaker/less resistant lithologies [Whipple and Tucker, 1999], suggesting that the metavolcanics may be less resistant to fluvial erosion than the granodiorite. However, the Little North Fork is a slightly larger basin (120 km² compared to Cascade River 85 km²). The tendency for hillslopes in the metavolcanics to have steep slopes at low hilltop curvature results in them plotting above the steady state line in Figure 8, and therefore, they may still be responding to accelerated incision [Hurst et al., 2012]. Hillslope morphology in the granodiorite conforms better to the steady state predictions (Figure 8). Hillslopes in the granodiorite may be able to keep pace with channel incision due to having a higher sediment transport coefficient [Roering et al., 2001b]. Additionally, since the knickpoint has not migrated as far into the granodiorite, it is likely propagating slower than in the metavolcanics. Gallen et al. [2011] demonstrated that the passing of a knickpoint results in hillslope steepening and an increase in hillslope relief immediately downstream; these results mirrored the theoretical predictions of Mudd and Furbish [2007]. However, with increasing distance downstream from the knickpoint, Gallen et al. [2011] found that these metrics begin to reduce again, suggesting that hillslopes are relaxing following the passing of a knickpoint and initial hillslope steepening. In the Feather River, we have been unable to observe such relaxation on hillslopes and are as yet unable to assert whether the increased erosion rates that have carved the Feather River canyon are a persistent response to a change in tectonic forcing or alternatively reflect a baselevel adjustment similar to that observed by Gallen et al., [2011]. Such a problem has important bearing on the calibrated values of \( D \), since we have assumed in section 6.1 that erosion rates are the same, and persistently
high, in order to solve equation (4) to derive \( D \). Analysis of cosmogenic radionuclides in cave sediments elsewhere in the Sierra Nevada suggests that the erosion history in the late Cenozoic is characterized by a pulse of incision moving through the landscape [Stock et al. 2004]. Incision is inferred to be a response to accelerated uplift in the late Cenozoic [Wakabayashi and Sawyer, 2001]. The likely mechanism of uplift is an isostatic response to delamination of an eclogite root beneath the mountain range [Saleeby and Foster, 2004; Jones et al., 2004]. Therefore, it seems likely that uplift rates will decrease through time as new isostatic equilibrium is approached.

6.4. Hillslope Lengths and Drainage Density

Hillslope lengths tend to be longer in the metavolcanics than in the granodiorite (Figure 9). A recent study by Perron et al. [2008] postulated that drainage density and its inverse, hillslope length, should be set by the relative efficiency of diffusive (hillslope) and advective (valley-forming) processes. Their analysis focused on low-relief settings, where hillslope processes could be assumed to be diffusive and sediment flux linearly related to slope, whilst valley-forming processes were dominated by channelization of overland flow. However, the present study was focused on a landscape responding to an order of magnitude increase in erosion rates, where zero-order basins may be predominantly eroded by debris flows and hillslopes approach a threshold gradient as a process transition to landslide-dominated sediment transport occurs. Perron et al. [2009] demonstrated that such a relationship breaks down in rapidly denuding landscapes with steep planar hillslopes such as the Oregon Coastal Range.

Hillslope length \( L_{H} \) in part controls slope steepness through setting the proportion of a hillslope that is planar and experiencing non-linearity in sediment transport [Roering et al., 2001b] (i.e., the longer the hillslope, the longer the proportion of its length that will be steep and planar). In the Feather River region, hillslope length decreases with erosion rate (assuming \( C_{HT} \) is a surrogate) (Figure 9). Mudd and Furbish [2005] demonstrated that hilltops may migrate when subject to differential erosion rate on either flank such that when erosion rate is raised on one side, the hillslope on that side increases its length. However, in their model, the extent of the drainage network was fixed, and an increase in hillslope length by divide migration was accommodated by shortening of the adjacent, low erosion rate hillslope. Contrary to the results presented here, drainage density (inverse of \( L_{H} \)) has been demonstrated to vary negatively with relief in steep, mountainous landscapes [Montgomery and Dietrich, 1988; Oguchi, 1997]. Such a result has been supported by analytical and numerical modeling studies which predict that where hillslope sediment transport is
dominated by landsliding, there should be an inverse relationship between drainage density and topographic relief [Howard, 1997; Tucker and Bras, 1998]. These studies have focused on landscapes (real or otherwise) that are assumed to be adjusted to their boundary conditions. The Feather River is still responding to a transient erosion signal, and as such, the patterns observed here may only be associated with processes of landscape response. In landscapes experiencing rapid erosion, where coupled landslide and debris-flow processes are likely to be the dominant erosion processes on hillslopes and in valleys respectively, drainage density is likely to be influenced by factors governing the frequency and magnitude of landslide events and the potential for these events to erode valleys as they translate into debris flows and scour the substrate [e.g., Stock and Dietrich, 2006]. In such settings, it may therefore be difficult to isolate the relative efficiencies of hillslope- and valley-forming processes. We find that hillslopes tend to get shorter with increased erosion rate in the Feather River and speculate that this may be due to an increase in debris flow frequency, allowing the valley-forming process to be more efficient at high erosion rates, so that valley heads migrate further into the landscape.

6.5. Mechanisms for Lithologic Control on Hillslope Sediment Transport

[37] Several workers have suggested that lithology may be important in setting $D$ [e.g., McKeen et al., 1993; Yoo et al., 2005], but as yet, we have limited quantitative understanding of such a relationship [c.f. Furbish et al., 2009]. This is in part due to an inability to isolate lithologic control from
that of climate, vegetation, and bioturbation. Here we have demonstrated that $D$ varies by a factor of two between two landscapes underpinned by different lithologies, despite similarity in vegetation, and presumably climate, given their spatial proximity and similar range in altitude. The sediment transport coefficient is not directly controlled by lithology, rather by the characteristics of the soil produced and the processes that act to transport sediment, which are in turn influenced by lithology. We have demonstrated that LiDAR metrics for AGB are similar between the two lithologies on hillslopes; therefore, energy expended in root growth/decay and tree throw should be similar, yet the amount of soil moved per transport event must differ. Future research should attempt to quantify mechanical properties of bedrock and chemical weathering in settings where $D$ can be determined, and a number of possible mechanisms by which lithology may influence sediment transport can be anticipated.

[38] The chemical and physical properties of soils are set by lithology, which influence the efficacy of sediment transport. Disparity in the degree of chemical and physical weathering in soil and saprolite may result in different volumes of material mobilized by tree throw and root growth. Furbish et al. [2009] derived a diffusion-like equation through which the disturbing and settling motions of individual particles within a soil. They parameterized $D$ as controlled by active soil depth, characteristic particle size, porosity (partially set by particle size), and a rate of particle activation (frequency of disturbance per unit time). The grain size distribution in a soil may then exert control on sediment transport efficiency by adjusting the mean free path length a grain can travel when disturbed or settling. Coarser grain size distributions have larger pore spaces and therefore facilitate longer travel distances per disturbance event. Rapid erosion rates lead to shorter residence times of soil material, and therefore, less time is available for the production of fine-grained weathering products such as clays and pedogenic crystalline iron [Mudd and Yoo, 2010; Yoo et al., 2011], and accelerated erosion rates are expected to result in a greater proportion of rock fragments in soils [Marshall and Sklar, 2011]. Furthermore, variation in grain size distributions may influence the hydrology [e.g., Poesen and Lavee, 1994], potentially impacting upon cyclical wetting/drying expansion/contraction within the soil. Because grain size distributions in soils are likely to be positively correlated with erosion rates, we also expect that $D$ may increase with erosion rates. Therefore, the common assumption that $D$ is independent of $E$ (which we apply in equation (4)) may not hold for disturbance-driven sediment transport; if true, this would introduce significant complexity to efforts to utilize hillslope topography to predict erosion rates.

7. Conclusions

[39] Despite similar vegetation, the hillslope morphology in two distinct lithologies in the Feather River region of California varies significantly, with hillslopes in metamorphosed volcanic rocks tending to be steeper and longer than those in a granodiorite pluton. Variation of the sediment transport coefficient is inferred from hilltop curvature in rapidly eroding portions of the landscape, with the sediment transport coefficient being lower in metavolcanics ($4.8 \pm 1.8$ m$^2$ ka$^{-1}$) than in the granodiorite ($8.8 \pm 3.3$ m$^2$ ka$^{-1}$). The study area is undergoing a transient adjustment to accelerated baselevel fall and therefore exhibits a large range in hilltop curvature (considered an indicator of erosion rate at steady state). Hillslope gradient increases monotonically with hilltop curvature until approaching a critical hillslope gradient $S_c$. The range of erosion rates facilitates estimation of $S_c$ at 0.79 and 0.85 in the granodiorite and metavolcanics, respectively. Hillslopes on metavolcanics tend to have a steeper mean gradient at low hilltop curvature, indicating that they are in a transient stage of adjustment to increased erosion rate. Hillslope lengths get shorter as hilltop curvature increases, suggesting that drainage density is coupled to the rate of erosion during adjustment. We conclude that lithologic variability in a landscape can influence rates of sediment transport, influencing the topographic form of hillslopes and that lithology influences the degree of landscape dissection.

[40] Acknowledgments. This work was supported by a National Environmental Research Council (NERC) Doctoral Training Grant NE/G524128/1 awarded to M. D. Hurst and NERC NE/H001174/1 to S. M. Mudd. Funding for this work was also provided by the National Science Foundation EAR0819064 (Empirical and Theoretical Integration of Geochemical and Morphologic Evolution of Soil-Covered Hillslopes: Responses to Channel Incision) to K. Yoo and S. M. Mudd for which LiDAR topographic data were acquired by the National Center for Airborne Laser Mapping. We are grateful to Edward Mitchard for guidance on estimating AGB from discrete return LiDAR data and Lyndsey Mackay for GIS guidance. We also thank the associate editor Simon Brocklehurst, Josh Roering, and two anonymous reviewers whom comments and insights helped refine and improve this contribution. We are particularly grateful to editor Alex Densmore for his insights and diligence.

References

Ahern, F. (1970), Functional relationships between denudation, relief, and uplift in large mid-latitude drainage basins, Am. J. Sci., 268(3), 243–263, doi:10.2475/ajs.268.3.243.

Almond, P. C., J. J. Roering, M. W. Hughes, F. S. Lutter, and C. Lebouteiller (2008), Climatic and anthropogenic effects on soil transport rates and hillslope evolution, Sediment Dynamics in Changing Environments, Christchurch, New Zealand. IAHS Publication, 325, 417–424.

Anderson, R. S. (1994), Evolution of the Santa Cruz Mountains, California, through tectonic growth and geomorphic decay., J. Geophys. Res., 99(B10), 20161–20179, doi:10.1029/94JB00713.

Anderson, D. J., and R. C. Buckman (1987), Fitting degradation of shoreline scarp by a nonlinear diffusion-model, J. Geophys. Res., 92(B12), 12857–12867, doi:10.1029/JB092iB12p12857.

Armitage J. J., R. A. Duller, A. C. Whittaker, and P. A. Allen (2011), Transformation of tectonic and climatic signals from source to sediment archives, Nature Geosci., 4, 231–235.

Arrawosmith, J. R., D. D. Rhodes, and D. D. Pollard (1998), Morphologic dating of scarp forms by repeated slip events along the San Andreas Fault, Carrizo Plain, California, J. Geophys. Res. 103(B5), 10311–10360, doi:10.1029/98JB05005.

Avouac, J. P., and G. Peltzer (1993), Active tectonics in southern Xinjiang, China: Analysis of terrace rise and normal fault scarp degradation along the Hotan-Qira fault system, J. Geophys. Res., 98, 21773–21807, doi:10.1029/93JB02172.

Avouac, J. P., P. Tappinier, M. Bai, H. You, and G. Wang (1993), Active thrusting and folding along the northern Tien-Shan and late Cenozoic rotation of the Tarim relative to Dzungaria and Kazakhstan, J. Geophys. Res. 98(B4), 6755–6804, doi:10.1029/92JB01963.

Begin, Z. B. (1992), Application of quantitative morphologic dating to paleo-seismicity of the northwestern Negev, Israel, Isr. J. Earth Sci., 41, 95–103.

Binnie, S. A., W. M. Phillips, M. A. Summerfield, and L. K. Fifield (2007), Tectonic uplift, threshold hillslopes, and denudation rates in a developing mountain range, Geology, 35(8), 743–746, doi:10.1130/G23641A.1.

Burbank, D., and R. Gerson (1986), Morphology of the latest Quaternary surface-faulting in the Gulf of Elat region, eastern Sinai, Tectonophysics, 128, 97–119, doi:10.1016/0040-1951(86)90310-0.

Bowman, D., and T. Gross (1986), Neotectonics in the northern Araba: Research report to the Israel Department of Energy (in Hebrew).

Burbank, D. W., J. Leland, E. Fielding, R. S. Anderson, N. Brozovic, M. R. Reid, and C. Duncan (1996), Bedrock incision, rock uplift and threshold
Schmidt, K. M., and D. R. Montgomery (1995), Limits to relief, Science, 270(5236), 617–620.

Sklar, L. S., and W. E. Dietrich (2004), A mechanistic model for river incision into bedrock by saltating bed load, Water Resour. Res., 40, W06301, doi:10.1029/2003WR002496.

Small, E. E., R. S. Anderson, and G. S. Hancock (1999) Estimates of the rate of regolith production using 10Be and 26Al from an alpine hillslope, Geomorph., 27, 131–150.

Spelz, R. M., J. M. Fletcher, L. A. Owen, and M. W. Caffee (2008), Quaternary alluvial-fan development, climate and morphological dating of fault scarps in Laguna Salada, Baja California, Mexico, Geomorphology, 102, 578–594, doi:10.1016/j.geomorph.2008.06.001.

Stock, J. D., and W. E. Dietrich (2006), Erosion of steepland valleys by debris flows, Geol. Soc. Am. Bull., 118(9–10), 1125–1148, doi:10.1130/B25902.1.

Stock, G. M., R. S. Anderson, and R. C. Finkel (2004), Pace of landscape evolution in the Sierra Nevada, California, revealed by cosmogenic dating of cave sediments, Geology, 32(3), 193–196, doi:10.1130/G20197.1.

Tucker, G. E., and R. L. Bras (1998), Hillslope processes, drainage density and landscape morphology, Water Resour. Res., 34(10), 2751–2764, doi:10.1029/98WR01474.

Tucker, G. E., and D. N. Bradley (2010) Trouble with diffusion: Reassessing hillslope erosion laws with a particle-based model, J. Geophys. Res., 115, F00A10, doi:10.1029/2009JF001264.

Tucker, G. E., and G. R. Hancock (2010), Modelling landscape evolution, Earth Surf. Proc. Land., 35, 28–50.

Walther, S. C., J. J. Roering, P. C. Almond, M. W. Hughes (2009), Long-term biogenic soil mixing and transport in a hilly, loess-mantled landscape: Blue Mountains of southeastern Washington, Catena, 79, 170–178, doi:10.1016/j.catena.2009.08.003.

Warbington, R. and D. Beardsley (2002), Estimates of old growth forests on the 18 National Forests of the Pacific Southwest Region, USDA Forest Service.

Walther, S. C., J. J. Roering, P. C. Almond, M. W. Hughes (2009), Long-term biogenic soil mixing and transport in a hilly, loess-mantled landscape: Blue Mountains of southeastern Washington, Catena, 79, 170–178, doi:10.1016/j.catena.2009.08.003.

Whipple, K. X., and G. E. Tucker (1999), Dynamics of the stream-power river incision model: Implications for height limits of mountain ranges, landscape response timescales, and research needs, J. Geophys. Res., 104(B8), 17661–17674, doi:10.1029/1999JB900120.

Whittaker, A. C., M. Attal, P. A. Allen (2010), Characterising the origin, nature and fate of sediment exported from catchments perturbed by active tectonics, Basin Res., 22, 809–828.

Yoo, K., R. Amundson, A. M. Heimsath, and W. E. Dietrich (2005), Process based model linking pocket gopher (Thomomys bottae) activity to sediment transport and soil thickness, Geology, 33(11), 917–920.

Yoo, K., and S. M. Mudd (2008), Toward process-based modeling of geochemical soil formation across diverse landforms: A new mathematical framework, Geoderma, 146(1–2), 248–260.

Yoo, K., B. Weinman, S. M. Mudd, M. Hurst, M. Attal, and K. Maher (2011), Evolution of hillslope soils: The geomorphic theater and the geochemical play, Appl. Geochem., 26, S149–S153, doi:10.1016/j.apgeochem.2011.03.054.