

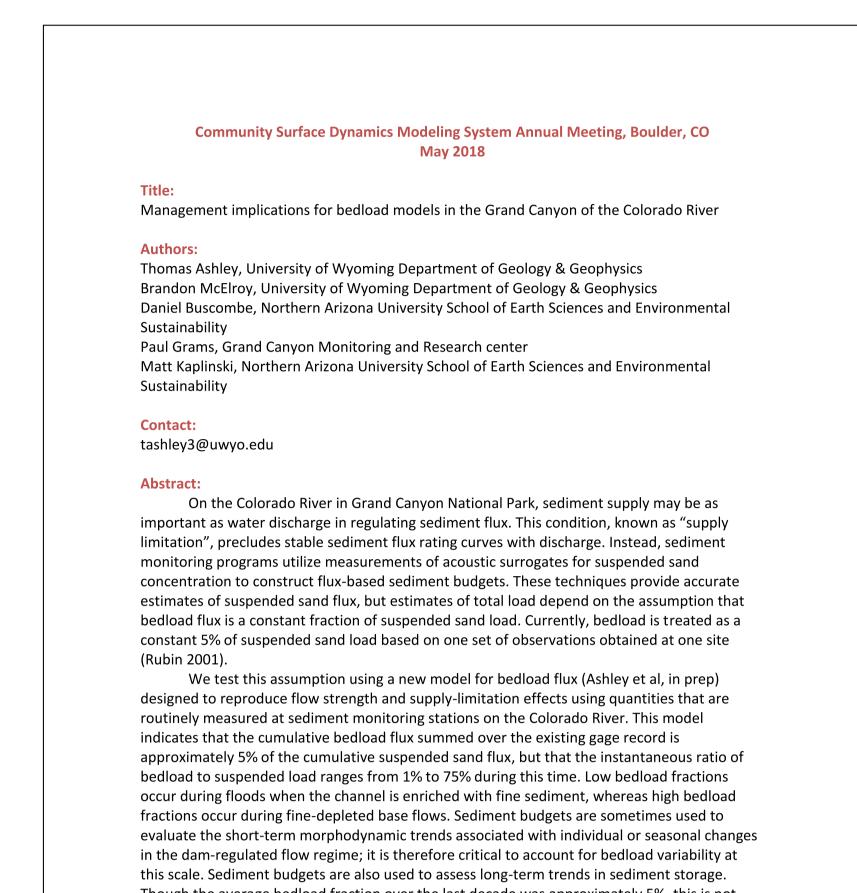


Management Implications for Bedload Models in the Grand Canyon of the Colorado River

Thomas Ashley^{1,*}, Brandon McElroy¹, Daniel Buscombe³, Paul Grams², Matt Kaplinski³

¹ University of Wyoming, Department of Geology and Geophysics, Laramie, WY ² Grand Canyon Monitoring and Research Center, U.S. Geological Survey, Flagstaff, AZ ³ Northern Arizona University, School of Earth Sciences and Environmental Sustainability, Flagstaff, AZ

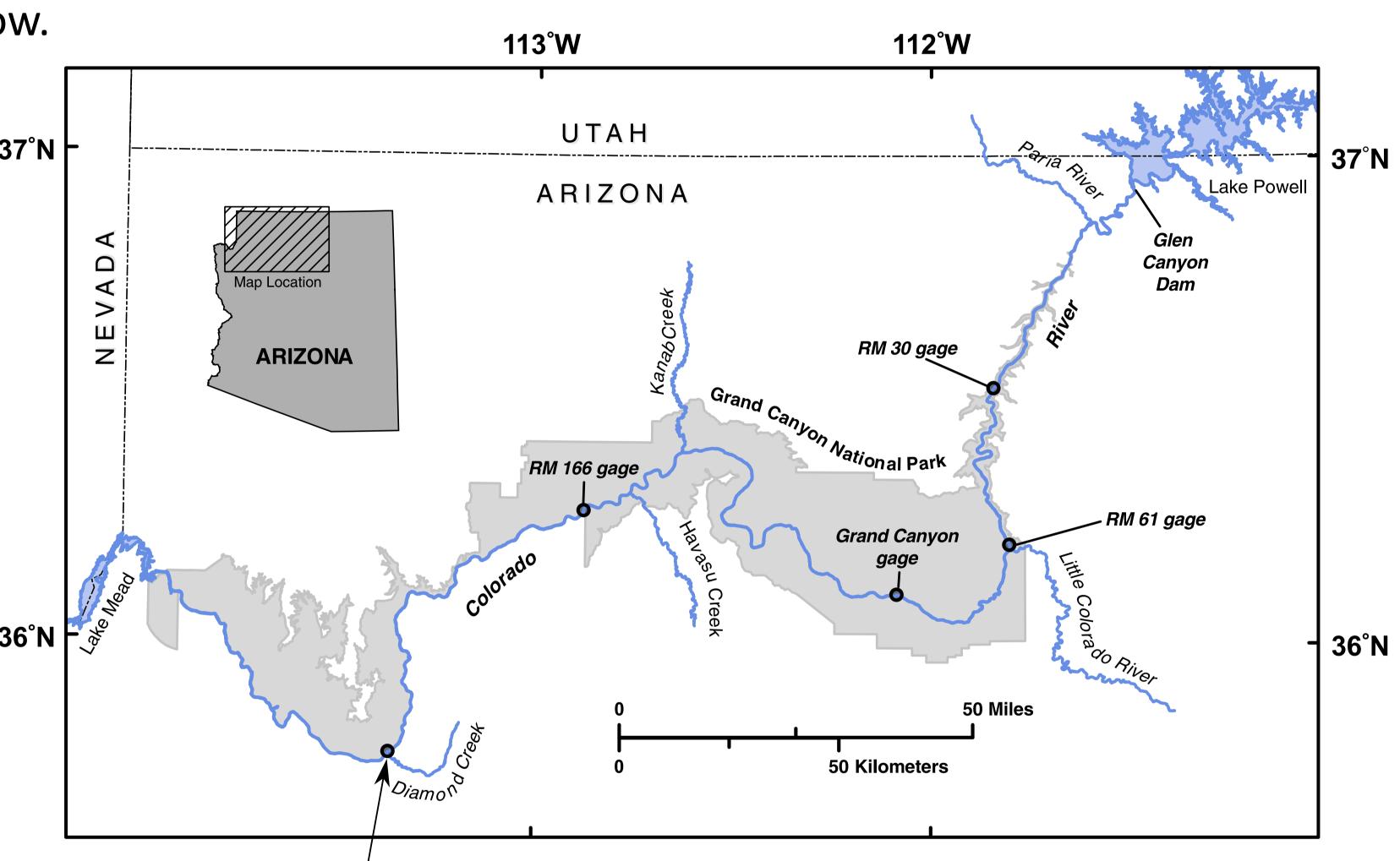
ABSTRACT



land use and climate are likely to impact the rate of tributary sand delivery.

(1) SETTING

Gage estimates of sediment flux are an imortant tool for quantifying large-scale geomorphic change. On the Colorado River, estimates of sediment flux are used to constrain sediment mass-balance in five discrete reaches. Acoustic sediment monitoring instruments provide reliable estimates of suspended load but bedload is estimated as a constant 5% of suspended sand load. Here, we utilize a recently developed bedload model to test this assumption at one monitoring station shown



Diamond Creek sediment monitoring station (River Mile 225)

2 BEDLOAD MODEL

Independent Variables:

Unit water discharge S[-]Reach slope

Median diameter of bed material

 $D_b [m]$

Dependent Variables:

 $q_b \ [m^2/s]$ Unit bedload flux

suspended sand

Suspended sand concentration Median diameter of D_s [m]

Power-law approximation for underlying physics:

$$egin{align} q_b \propto Q^{lpha_{11}} S^{lpha_{12}} D_b^{lpha_{13}} \ C_s \propto Q^{lpha_{21}} S^{lpha_{22}} D_b^{lpha_{23}} \ D_s \propto Q^{lpha_{31}} S^{lpha_{32}} D_s^{lpha_{33}} \ \end{array}$$



PHYSICS __

(Garcia, 2008;

Wright & Parker,

$q_b = Ae^{\beta_0}Q^{\beta_1}C^{\beta_2}Q^{\beta_3}$ where $A=q_{b0}/Q_0^{eta_1}C_{s0}^{eta_2}D_{s0}^{eta_3}$

For more information on modeling, see "Statistical bedload modeling on the Colorado River in Grand Canyon National Park" (T3. Advances in River Science in the Intermountain West I, Thursday May 17 (tomorrow), 9:45 AM)

Works Cited:

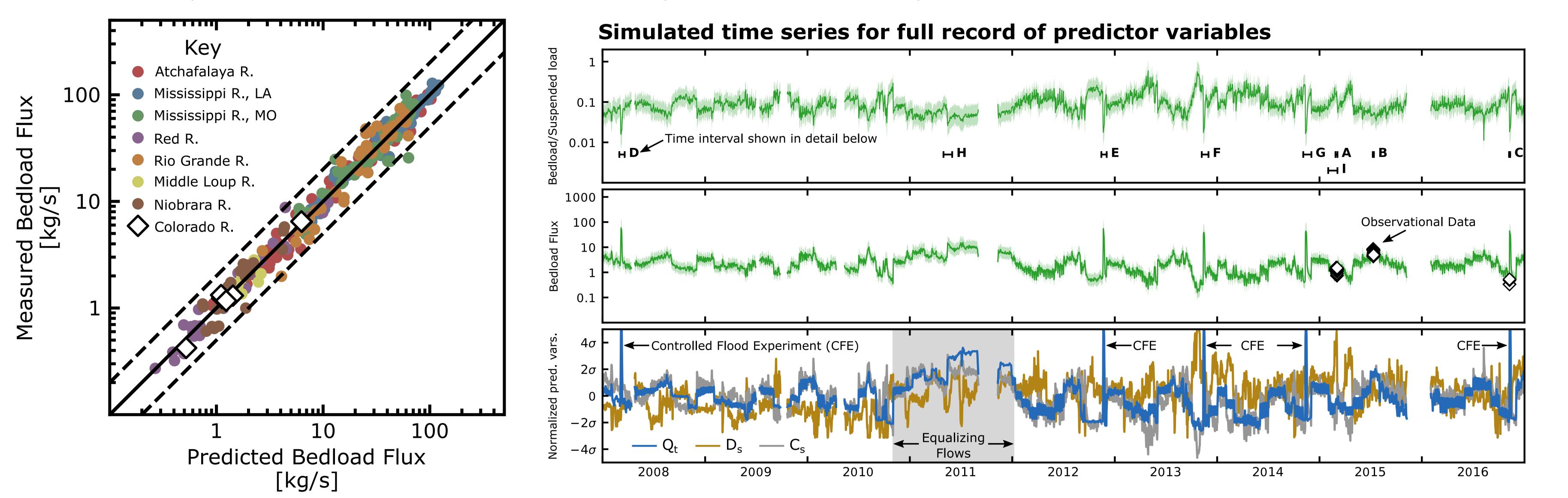
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Gelman, A., Carlin, J. B., Stern, H. S., Dunson, D. B., Vehtari, A., & Rubin, D. B. (2014). Bayesian data analysis (Vol. 2). Boca Raton, FL: CRC press.

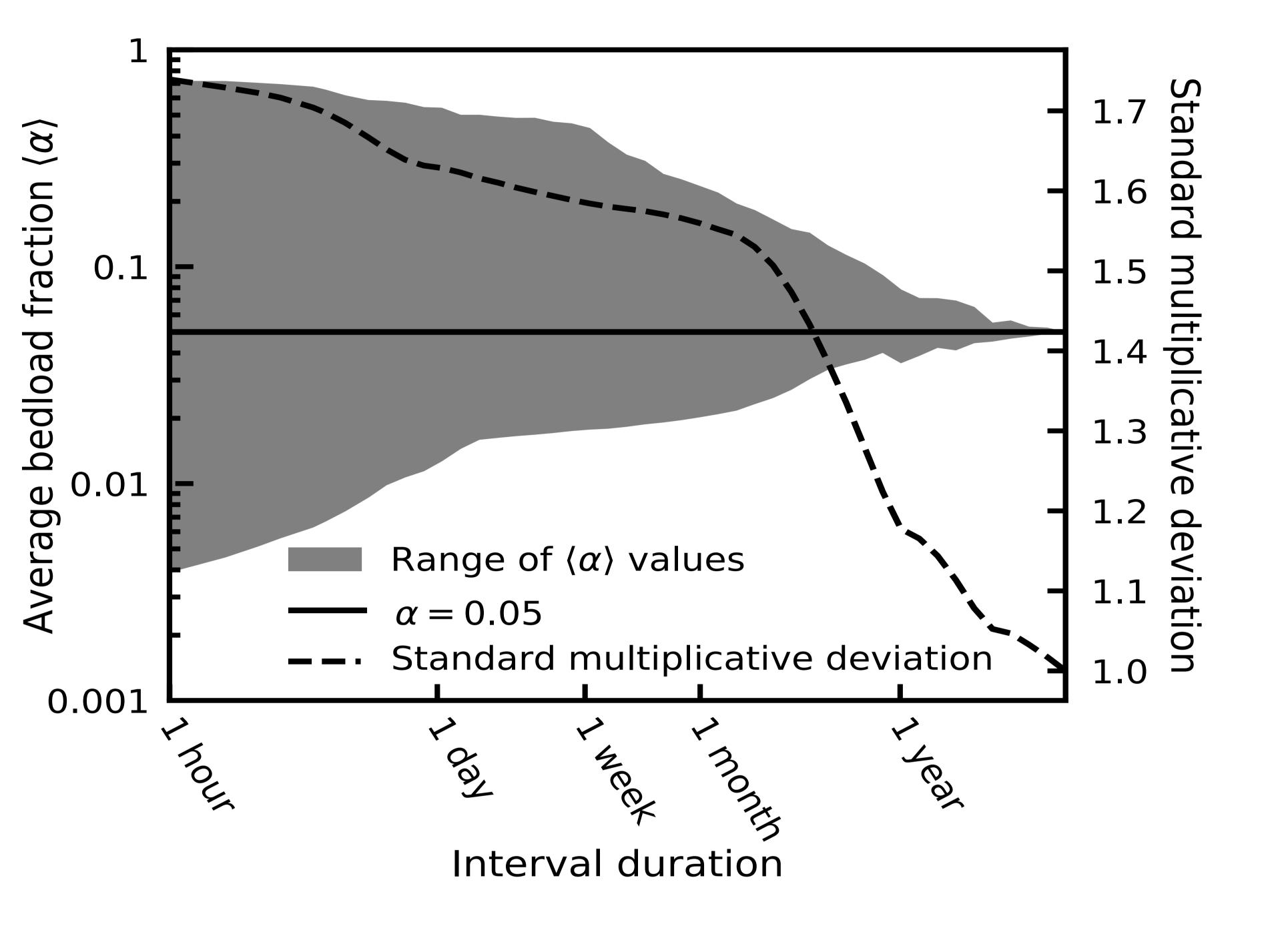
Toffaleti, F. B. (1968). A Procedure for Computation of the Total River Sand Discharge and Detailed Distribution, Bed to Surface (No. TR-5). Army Cor. of Eng. Comm. on Chan. Stab. Wright, S., and G. Parker (2004), Flow Resistance and Suspended Load in Sand-Bed Rivers: Simplified Stratification Model, J. Hydraul. Eng., 130(8), 796–805.

3 PARAMETER ESTIMATION AND PREDICTION

Parameter estimation and prediction is performed using Bayesian hierarchical linear regression (Gelman et al., 2014). Data collected at our study site was supplemented by data from seven other sites on other rivers reported by Toffaleti (1968). Bedload flux and the ratio of bedload to suspended load were then simulated over the full record of predictor variables at our study site.

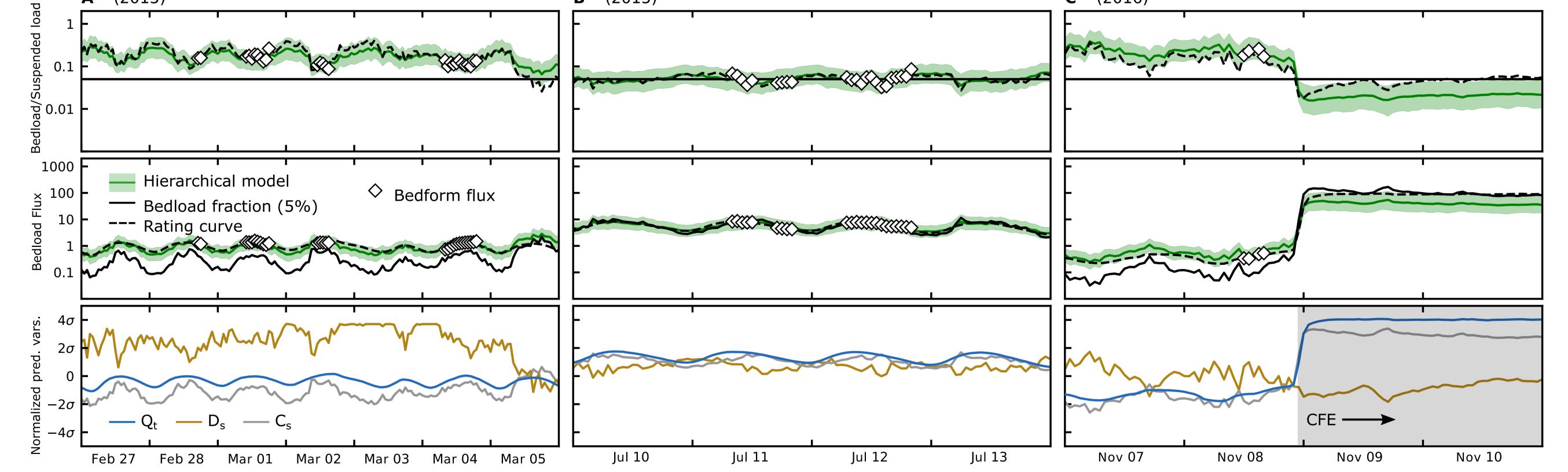


5 TIMESCALE-DEPENDENT AVERAGING

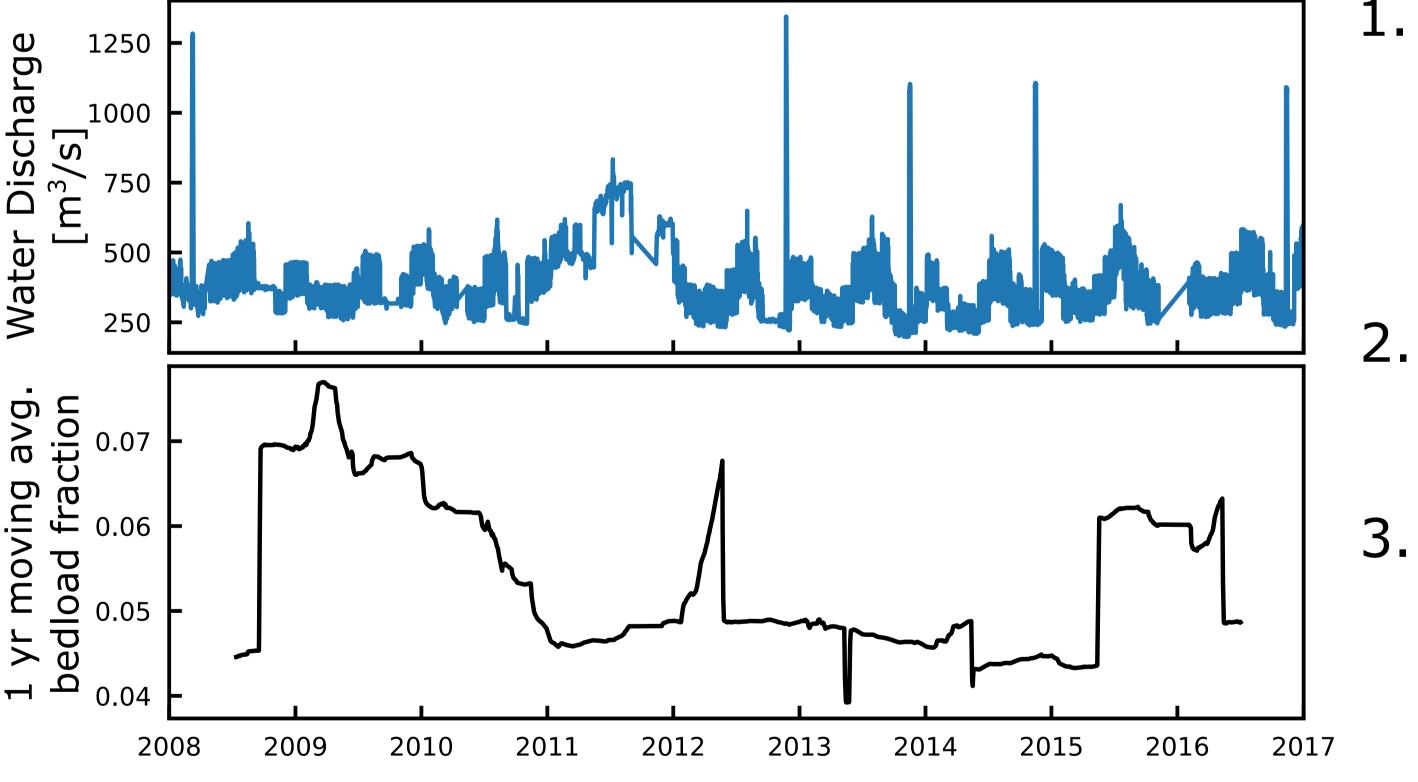


4 SIMULATED BEDLOAD TIME SERIES EXAMPLES

Compared with estimates obtained from repeat bathymetric surveys of dune migration:



6 CONCLUSIONS



- . Cumulative bedload flux over the full record is approximately 5% of cumulative suspended sand flux. However, future changes in sediment supply or water discharge could cause changes in the long-term average bedload fraction
- . Instantaneous bedload flux may range from less than 1% to as much as 75% of suspended sand
- 3. High bedload fractions occur during fine sedimentdepleted base flow conditions, while low bedload fractions occur during fine sediment-enriched flood



