

Alternative(s) to fractional-diffusion equations in bedload transport models

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Motivation

Approaches to bedload
transport

Model extension

Bedform dynamics

Conclusions

Outline

- Fluctuations in bedload transport rates
- Approaches to ‘stochastic’ bedload transport rate
- Extension of a birth-death Markov process
- Comparison with experiments

Motivation

- Fluctuations in sediment transport
- Bedform

Approaches to bedload transport

Model extension

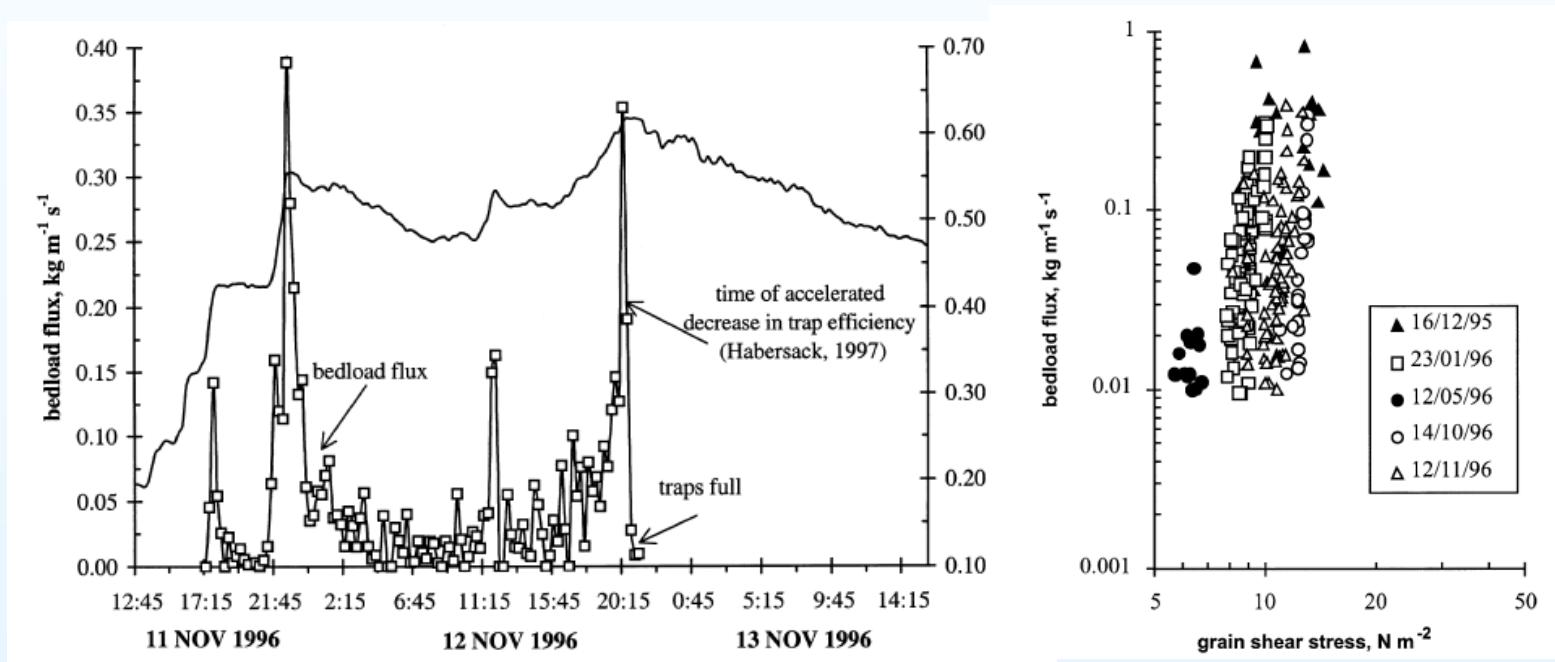
Bedform dynamics

Conclusions

Fluctuations in sediment transport

Evidence that bedload transport is highly fluctuating, with pulses more or less correlated with the water discharge

Tordera River (Spain) gravel-bed river (2% slope)



Garcia et al., Geomorphology, 34, 23–31, 2000.

Bedform

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- Fluctuations in sediment transport
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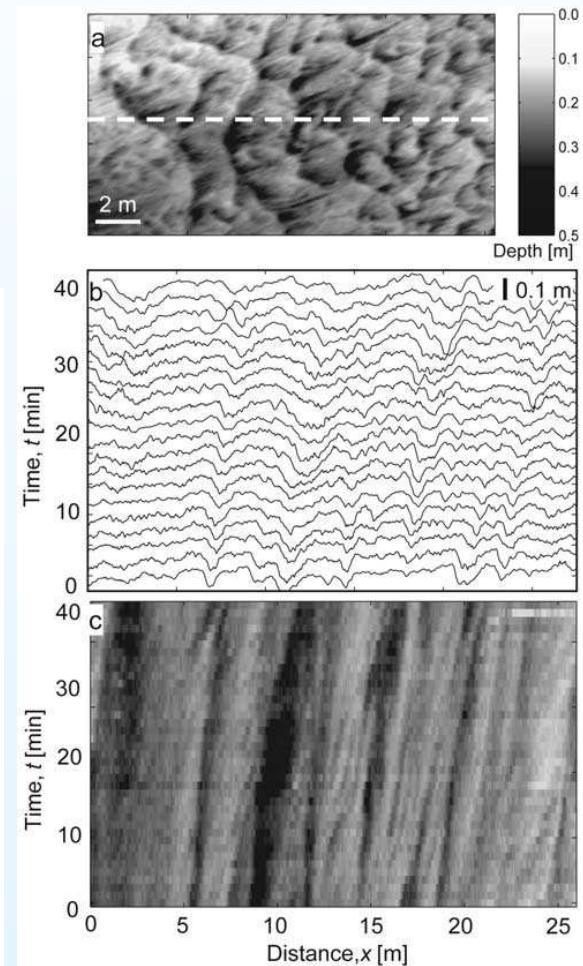
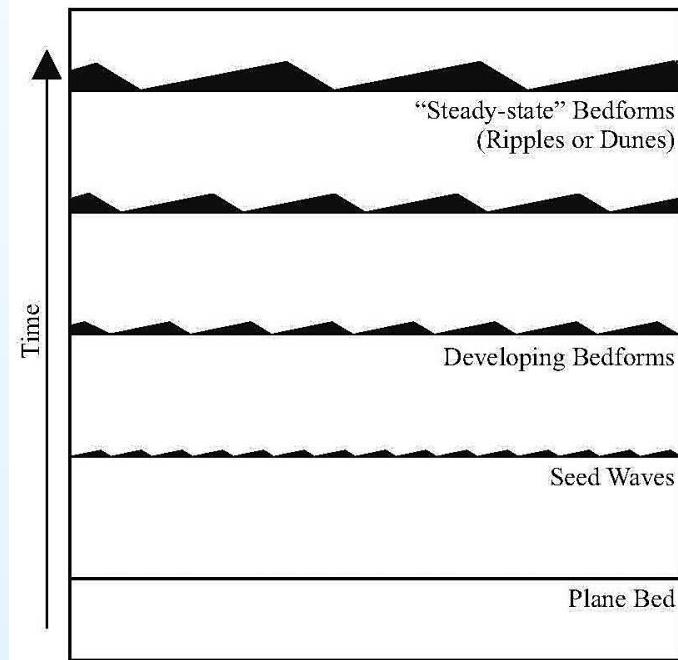
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Bedforms: chaotic or regular forms?



Coleman & Melville, *J. Hydr. Eng.*, **122** 301–310, 1996.
Jerolmack & Mohrig, *Water Resour. Res.*, **41** W12421, 2005.

Motivation

Approaches to bedload transport

- Einstein and Hamamori's distributions

- Advection diffusion equation

- Markov model

- Steady-state solution

Model extension

Bedform dynamics

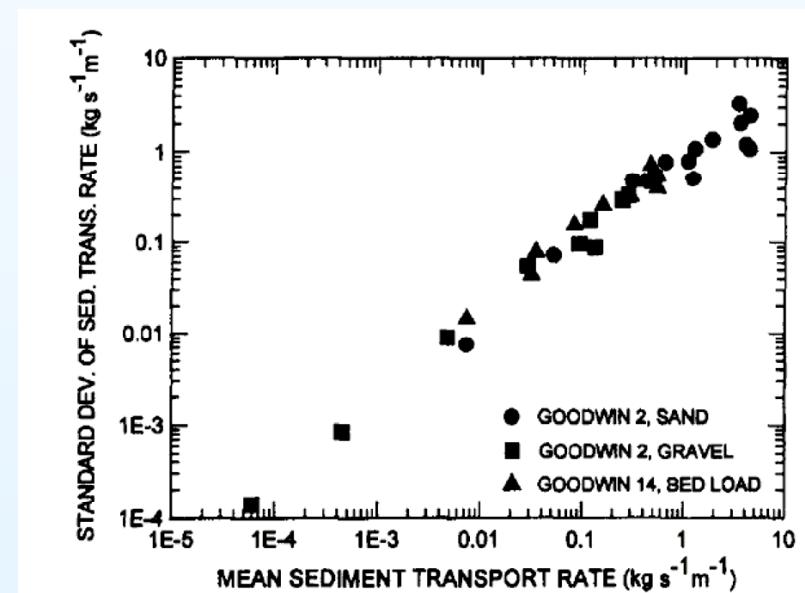
Conclusions

Einstein and Hamamori's distributions

$$P(z) = -\frac{\sqrt{7}}{12} \ln \left(\frac{\sqrt{7}}{12} z + \frac{1}{4} \right)$$

with $z = (Q_s - \bar{Q}_s)/\sqrt{\text{var } Q_s}$ and $\text{var } Q_s = 7\bar{Q}_s^2/9$.

Goodwin Creek data



Kuhnle & Willis, *J. Hydr. Eng.* **124**, 1109–1114 1998.

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Advection diffusion equation

The classic advection diffusion equation (ADE)

$$\frac{\partial c}{\partial t} = -u \frac{\partial c}{\partial x} + \frac{\partial}{\partial x} \left(K \frac{\partial c}{\partial x} \right)$$

can be generalized into a fractional ADE

$$\frac{\partial c}{\partial t} = u \frac{\partial c}{\partial x} + K \frac{\partial^\gamma c}{\partial x^\gamma}$$

with c particle concentration, K diffusivity, γ fractional exponent ($0 < \gamma < 2$, $\gamma = 2$ gives the classic ADE).

Few experimental evidence to date, but Nikora *et al.* found

- var $d(t) \propto t^\gamma$ with $\gamma \geq 2$ for short timescales (i.e., seconds)
- $\gamma < 2$ for long timescales (i.e., minutes)

Nikora *et al.*, *Water Resour. Res.* **38**, WR000513, 2002

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- **Markov model**
- Steady-state solution

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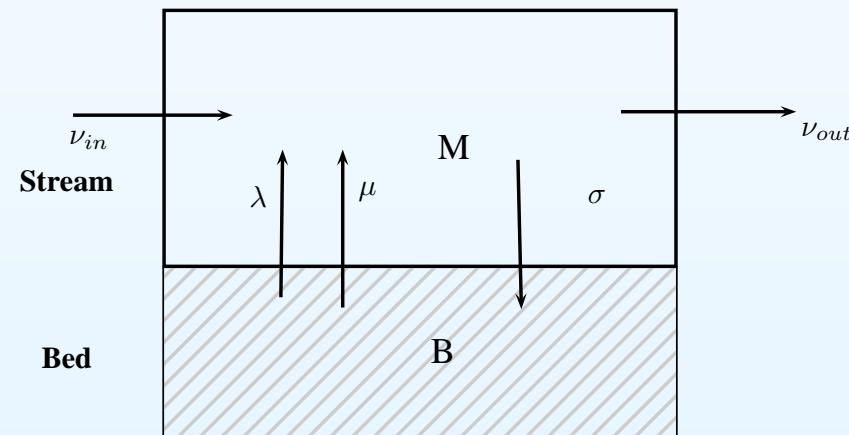
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Markov model

Emigration-immigration birth-death model

$$\frac{\partial}{\partial t} P(n; t) = (n+1)\alpha P(n+1; t) + (\beta + (n - 1)\mu) P(n-1; t) - (\beta + n(\alpha + \mu)) P(n; t),$$



Ancey et al., *J. Fluid. Mech.* **595**, 83–114, 2008;
 Ancey, *J. Geophys. Res.*, **115** F00A11, 2010.

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Steady-state solution

Under steady state conditions, the solution is the negative binomial distribution

$$P(n) = \text{NegBin}(n; r, p) = \frac{\Gamma(r+n)}{\Gamma(r)n!} p^r (1-p)^n, n = 0, 1, \dots,$$

with $r = (\lambda + \nu_{in})/\mu$ and $p = 1 - \mu/(\sigma + \nu_{out})$, and where Γ denotes the gamma function.

$$\bar{N} = r \frac{1-p}{p} = (\lambda + \nu_{in})/(\sigma + \nu_{out} - \mu)$$

$$\text{var}N = r \frac{1-p}{p^2} = \frac{(\lambda + \nu_{in})(\sigma + \nu_{out})}{(\sigma + \nu_{out} - \mu)^2}.$$

Note: (i) a tail falling off like $n^r(1-p)^n$. (ii) When $p \rightarrow 0$, $\sqrt{\text{var}N}/\bar{N} \rightarrow \sqrt{r}$

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- Exner equation

- Large-number approximation
- Expansion
- Comparison with data
- Comparison with data

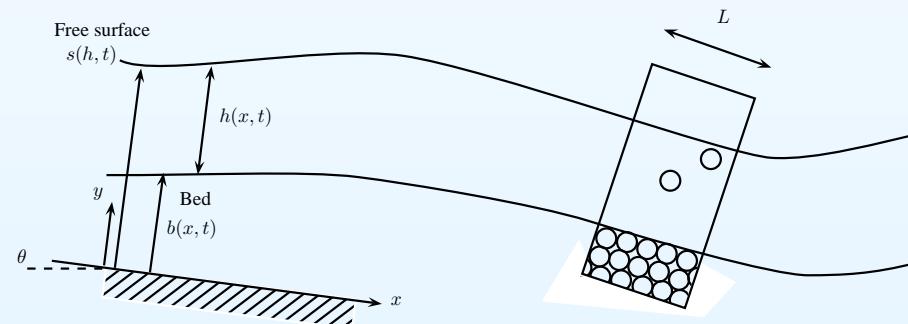
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Exner equation

Bed mass conservation

$$(1 - \phi_p) \frac{\partial b}{\partial t} = D - E = -\frac{\partial q_s}{\partial x},$$

ϕ_p bed porosity, D (E , resp.) volume rate of deposition (entrainment, resp.) per unit time and per unit bed surface onto the bed, and q_s solid discharge per unit width.



$$E = \frac{v_p}{L}(\lambda + \mu N) = v_p(\tilde{\lambda} + \mu \tilde{N}) \text{ and } D = \frac{v_p}{L}\sigma N = v_p\sigma \tilde{N}.$$

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Large-number approximation

The master equation reads

$$\frac{\partial}{\partial t} P(n; t) = (n+1)\alpha P(n+1; t) + \{\beta + (n-1)\mu\} P(n-1; t) - \{\beta + n(\alpha + \mu)\} P(n; t),$$

In the $N \rightarrow \infty$ limit, $P(n+1; t) \approx P(N) + 1 \cdot P'(N)$; the master equation is approximated by a Fokker-Planck equation

$$\frac{\partial}{\partial t} P(n; t) = -\frac{\partial}{\partial n}(AP) + \frac{1}{2}\frac{\partial^2}{\partial n^2}(BP),$$

with $A = \beta + n(\mu - \alpha)$ (the *drift function*) and $B = (\alpha + \mu)n + \beta$ (the *diffusion function*). This PDE is equivalent to the SDE

$$dN = A(N)dt + B^{1/2}(N)d\mathcal{W}(t),$$

where $\mathcal{W}(t)$ is the Wiener process.

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Expansion

Introducing $\epsilon = L^{-1/2}$ and \tilde{N} (number of moving particles per unit length)

$$\tilde{N} = \underbrace{\tilde{N}_*}_{\text{deterministic component}} + \underbrace{\epsilon \eta_1(t)}_{\text{first-order correction}} + \dots .$$

Deterministic component (“sure function”)

$$\tilde{N}_*(t) = \frac{\tilde{\lambda} - e^{-t(\sigma-\mu)}(\tilde{\lambda} + N_0(\mu - \sigma))}{\sigma - \mu}.$$

The long-term behavior can be approximated by an Ornstein-Uhlenbeck process

$$t \rightarrow \infty, \quad \eta_1 \sim \mathcal{N} \left(0, \frac{D_o}{2k_o} \right) = \mathcal{N} \left(0, \frac{\tilde{\lambda}\sigma}{(\sigma - \mu)^2} \right).$$

Motivation

Approaches to bedload transport

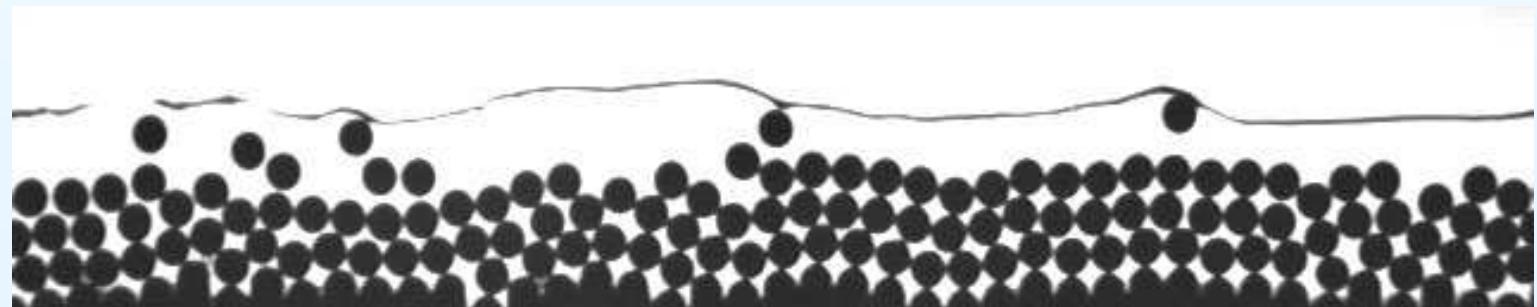
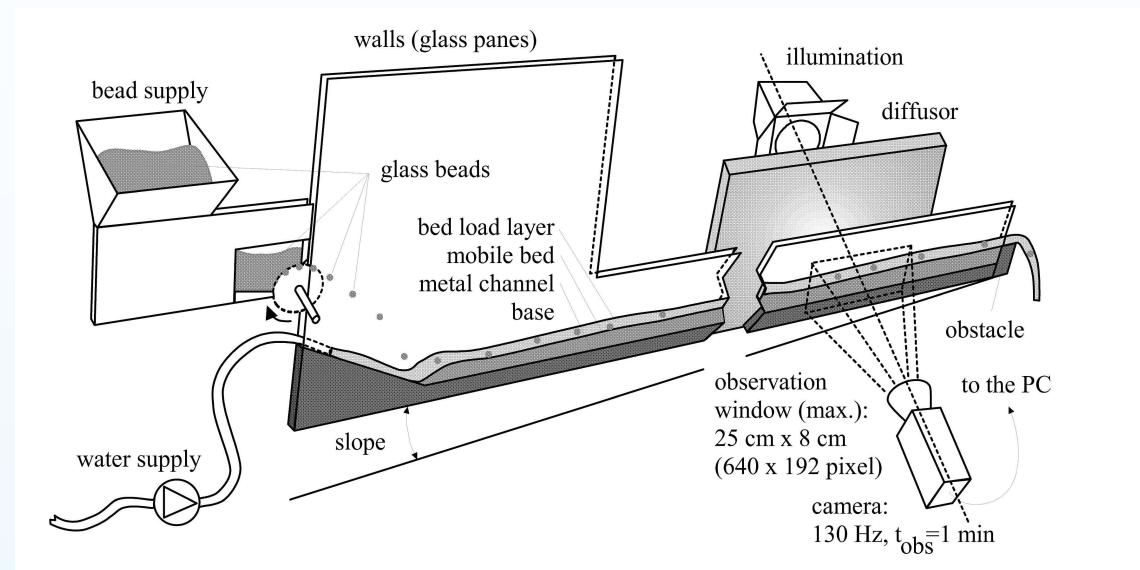
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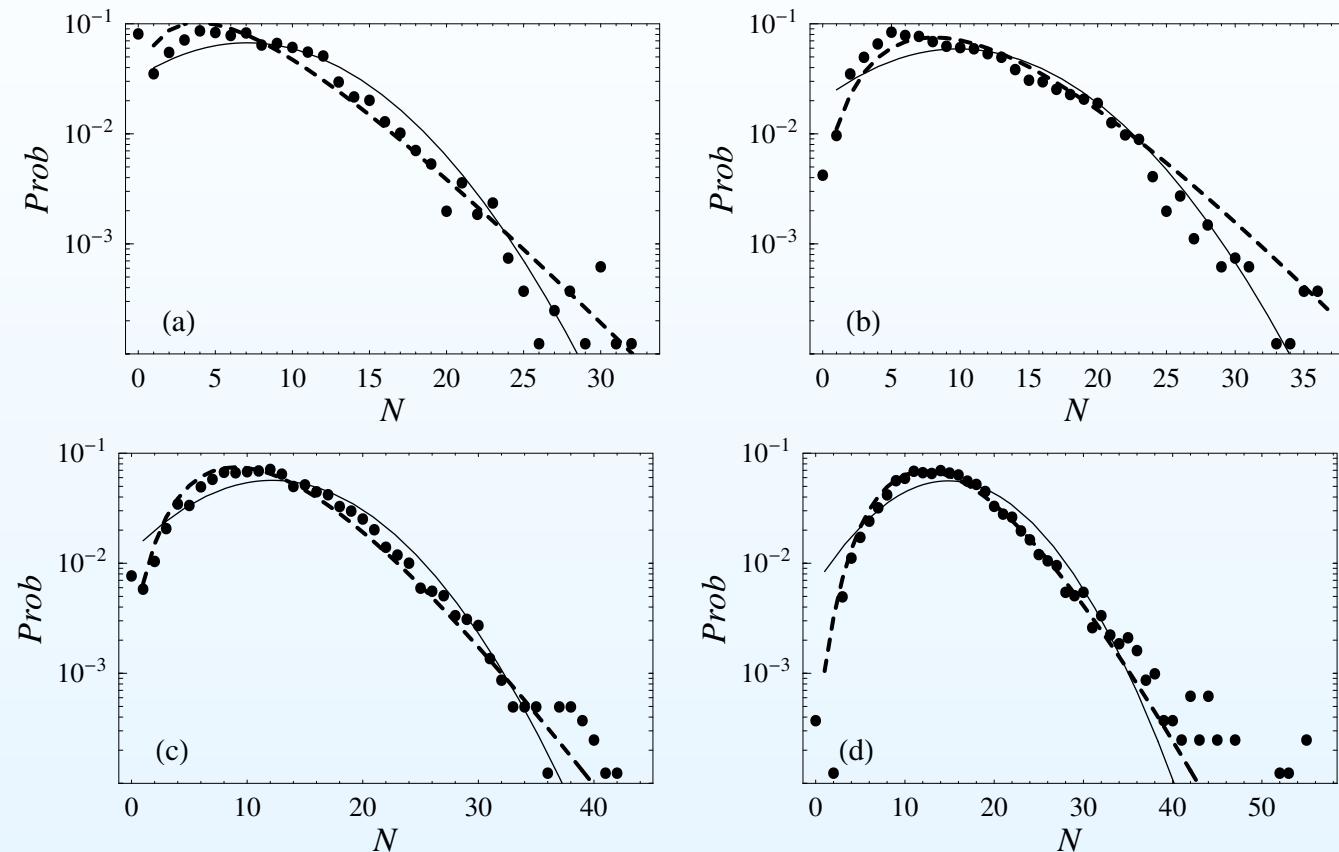


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Comparison with data



dots: data. Dashed line: binomial distribution (steady-state solution).
 Solid line: Gaussian approximation.

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- Implications for the Exner equation

- Stochastic simulations

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Implications for the Exner equation

Recall that in the large-system limit and at long times

$$\tilde{N} = \frac{\tilde{\lambda}}{\sigma - \mu} + \frac{\sqrt{\tilde{B}(n_*)}}{\sigma - \mu} \frac{d\mathcal{W}}{dt},$$

Substituting $b(x, t) = b_0(x, t) + \epsilon b_1(x, t) + \dots$ into the Exner equation gives a hierarchy of equations

$$\psi \frac{\partial b_0}{\partial t} = (\sigma - \mu) \tilde{n}_* - \tilde{\lambda},$$

$$\psi \frac{\partial b_1}{\partial t} = (\sigma - \mu) \eta_1,$$

The first moments of b_1 are $\langle b_1 \rangle = 0$ and $\langle b_1^2 \rangle = \tilde{B}(N_\infty)t$.

$$\mathcal{S}(\omega) = \frac{\tilde{B}(N_\infty)}{2\pi} \frac{1}{\omega^2 + (\sigma - \mu)^2} \propto \omega^{-2} \text{ for } \omega \rightarrow \infty$$

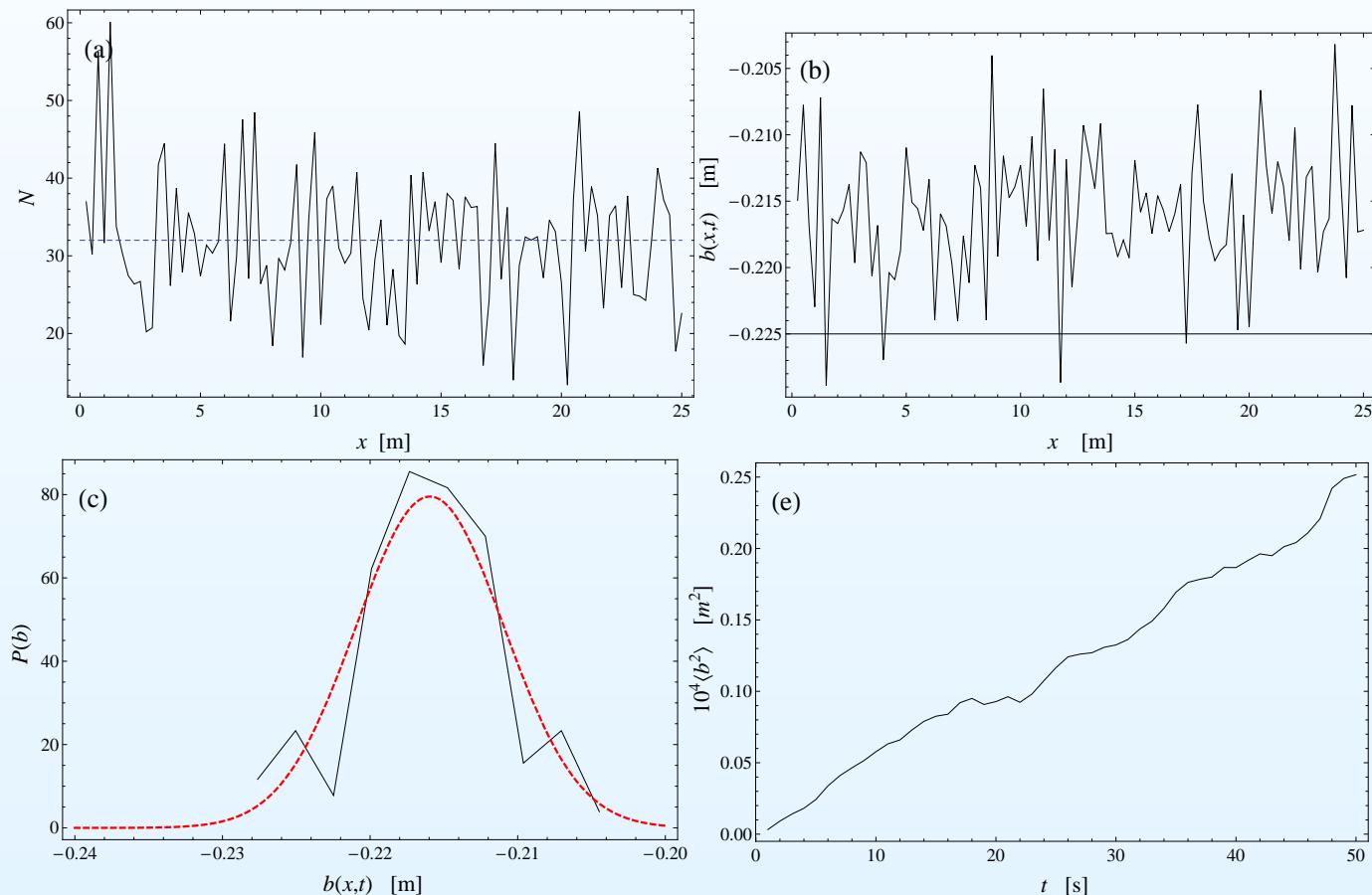
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- Implications for the Exner equation
- **Stochastic simulations**

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Stochastic simulations

Bed evolution can be computed using SDE solvers (e.g., Euler-Maruyama method)



Issue: coupling with the stream

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Conclusion

- reasonably simple framework for computing bedload transport in steep streams
- analytical and numerical results in agreement with experimental data
- explain and describe why wide fluctuations arise

but

- no fluid-bed interaction at the moment
- simplified physical picture (1D stream!)
- very slow numerical schemes