Construction of the zero range process and a deposition model with superlinear growth rates

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Joint work with

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and

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- 1. The zero range process and the bricklayers' process
- 2. Construction materials and the construction
- 3. What have we constructed? Properties
- 4. What we didn't succeed in...





with rate $r(\omega_i)$,











7



8





10

















 ω_i = negative discrete gradient





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with rate $r(\omega_i) + r(-\omega_{i+1})$,

 $\omega_{i} = \text{negative discrete gradient}$ $\omega_{i} \in \mathbb{Z}$ $\omega_{i} = 2$ $\omega_{i+1} = -1$

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→ Independent μ^{θ} -distributed ω_i 's is (formally) the equilibrium of the process. The parameter θ sets the mean of ω_i 's, that is, the slope of the wall. • The construction was available for the case

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• B. 2001 and 2004 finds nice distributions related to shocks in the *exponential* BL process:

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Unfortunately, the process is not constructed at that time.
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+ the previous assumptions for attractivity and the μ^{θ} -equilibrium. Estimates used by Andjel do not work.

2. Construction materials: Equilibrium in finite volume



 \frown : with rate $r(\zeta_i)$ \frown : with rate $r(-\zeta_i)$ \downarrow : with rate $\mathbf{E}^{\mu^{\theta}}r(\zeta_i)$ \checkmark : with rate $\mathbf{E}^{\mu^{\theta}}r(-\zeta_i)$

 $\zeta_i = \text{negative discrete gradient}$

2. Construction materials: Equilibrium in finite volume



 $\begin{array}{c} \checkmark & (\zeta_i) \\ & (\zeta_i)$

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- → Independent μ^{θ} -distributed ζ_i 's ($i = \ell \dots r$) is the equilibrium of the process.
 - θ sets the mean of ζ_i 's, that is, the slope of the wall.

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→ This process is far from equilibrium!!









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- $\stackrel{\text{oupling 1:}}{\Rightarrow} \text{ We have a limit of the monotone processes. Is the limit finite?}$





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- Fix a state $\underline{\omega}(0) \in \widetilde{\Omega}$. Start a monotone process from this state.
- → Coupling 1: The column heights are monotone in ℓ and \mathfrak{r} . → We have a limit of the monotone processes. Is the limit finite? Yes, it is.
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$$\widetilde{\Omega} = \{ \underline{\omega} : \begin{cases} \limsup_{i \to -\infty} \frac{1}{|i|} \sum_{j=i+1}^{0} |\omega_j| < \infty \\ \limsup_{i \to \infty} \frac{1}{i} \sum_{j=1}^{i} |\omega_j| < \infty \end{cases} \}$$

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→ In this case we have a limiting process in infinite volume, for which it is almost impossible to blow up.

 \rightsquigarrow But what is this process?

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- d.) The product measure $\underline{\mu}^{\theta}$ is stationary, and the process in this distribution is ergodic.

3a.) Right-continuity and the Markov property

<u>Idea</u>: the limit for the space-time box $[a, b] \times [0, t]$ is already achieved in finite volume.

That is, a.s. there exist (random) ℓ and \mathfrak{r} , such that the heights $h_i^{[\ell,\mathfrak{r}]}(s)$ of columns of the $[\ell,\mathfrak{r}]$ monotone process agree to the column heights $h_i(s)$ of the limiting process for all $a \leq i \leq b$ and $s \in [0, t]$.

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Right-continuity is OK, and a bit of extra work yields the Markov property as well.



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With the conditional coupling



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it is enough to give a moment bound on the columns $g_i(t)$ (B-C). First moment: $\frac{d}{dt} E g_i(t) = E [r_i(t)] < \text{const.}$

(where $r_i(t) = r(\zeta_i(t)) + r(-\zeta_{i+1}(t))$ is the rate of growth at site *i*).

Second moment:

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$$y' \leq 2\sqrt{y} \cdot \sqrt{\mathbf{E} r_i^2} + \mathbf{E} r_i.$$

Similar procedure works for higher moments as well.

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Formal generator (on cylinder functions φ):

$$L\varphi(\underline{\omega}) = \sum_{i} r_i(\underline{\omega}) \cdot [\varphi(\underline{\omega}^{(i,i+1)}) - \varphi(\underline{\omega})],$$

where $\underline{\omega}^{(i,i+1)} = \underline{\omega}$ + one brick = ..., ω_{i-1} , $\omega_i - 1$, $\omega_{i+1} + 1$, ω_{i+2} , ...

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$$S(t)\varphi(\underline{\omega}) - \varphi(\underline{\omega}) = \int_{0}^{t} S(s)L\varphi(\underline{\omega}) \, \mathrm{d}s$$

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Hence most of these growing columns have grown while the growth rates were moderate. The probability of this is very small.

 \rightarrow <u>So:</u> For small enough t

 \mathbf{P} {every column grew by time t in [0, i]} $\leq e^{-C \cdot i}$.







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 \rightsquigarrow Up to some time $T = T^{\underline{\omega}}$, we have the Kolmogorov forward and backward equations (also in differential form).

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3*d*.) Equilibrium and ergodicity

Based on the equilibrium ζ process, it is natural, and not difficult either, that the measure μ^{θ} is stationary.

Ergodicity is a bit more difficult. The *ergodicity* of such a noncountable state space process is characterized by:

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- Constructing the version in which bricklayers are also allowed to remove bricks from columns (that is, particles are also allowed to jump to the left (ZR)). We haven't tried, but it didn't seem easy.

It exists.

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Thank you.