

I. Review

Showed

$$\text{Var}(\log Z_{N,t}^\theta) \leq CN^{2/3}$$

using i) existence of invariant measure (O'Connell-Yor)

ii) integration by parts

$$E[\log Z_{N,t}^\theta | B_0] = E[s_0^+]$$

iii) estimate $E[s_0^+]$ by varying parameter θ

II. Alternative approach to OY polymer.

$$Z_{N,t}^\theta = \int e^{s_0\theta - B_0(s_0) + \dots + B_N(t) - B_N(s_{N-1})} ds$$

$s_0 < \dots < s_{N-1} < t$

Itô's formula:

$$dZ_{N,t}^\theta = Z_{N-1,t}^\theta dt + Z_{N,t}^\theta dB_N + \frac{1}{2} Z_{N,t}^\theta dt$$

$$d\log Z_{N,t}^\theta = e^{-\log Z_{N-1} + \log Z_N} dt + dB_N$$

Define: $r_j = \log Z_j - \log Z_{j-1} \quad j = 1, \dots, N$ (Recall $Z_0 := e^{B_0 - \theta t}$)

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$$\Rightarrow \begin{cases} dr_i = (e^{-r_i} - \theta) dt + dB_0 + dB_i \\ dr_j = (e^{-r_j} - e^{-r_{j-1}}) dt + dB_j - dB_{j-1}, j \geq 2. \end{cases}$$

This system (or similar ones) have appeared in several places

in the literature: Ferrari, Spohn, Weiss, O'Connell, Diehl, Gubinelli,
Tava, Moreno-Flores, ...

If we let $V(r) = e^{-r}$, we have:

$$\begin{cases} dr_i = (-V'(r_i) - \theta) dt + dB_0 + dB_i \\ dr_j = (-V'(r_j) + V'(r_{j-1})) dt + dB_j - dB_{j-1}. \end{cases}$$

Assuming the system of SDEs has a solution:

Generator

$$L = \frac{1}{2} \partial_{r_i}^2 + \frac{1}{2} \sum_{j=2}^N (\partial_{r_j} - \partial_{r_{j-1}})^2 + \frac{1}{2} \partial_{r_N}^2 - (V'(r_i) + \theta) \partial_{r_i} + \sum_{j=2}^N (V'(r_{j-1}) - V'(r_j)) \partial_{r_j}.$$

Define

$$\omega = \prod_{i=1}^N \frac{e^{-\theta r_i} - V(r_i)}{Z(\theta)},$$

product form

$$Z(\theta) = \int e^{-\theta x} e^{-V(x)} dx \quad (= \Gamma(\theta) \text{ in OY case})$$

$V(x) = e^{-x}$

By direct computation,

$$L^* \omega = 0$$

where L^* is the formal adjoint of L , so

ω is an invariant measure.

$$\mathbb{E}^\omega [F(r_1(0), \dots, r_N(0))] = \mathbb{E}^\omega [F(r_1(t), \dots, r_N(t))].$$

In case $V(x) = e^{-x}$, we recover stationarity of

the increments $\log Z_j - \log Z_{j-1}$.

III. Height function

Analysis in previous lectures was based on decomposition:

$$\begin{aligned} \log Z_{N,t}^\theta &= \sum_{j=1}^N \underbrace{r_j(t)}_{\text{correlated}} - B_\theta(t) + \theta t \\ &= \log Z_j - \log Z_{j-1} \end{aligned}$$

For general V , we can also consider:

$$W_{N,t}^\theta := \sum_{j=1}^N \underbrace{r_j(t)}_{\text{solutions to SDE}} - B_\theta(t) + \theta t$$

It is expected that for "most" V , $W_{N,t}^\theta$ has KPZ

It is expected that for "most" V , $W_{N,t}^6$ has KPZ fluctuations (e.g. Ferrari-Spohn-Weiss, Sasamoto-Spohn ...).

Try to follow the proof for polymers.

$$\text{Var}(W_{N,t}^6) = \text{Var}\left(\sum_{j=1}^N r_j\right) - t + 2E[W_{N,t}^6 B_0(t)]$$

↑
arithmetic

How to analyze $E[W_{N,t}^6 B_0(t)]$?

For polymer, we used Gaussian integration by parts:

$$E[\log Z B_0(t)] = \frac{d}{ds} E\left[\log \int_{-\infty < s_0 < \dots < t} e^{B_{s_0} - \delta s_0^+ - B_0(s_0) + \dots} ds\right]$$

$$= E[E[s_0^+]]$$

If we start from the equations, not clear what $E[\cdot]$ would be, but we can still integrate by parts.

$$E[W_{N,t}^6 B_0(t)] = \frac{d}{ds} E\left[W_{N,t}^6 [B_0(s) + \delta s^+, s < t]\right] \Big|_{\delta=0}$$

Formally:

$$\text{Var}(W_{N,t}^6) = N\psi_6(t) - t + 2E[\partial_{\delta, \text{eq}} W_{N,t}^6]$$

↑

$$\frac{d^2}{d\theta^2} \log Z(\theta)$$

$\partial_{\theta, \text{eq}} W_{N,t}^\theta$: differentiate with respect to parameter θ
only for $t \geq 0$ (not initial data).

IV. Computing $E[\partial_{\theta, \text{eq}} W_{N,t}^\theta]$.

In case $V(u) = e^{-x}$, we can compute $W_{N,t}^\theta$ explicitly
and

$$\partial_{\theta, \text{eq}} \log Z_{N,t}^\theta$$

$$= \partial_\theta \log \int_{-\infty < s_0 < \dots < t} e^{-\gamma s_0^- + \theta s_0^+ - B_0(s_0) + \dots + B_N(t) - B_N(s_{N-1})} ds \Big|_{\gamma=0}.$$

In general, not possible to compute exactly, but we can
differentiate the equations:

$$\begin{cases} dr_i = (-v'(r_i) - \theta)dt + dB_0 + dB_1 \\ dr_j = (-v'(r_j) + v'(r_{j-1}))dt + dB_j - dB_{j-1} \\ r_j(0) \sim \text{stationary} \end{cases}$$

\downarrow

∂_θ

$$\begin{cases} dv_i = (-v''(r_i)v_i - 1)dt \\ dv_j = (-v''(r_j)v_j + v''(r_{j-1})v_{j-1})dt \\ v_j(0) = 0 \end{cases}$$

The equations are linear but depend on solutions:

$$\Rightarrow v_i(t) = - \int_0^t e^{- \int_s^t V''(r_i(u)) du} ds \leq 0$$

$$v_j(t) = \int_0^t e^{- \int_s^t V''(r_j(u)) du} V''(r_{j-1}(s)) v_{j-1}(s) ds \leq 0$$

if $V'' \geq 0$.

If $V(x) = \frac{x^2}{2} \Rightarrow V'' = 1$, we can compute exactly:

$$E[\partial_{\theta, \text{eq}} W_{N,t}] = \int_0^t s^{N-1} e^{-s} \frac{ds}{(N-1)!}$$

$$\text{Var}(W_{N,t}) \sim N^{\frac{1}{2}} \text{ if } t=N.$$

\uparrow
not $\frac{1}{2}$.

In general, we use monotonicity properties:

$$dW_{N,t}^\theta = -V'(r_N) dt + dB_N$$

$$\left\{ \begin{array}{l} \partial_\theta \\ \end{array} \right.$$

$$d\partial_\theta W_{N,t}^\theta = -V''(r_N) \underbrace{v_N}_{\leq 0} dt \geq 0$$

So parameter in initial data $\theta \mapsto W_{N,t}^{\eta, \theta}$ is increasing.

parameter in SDE

Can also show: if $V'' \geq 0$, $V''' \leq 0$, then

$$\theta \mapsto W_{N,t}^{\eta, \theta} \text{ convex.}$$

Want to apply the strategy we used for OY polymer:

$$\text{Var}(\log Z_{N,t}^\theta) = 2\mathbb{E}[E[s_0^+]] \leq \mathbb{E}[E[s_0]^2]^{1/2} + O(N^{1/2})$$

$$\left. \begin{aligned} &\leq \frac{1}{h} \mathbb{E}[(\log Z^{\theta+h} - \log Z^\theta)^2]^{1/2} + O(N^{1/2}) \\ &+ \frac{1}{h} \mathbb{E}[(\log Z^{\theta-h} - \log Z^\theta)^2]^{1/2} \end{aligned} \right\} \text{Concavity}$$

$$V(x) = e^x \text{ (OY)}$$

$$\log Z_{N,t}^\theta$$

General V

$$W_{N,t}^\theta$$

$$E[s_0^+]$$

$$\partial_{\theta, \text{eq}} W_{N,t}^\theta$$

$$E[s_0]$$

$$\frac{d}{d\theta} W_{N,t}^\theta$$

II. Initial data

For polymer, initial data is built in:

$$Z_{0,t}^{\eta, \theta} = \int e^{\theta s_0^+ - \eta s_0^- - B_0(s_0) + \dots + B_N(t) - B_N(s_{N-1})} d_s$$

$$Z_{N,t}^{\eta, \theta} = \int_{-\infty < s_0 < \dots < t} e^{\theta s_0^+ - \eta s_0^- - B_0(s_0) + \dots + B_N(t) - B_N(s_{N-1})} ds$$

For general V , we defined

$$W_{N,t} = \sum_{j=1}^N v_j(t) - B_0(t) + \theta t$$

where

$$\begin{cases} dv_1 = -V'(r_1)dt - \theta dt + dB_0 + dB_1 \\ dv_j = (-V'(r_j) + V'(r_{j-1}))dt + dB_j - dB_{j-1} \\ v_j(0) \sim f_j(\eta) \end{cases}$$

Want to choose initial data so $\eta \mapsto W_{N,t}^{\eta, \theta}$ has good properties. For example:

$$\frac{d^2}{d\theta^2} W_{N,t}^{\theta, \theta} \geq 0.$$

Idea: use the equations to create initial data:

$$f_j(\eta) = \lim_{T \rightarrow \infty} \tilde{r}_j(\eta)(T)$$

$$\begin{cases} d\tilde{r}_1 = (-V'(\tilde{r}_1) - \eta)dt + dB_0 + dB_1 \\ d\tilde{r}_j = (-V'(\tilde{r}_j) + V'(\tilde{r}_{j-1}))dt + dB_j - dB_{j-1} \end{cases}$$

Claim: with this choice,

$$\partial_\eta W_{N,t}^{\eta, \theta} \leq 0,$$

$$\partial_\eta^2 W_{N,t}^{\eta, \theta} \geq 0,$$

$$\partial_\theta \partial_\theta W_{N,t}^{\theta,\theta} \geq 0.$$

VI. Estimate

$$\text{Var}(W_{N,t}^{\theta,h}) = N\psi_1(\theta) - t + 2\mathbb{E}[\partial_{\theta,\text{eq}} W_{N,t}^{\theta,0}]$$

$$\leq N\psi_1(\theta) - t + 2\mathbb{E}\left[\left(\frac{d}{dh} W_{N,t}^{\theta,0}\right)^2\right]^{1/2} + O(N^{1/2})$$

$$\leq N\psi_1(\theta) - t$$

$$+ \frac{1}{h} \mathbb{E}\left[(W_{N,t}^{\theta+h} - W_{N,t}^{\theta})^2\right]^{1/2} + O(N^{1/2})$$

If: • $|\mathbb{E}[W_{N,t}^{\theta+h}] - \mathbb{E}[W_{N,t}^{\theta}]| \leq CNh^2$ ✓ (same as ④)

• $|\text{Var}(W_{N,t}^{\theta+h}) - \text{Var}(W_{N,t}^{\theta})| \leq CNh$ (use coupling)

$$\text{Var}(W_{N,t}^{\theta+h}) = N\psi_1(\theta) - t + 2\mathbb{E}[\partial_{\theta,\text{eq}} W_{N,t}^{\theta}]$$

increasing in parameter in equations

decreasing in parameter in initial data

$$\Rightarrow \text{Var}(W_{N,t}^{\theta}) \leq N^{2/3} + N^{1/3} (\text{Var}(W_{N,t}^{\theta}))^{1/2}$$

$$\text{if } |N\psi_1(\theta) - t| \leq CN^{2/3}.$$

$$\Rightarrow \text{Var}(W_{N,t}^{\theta}) \leq CN^{2/3}.$$

$$\Rightarrow \text{Var}(W_{N,t}^{\theta}) \leq C N^{2/3}.$$

Remains to show:

$$E[\partial_{\theta, \text{eq}} W_{N,t}^{\theta,0}] \leq E\left[\left(\frac{d}{d\theta} W_{N,t}^{\theta,0}\right)^2\right]^{1/2} + O(N^{1/2})$$

O.Y.

$$E[E[s_0^+]] \leq [E[E[s_0]^2]]^{1/2} + O(N^{1/2})$$

Key estimate:

$$E[s_0^+] E[s_0^-] \leq C E[(s_0 - E[s_0])^2]$$

$$-\partial_\theta W_{N,t}^{\theta,0} \partial_\gamma W_{N,t}^{\theta,0} \Big|_{\gamma=0} \leq C \frac{d^2}{d\theta^2} W_{N,t}^{\theta,0}.$$

With Noack and Landon, we show:

$$-c_0 \partial_\theta W_{N,t} \partial_\gamma W_{N,t} \leq \partial_{\gamma, \theta}^2 W_{N,t} \quad \text{if } e^{c_0 x} V''(x) \text{ is non-increasing}$$

Idea of proof:

$$A_n = c_0 \partial_\theta W \partial_\gamma W + \partial_{\gamma, \theta}^2 W \Big|_{\gamma=0}$$

$$dA_n = -V''(r_n) A_n + V''(r_n) A_{n-1}$$

$$- [c_0 V''(r_n) + V'''(r_n)] \partial_x r_n \partial_y r_n$$

$$\Rightarrow dA_n \geq -V''(r_n) A_n$$