

A truly pathwise approach to polymer localization

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A truly pathwise approach to polymer localization

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So far, localization statements for directed polymers in random medium deal with the location of the endpoint ("favourite site" for polymer). We introduce a pathwise property, roughly: there exists a "favorite path" depending on the environment as well as the model parameters and time horizon, such that the polymer path has a significant overlap with the favorite path.

In a joint work with [Mike Cranston](#), we establish this property in the parabolic Anderson model. We also obtain complete localization, i.e., the overlap tends to its maximal value 1 as the product (diffusivity \times temperature²) tends to 0.

Outline

- 1 Parabolic Anderson Model
- 2 Replica Overlap
- 3 Polymer Localization
- 4 Complete localization as $\beta^2/\kappa \rightarrow \infty$

The Parabolic Anderson Model

$X = (X(t), t \geq 0)$: symmetric simple random walk in \mathbb{Z}^d with jump rate $\kappa > 0$ starting at 0. Law P_κ , expectation E_κ .

$W_x, x \in \mathbb{Z}^d$: i.i.d. Brownian motions.

Expectation: E .

Anderson polymer model: the Gibbs measure on $\mathcal{D}_T = \mathcal{D}([0, T], \mathbb{Z}^d)$ with

$$\mu_{\kappa, \beta, T}(f) = \frac{1}{Z_{\kappa, \beta, T}} E_\kappa \left[f(X) \exp \left\{ \beta \int_0^T dW_{X(s)}(s) \right\} \right].$$

for $f : \mathcal{D}_T \rightarrow \mathbb{R}$. ($T > 0$ is time horizon.)

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for $f : \mathcal{D}_T \rightarrow \mathbb{R}$. ($T > 0$ is time horizon.) The model has two parameters:

- inverse temperature $\beta > 0$ measuring the fluctuations of the environment,
- diffusivity $\kappa \in (0, \infty)$ of the path under the *a priori* measure P_κ .

Denote

$$H_T(X) = \int_0^T dW_{X(s)}(s); \quad \text{Then, } Z_{\kappa, \beta, T} = E_\kappa \left[\exp \{ \beta H_T(X) \} \right]$$

Lyapunov exponents

$N(T, X)$ = number of jumps of X on $[0, T]$.

Proposition (Microcanonical)

For $r \geq 0$, the following limit exists a.s. and in L^p , $p \in [1, \infty)$:

$$\Gamma(\beta, r) = \lim_{T \rightarrow \infty} T^{-1} \ln E_\kappa [\exp\{\beta H_T(X)\} | N(T, X) = [rT]]$$

It is deterministic, convex in β , continuous in r , independent of κ , and

$$\Gamma(\beta, r) = a \Gamma(a^{-1/2} \beta, a^{-1} r), \quad a > 0.$$

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Let I_{κ} be the Cramér transform of the Poisson distribution with parameter κ ,

$$I_{\kappa}(r) = r \ln(r/\kappa) - r + \kappa, \quad r \geq 0.$$

Proposition (Variational formula for free energy)

$$\Psi(\kappa, \beta) \stackrel{\exists}{=} \lim_{T \rightarrow \infty} T^{-1} \ln Z_{\kappa, \beta, T} = \sup\{\Gamma(\beta, r) - I_{\kappa}(r); r \geq 0\}$$

From scaling relation for Γ ,

$$\Psi(\kappa, \beta) = \beta^2 \Psi(\beta^{-2} \kappa, 1) = \kappa \Psi(1, \kappa^{-1/2} \beta).$$

Define $I_{\kappa, \beta}$,

$$I_{\kappa, \beta}(r) = -\Gamma(\beta, r) + I_{\kappa}(r) + \Psi(\kappa, \beta),$$

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Theorem (Large deviations principle for the number of jumps)

Then

$$\lim_{T \rightarrow \infty, n/T \rightarrow r} T^{-1} \ln \mu_{\kappa, \beta, T}(N(T, X) = n) = -I_{\kappa, \beta}(r), \text{ a.s.}$$

Moreover, for a.e. realization of the environment, and all subsets $B \subset \mathbb{R}_+$,

$$\begin{aligned} -\inf_{B^0} I_{\kappa, \beta}(r) &\leq \liminf_{T \rightarrow \infty} T^{-1} \ln \mu_{\kappa, \beta, T}(N(T, X)/T \in B) \\ &\leq \limsup_{T \rightarrow \infty} T^{-1} \ln \mu_{\kappa, \beta, T}(N(T, X)/T \in B) \leq -\inf_B I_{\kappa, \beta}(r). \end{aligned}$$

Remarks

Literature: R.Carmona-Molchanov '94 (intermittency)

Biskup, P.Carmona, Cranston, Gärtner, van der Hofstad, den Hollander, Hu, König, Korolov, Mörters, Mountford, Rovira, Shiga, Tindel, Viens, Zeldovich... many others!

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Totally asymmetric 1-dim, exactly solvable: O'Connell (with Moriarty, Yor)

(i) **Annealed bound.** By Jensen's inequality, and since $Ee^{\beta H_T(X)} = e^{\beta^2/2}$, both

$$\Psi(\kappa, \beta) \leq \beta^2/2 \quad \text{and} \quad \Gamma(\beta, r) \leq \beta^2/2$$

hold ($\forall \kappa, r$). Then,

$$\Psi(\kappa, \beta) = \beta^2/2 \iff \Gamma(\beta, \kappa) = \beta^2/2,$$

and, in such a case, $I_{\kappa, \beta}(r)$ has a unique minimum at $r = \kappa$.

(ii) **Weak versus strong disorder.**

$$\Psi(\kappa, \beta) = \beta^2/2 \iff \beta^2/\kappa \leq \Upsilon_c,$$

for some critical value $\Upsilon_c \in [0, \infty)$ depending only on the dimension.

- $d \geq 3 \implies \Upsilon_c > 0$
- $d = 1, 2$: $\Upsilon_c = 0$ is expected, in view of results for other models: discrete models (Lacoin, Vargas, ...) or continuous (Bertin)

Asymptotics of Lyapunov exponents

Theorem (Asymptotics of free energy)

As $\beta^2/\kappa \rightarrow \infty$,

$$\Psi(\kappa, \beta) \sim \frac{\alpha^2 \beta^2}{4 \ln(\beta^2/\kappa)}$$

Ref: R.Carmona-Koralov-Molchanov'01, Cranston-Mountford-Shiga'02;
Also R.Carmona-Molchanov-Viens'96.

$$\alpha = \lim_{n \rightarrow \infty} \frac{1}{n} A_n, \quad A_n = \sup_{\gamma: N(n, \gamma) = n} H_n(\gamma).$$

\exists limit: sup of Gaussian process indexed by a set with suitably bounded entropy.

Sketch: Fix T first:

$$\lim_{\beta \rightarrow \infty} \frac{1}{\beta T} \ln E_{\kappa} [e^{\beta H_T(X)} | N(T, X) = [rT]] = \sup_{\gamma: N(T, \gamma) = rT} H_T(\gamma) =: A_{T,r} .$$

$$\begin{aligned} \lim_{\beta \rightarrow \infty} \beta^{-1} \Gamma(\beta, r) &= \lim_{\beta \rightarrow \infty} \lim_{T \rightarrow \infty} \frac{1}{\beta T} \ln E_{\kappa} [e^{\beta H_T(X)} | N(T, X) = [rT]] \\ &= \lim_{T \rightarrow \infty} T^{-1} A_{T,r} \quad (\text{interchange limits}) \\ &= \sqrt{r} \lim_{T \rightarrow \infty} T^{-1} A_T \quad (\text{scaling : } A_{T,r} \stackrel{C}{=} \sqrt{r} A_T) \\ &= \alpha \sqrt{r} \end{aligned}$$

Then, for $\kappa = 1$ and $\beta \rightarrow \infty$,

$$\Psi(1, \beta) \sim \sup \{ \alpha \beta \sqrt{r} - I_1(r) : r \geq 0 \} = \frac{\alpha^2 \beta^2}{4 \ln(\beta^2)}$$

which yields the case of general κ by scaling. □

For the maximizer $r_{\max}(\kappa, \beta)$ in Variational Formula, $r_{\max}(\kappa, \beta) \sim \frac{\alpha^2 \beta^2}{4 \ln^2(\beta^2/\kappa)}$.

Note : For large β^2/κ ,

$$\Psi(\kappa, \beta) \ll \beta^2/2, \quad r_{\max}(\kappa, \beta) \gg \kappa$$

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Two versions of the Overlap

Localization properties are derived through the **overlap** between two independent polymer paths X, \tilde{X} sharing the same environment.

Two versions appear:

$$I_{\kappa,\beta,T} = \frac{1}{T} \int_0^T \mu_{\kappa,\beta,t}^{\otimes 2}(X(t) = \tilde{X}(t)) dt$$

$$J_{\kappa,\beta,T} = \frac{1}{T} \int_0^T \mu_{\kappa,\beta,T}^{\otimes 2}(X(t) = \tilde{X}(t)) dt = \mu_{\kappa,\beta,T}^{\otimes 2} \left(\underbrace{\frac{1}{T} \int_0^T \mathbf{1}\{X(t) = \tilde{X}(t)\} dt}_{\text{proportion of time together}} \right)$$

How do they appear: (i) By Itô's formula,

$$d \ln Z_{\kappa,\beta,t} = \beta \mu_{\kappa,\beta,t}(dW_{X(t)}(t)) + \frac{\beta^2}{2} \left(1 - \mu_{\kappa,\beta,t}^{\otimes 2}(X(t) = \tilde{X}(t)) \right) dt,$$

and, upon integration we get

$$\frac{1}{T} \ln Z_{\kappa,\beta,T} = \frac{1}{T} M_T + \frac{\beta^2}{2} (1 - I_{\kappa,\beta,T})$$

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Limits of overlaps and relation with free energy

As $T \rightarrow \infty$, the martingale term vanishes, so we get (i) below.

Theorem

(i) For all β and κ ,

$$\tilde{I}_{\kappa,\beta,\infty} \stackrel{\exists}{=} \lim_{T \rightarrow \infty} I_{\kappa,\beta,T} = 1 - \frac{2}{\beta^2} \Psi(\kappa, \beta)$$

(j) The limit

$$\tilde{J}_{\kappa,\beta,\infty} = \lim_{T \rightarrow \infty} E[J_{\kappa,\beta,T}]$$

exists except for exceptional values of β^2/κ , and

$$\tilde{J}_{\kappa,\beta,\infty} = 1 - \beta^{-1} \frac{\partial}{\partial \beta} \Psi(\kappa, \beta).$$

Integration by parts

Proof of (j): • Key is the identity

$$\frac{\partial}{\partial \beta} E[\ln Z_{\kappa, \beta, T}] = \beta T [1 - E[J_{\kappa, \beta, T}]] \quad (*)$$

It comes by **integration by parts**, like in spin-glass models (cf. Talagrand's books). The basic, Gaussian integration by parts formula writes

$$EgF(g) = \sigma^2 EF'(g) \text{ for } g \sim \mathcal{N}(0, \sigma^2).$$

With our Brownian environment, we need the integration by parts formula from Malliavin calculus is: for F a smooth function of the $(W_x(t))_{t,x}$ and $h(t, x)$ deterministic, $\|h\|_2 < \infty$,

$$E\left[F \times \int_0^T h(t, x) dW_x(t)\right] = E\left[\int_0^T h(t, x) D_{t,x} F dt\right]$$

The Malliavin derivative $D_{t,x}$ is heuristically equal to $\frac{\partial}{\partial (dW_x(t))}$.

Integration by parts

$$\begin{aligned}\frac{\partial}{\partial \beta} E[\ln Z_{\kappa, \beta, T}] &= E[\mu_{\kappa, \beta, T}(H_T(X))] \\ &= E \left[\mu_{\kappa, \beta, T} \left(\sum_{x \in \mathbb{Z}^d} \int_0^T dW_x(t) \delta_x(X(t)) \right) \right] \\ &= \sum_{x \in \mathbb{Z}^d} E_{\kappa} \left[E \left[\frac{e^{\beta H_T(X)}}{Z_{\kappa, \beta, T}} \int_0^T dW_x(t) \delta_x(X(t)) \right] \right]\end{aligned}$$

Integration by parts

$$\begin{aligned}
\frac{\partial}{\partial \beta} E[\ln Z_{\kappa, \beta, T}] &= E[\mu_{\kappa, \beta, T}(H_T(X))] \\
&= E \left[\mu_{\kappa, \beta, T} \left(\sum_{x \in \mathbb{Z}^d} \int_0^T dW_x(t) \delta_x(X(t)) \right) \right] \\
&= \sum_{x \in \mathbb{Z}^d} E_{\kappa} \left[E \left[\underbrace{\frac{e^{\beta H_T(X)}}{Z_{\kappa, \beta, T}}}_F \int_0^T dW_x(t) \delta_x(X(t)) \right] \right]
\end{aligned}$$

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&\stackrel{i.b.p.}{=} \sum_{x \in \mathbb{Z}^d} E_{\kappa} \left[E \int_0^T \left[D_{t,x} \frac{e^{\beta H_T(X)}}{Z_{\kappa, \beta, T}} \right] \delta_x(X(t)) \right] dt \\
&= \beta E_{\kappa} \left[E \sum_{x \in \mathbb{Z}^d} \int_0^T \left[\left(\delta_x(X_t) \frac{e^{\beta H_T(X)}}{Z_{\kappa, \beta, T}} - \frac{e^{\beta H_T(X)}}{Z_{\kappa, \beta, T}} \frac{E_{\kappa}[\delta_x(\tilde{X}_t) e^{\beta H_T(\tilde{X})}]}{Z_{\kappa, \beta, T}} \right) \delta_x(X_t) \right] dt \right] \\
&= \beta E \sum_{x \in \mathbb{Z}^d} \int_0^T \left[\mu_{\kappa, \beta, T}(\delta_x(X(t))) - \mu_{\kappa, \beta, T}(\delta_x(X(t)))^2 \right] dt = \beta T [1 - E[J_{\kappa, \beta, T}]].
\end{aligned}$$

Hence we got the key identity

$$\frac{\partial}{\partial \beta} E[\ln Z_{\kappa, \beta, T}] = \beta T [1 - E[J_{\kappa, \beta, T}]] \quad (*)$$

- By standard convexity arguments: Ψ is differentiable in β outside an at most countable set; At such point we have

$$\begin{aligned} \tilde{J}_{\kappa, \beta, \infty} &: \stackrel{\exists}{=} \lim_{T \rightarrow \infty} E[J_{\kappa, \beta, T}] \\ &= 1 - \beta^{-1} \frac{\partial}{\partial \beta} \Psi(\kappa, \beta) \end{aligned}$$

□

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From (i): results in a classical formulation

The critical value corresponds to the localization transition:

$$\beta^2/\kappa > \Upsilon_c \iff \tilde{I}_{\kappa,\beta,\infty} > 0$$

For fixed κ, β , define the **favourite site** $x^*(t)$ for the polymer at time t by

$$x^*(t) = \arg \max \left\{ E_\kappa [\exp\{\beta H_t(X)\}; X(t) = x] : x \in \mathbb{Z}^d \right\}$$

Theorem

$$\beta^2/\kappa > \Upsilon_c \iff \begin{cases} \exists C = C(\beta^2/\kappa) > 0 : \\ \liminf_{T \rightarrow \infty} \frac{1}{T} \int_0^T \mu_{\kappa,\beta,t}(X(t) = x^*(t)) dt \geq C \quad \text{a.s.} \end{cases}$$

Pointwise result, in the style of Carmona-Hu'02, FC-Shiga-Yoshida'03.

From (j)

- The critical value is when localization starts:

$$\Upsilon_c = \inf\{\beta^2/\kappa : \tilde{J}_{\kappa,\beta,\infty} > 0\}.$$

- Let

$$\mathcal{D}_\kappa = \{\beta > 0 : \exists \frac{\partial}{\partial \beta} \Psi(\kappa, \beta), \frac{\partial}{\partial \beta} \Psi(\kappa, \beta) < \beta\}$$

We have:

$$\beta \in \mathcal{D}_\kappa \implies \tilde{J}_{\kappa,\beta,\infty} > 0.$$

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Clearly, $\mathcal{D}_\kappa \subset [\kappa^{1/2}\Upsilon_c, \infty)$. **Conjecture:** $\mathcal{D}_\kappa = (\kappa^{1/2}\Upsilon_c, \infty)$.

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- Weaker statements on the (larger) set of increase of $\beta \mapsto \beta^2/2 - \Psi(\kappa, \beta)$: for such a β , $\exists \beta_n \rightarrow \beta$ such that

$$\lim_{T \rightarrow \infty} E \mu_{\kappa, \beta_n, T}^{\otimes 2} \left(\frac{1}{T} \int_0^T \mathbf{1}_{X(t) = \tilde{X}(t)} dt \right) > 0.$$

On the contrary, for β where the difference is locally constant, then the limit is 0 at β and around.

From (j): favourite path

For fixed κ, β , define the "favourite path" y_T^* for the polymer with time horizon T as the function

$$y_T^*(t) = \arg \max \left\{ E_\kappa (\exp\{\beta H_T(X)\}; X(t) = y); y \in \mathbb{Z}^d \right\}$$

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Theorem (FC-Cranston'11)

$$\beta \in \mathcal{D}_\kappa \implies \liminf_{T \rightarrow \infty} E \mu_{\kappa, \beta, T} \left(\frac{1}{T} \int_0^T \mathbf{1}\{X(t) = y_T^*(t)\} dt \right) \geq C(\beta, \kappa) > 0.$$

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Provides an information on the path itself.

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Similar to above, a weaker form holds on the set of increase of the difference.

On favourite attributes

What are those favourite attributes ? x^* and y_T^* :

Both depend on κ , β (also T) and on the environment \mathcal{W} .

Both have long jumps: in particular the "favourite path" is not a path...

y_T^* and x^* are equal at time $t = T$, but they are not related otherwise.

Fundamental difference: The mapping $t \mapsto x^*(t)$ has oscillations at those times t when there are many maximizers: the set of jump times then looks locally like the set of zeros of Brownian motion. In contrast, from differentiability below, we see that $t \mapsto y_T^*(t)$ has no oscillations, typically.

The favourite path is much smoother than the favourite end-point.

Proposition

(i) The function $t \mapsto E_\kappa [\exp\{\beta H_t(X)\} \delta_x(X(t))]$ is a.s. Hölder continuous of every order less than $1/2$.

(ii) The function $t \mapsto E_\kappa [\exp\{\beta H_T(X)\} \delta_x(X(t))]$ is almost surely of C^1 class on $[0, T]$.

□ (i) $Z(t, x) = E_\kappa [\exp\{\beta H_t(X)\} \delta_x(X(t))]$ solves the stochastic heat equation

$$dZ(t, x) = \kappa \Delta Z dt + \beta Z \circ dW_x(t).$$

It has the regularity of Brownian motion.

(ii) On the other hand,

$$E_\kappa [\exp\{\beta H_T(X)\} \delta_x(X(t))] \times \exp -\beta W_z(T)$$

is (continuously) differentiable at t when $z = x$. □

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As $\beta^2/\kappa \rightarrow \infty$:

- Recall that $\Psi(\kappa, \beta) \sim \frac{\alpha^2 \beta^2}{4 \ln(\beta^2/\kappa)} \ll \beta^2/2$, and moreover, that

$$\frac{\beta^2}{2} - \Psi(\kappa, \beta) = \frac{\beta^2}{2} \tilde{I}_{\kappa, \beta, \infty}, \quad \text{and} \quad \frac{\partial}{\partial \beta} \left(\frac{\beta^2}{2} - \Psi(\kappa, \beta) \right) = \beta \tilde{J}_{\kappa, \beta, \infty}.$$

- This implies that both $\tilde{I}_{\kappa, \beta, \infty}$ and $\tilde{J}_{\kappa, \beta, \infty}$ tend to 1 at rate $\mathcal{O}(1/\ln(\beta^2/\kappa))$. Since 1 is the maximal value, the localization is **complete**.
- Similarly,

$$\limsup_{T \rightarrow \infty} \frac{1}{T} \int_0^T \mu_{\kappa, \beta, t}(X(t) \neq x^*(t)) dt = \mathcal{O}(1/\ln(\beta^2/\kappa))$$

$$\limsup_{T \rightarrow \infty} E \mu_{\kappa, \beta, T} \left(\frac{1}{T} \int_0^T \mathbf{1}\{X(t) \neq y_T^*(t)\} dt \right) = \mathcal{O}(1/\ln(\beta^2/\kappa))$$

(No need to restrict $\beta \in \mathcal{D}_{\kappa}$.)

Conclusion

- When the discrepancy between quenched and annealed free energy increases, the random path sticks to the favourite one for a significant fraction of the time.
The fraction of times grows to 1 as diffusivity \times temperature² vanishes.
- Path localization and complete localization also hold in other models. We proved that for Parabolic Anderson model, but this works as soon as the environment is **Gaussian**.

Another example: Brownian polymer in **Poissonian** medium (FC-Yoshida). Poisson variables have a nice integration by parts formula.