

## The Setup:

For  $\omega_j \in \mathbb{Z}^d$ ,  $j \in \mathbb{N}$ , let the sequence  $\omega = (\omega_1, \omega_2, \omega_3, \dots)$  represent the path of a  $d$ -dimensional directed random polymer. We construct the  $n$ -th step measure on this path as follows,

$$\mu_n(\{\omega\}) = \frac{1}{Z_n} \exp\left(\beta \sum_{j=1}^n \eta(j, \omega_j)\right) P(\{\omega\})$$

where:

- $P$  - the  $\mathbb{Z}^d$  simple-symmetric random walk measure of a path  $\omega$ .
- $\beta$  - a parameter whose physical interpretation is inverse temperature.
- $(\eta)_{\mathbb{N} \times \mathbb{Z}^d}$  - a family of iid rvs we call the random environment.
- $Z_n$  - a normalisation factor we call the the partition function.

Then  $\mu_n$  reweights the measure  $P$ , giving more weight to those paths which often encounter a positive environment ( $\eta(j, \omega_j) > 0$ ) and less to those which often encounter a negative environment ( $\eta(j, \omega_j) < 0$ ). Hence the model attempts to capture the idea that a polymer will grow towards an environment which is favourable. (fig. 1)

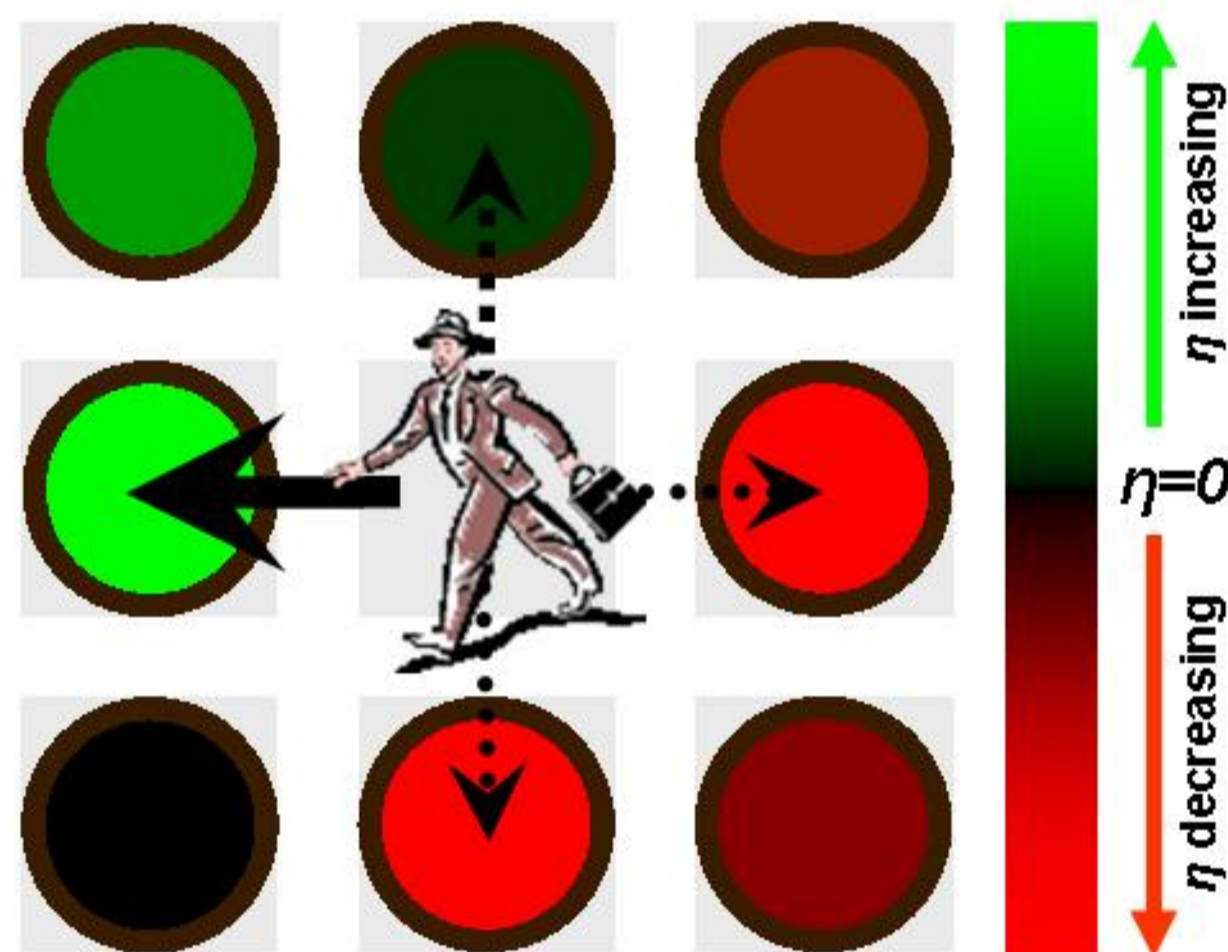


fig. 1

Under the measure  $\mu_n$  the walker's next step is more likely to be left, where the environment is +ve and less likely to be right or down.

## The Random Environment:

The random environment  $(\eta)_{\mathbb{N} \times \mathbb{Z}^d}$  is a family of real valued, non constant iid rvs. We define  $Q$  to be the measure on members of this family. Furthermore,  $\eta$  has a finite moment generating function. Formally,  $Q(e^{\beta\eta}) < \infty$  for all  $\beta \in \mathbb{R}$ . Additionally we will introduce notation for the log of the moment generating function as a function of the inverse temperature, which we will denote by  $\lambda_\eta(\beta)$ . Formally,  $\lambda_\eta(\beta) = \log(Q(e^{\beta\eta}))$ .

## Existence of a Phase Transition:

We study the normalised partition function,  $W_n$ , defined as follows,

$$W_n = Z_n e^{-n\lambda_\eta(\beta)}$$

Notice  $Q(W_n) = 1$  for all  $n$  and  $W_n$  is a martingale wrt filtration  $(\mathcal{G}_n)$ , where  $\mathcal{G}_n = \sigma\{\eta(j, x); j \leq n, x \in \mathbb{Z}^d\}$ .

It can be shown analytically that  $W_\infty := \lim_{n \rightarrow \infty} W_n$  converges  $Q$ -as and either  $W_\infty > 0$ ,  $Q$ -as or  $W_\infty = 0$ ,  $Q$ -as.

Further more this transition is monotonic in  $\beta$  and hence the model shows a phase transition namely,

$$\begin{aligned} W_\infty &> 0 \text{ } Q\text{-as for } \beta < \beta_c \\ W_\infty &= 0 \text{ } Q\text{-as for } \beta > \beta_c \end{aligned}$$

where  $\beta_c$  is the critical point at which phase transition occurs.

$W_\infty > 0$   $Q$ -as corresponds to a diffusive behaviour. Formally

$$\lim_{n \rightarrow \infty} \frac{\mu_n(|\omega_n|^2)}{n} = 1, \text{ } Q\text{-as}$$

$W_\infty = 0$   $Q$ -as corresponds to a regime of strong disorder. (fig. 2)

For  $d = 1, 2$  it has been shown that  $\beta_c = 0$ .

For  $d \geq 3$  the following analytical bounds exist for  $\beta_c$ [1],

- $\lambda_\eta(2\beta_c) - 2\lambda_\eta(\beta_c) \geq \log \frac{1}{\pi_d}$ , (lower)
- $\beta_c \lambda'_\eta(\beta_c) - \lambda_\eta(\beta_c) \leq \log 2d$ , (upper)

where  $\pi_d$  is the probability that a  $d$ -dimensional random walk returns to its starting point.

Recent work has shown that the lower bound is not tight for  $d \geq 4$ . A major objective of this project is to investigate numerically whether the lower bound is tight for  $d = 3$ .

### Strong disorder:

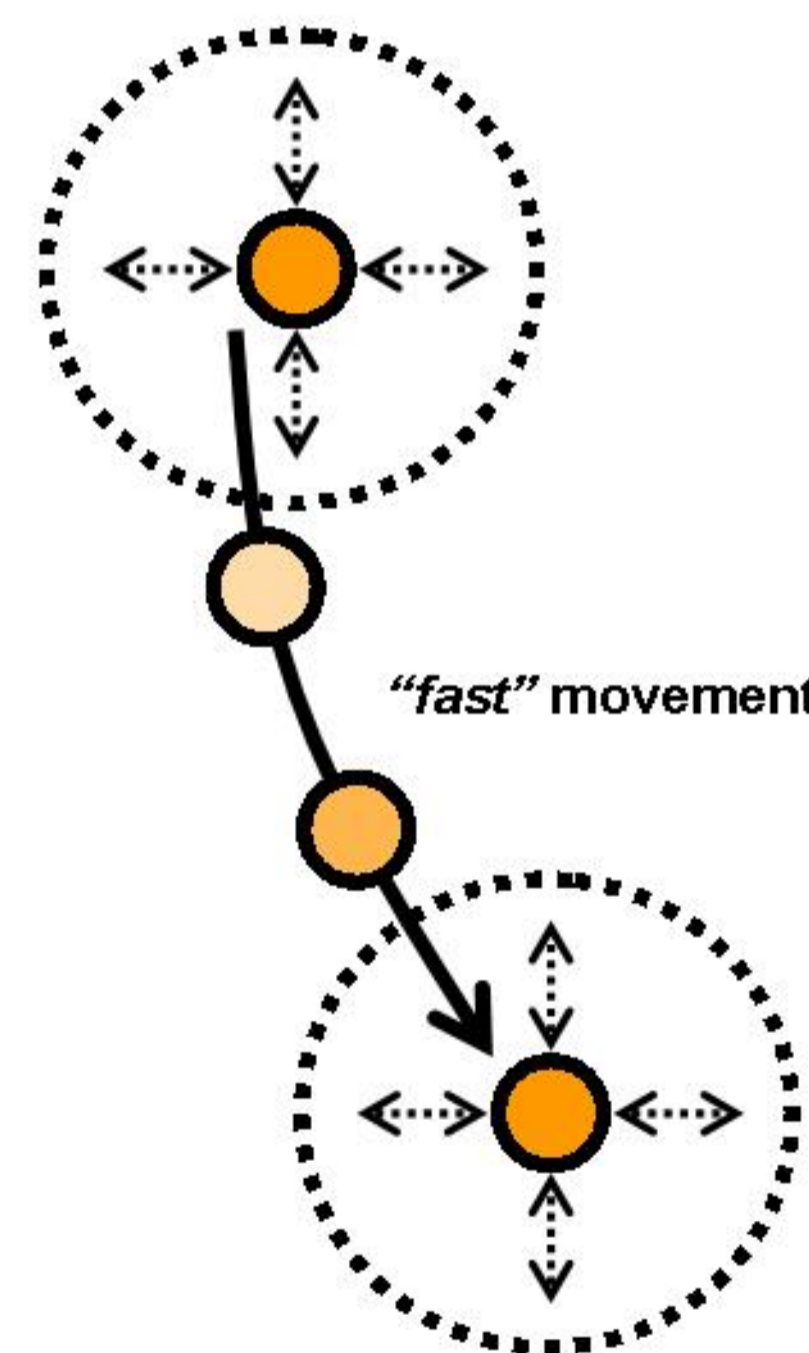


fig. 2

Strong disorder: Favourable paths stay localised at a favourite site and exhibit occasional fast movement through a +ve environment.

## Calculation:

We will attempt to calculate  $W_n$  for  $d = 3$  and a Gaussian random environment,  $\eta \sim N(0, 1)$ . For large  $n$  we hope to observe the phase transition and estimate  $\beta_c$ .

We will exploit the following recursion relation,

$$W_n(x) = \frac{e^{\beta\eta(n,x) - \lambda_\eta(\beta)}}{2d} \sum_{\delta \in \Delta} W_{n-1}(x + \delta)$$

with  $W_0(x) = 1$  if  $x = 0$  and  $W_0(x) = 0$  if  $x \neq 0$ , where

$$W_n(x) = P\left(e^{\sum_{j=1}^n (\beta\eta(j, \omega_j) - \lambda_\eta(\beta))} \mathbb{I}_{\{\omega_n = x\}}\right)$$

and

$$\Delta = \{\delta \in \mathbb{Z}^d : \|\delta\|_1 = 1\}^\dagger$$

We calculate  $W_n$  by summing over lattice points,

$$W_n = \sum_{x \in \mathbb{Z}^d} W_n(x)$$

remembering that  $W_n(x) = 0$  when  $\|x\|_1 > n$  because a random walker could not have traveled more than  $n$  steps away from the origin.

Another aim of the project is to investigate a possible scaling law of of the form  $W_n \sim e^{-nf(\beta - \beta_c)}$  for  $\beta > \beta_c$ , where the function  $f$  will be some scaling exponent.

## Preliminary Findings:

We have written c code to compute  $W_n$  up to  $n = 290$  for several realisations of the Gaussian random environment. Figure 3 shows  $W_{290}$  plotted against  $\beta$  for one such environment. The phase transition can be clearly observed and the preliminary findings indicate it is plausible that the analytical lower bound for  $d = 3$  and  $\eta \sim N(0, 1)$  of  $\beta_c \geq 1.03$  may not be tight. It will be necessary to compute  $W_n$  to larger  $n$  to confirm this.

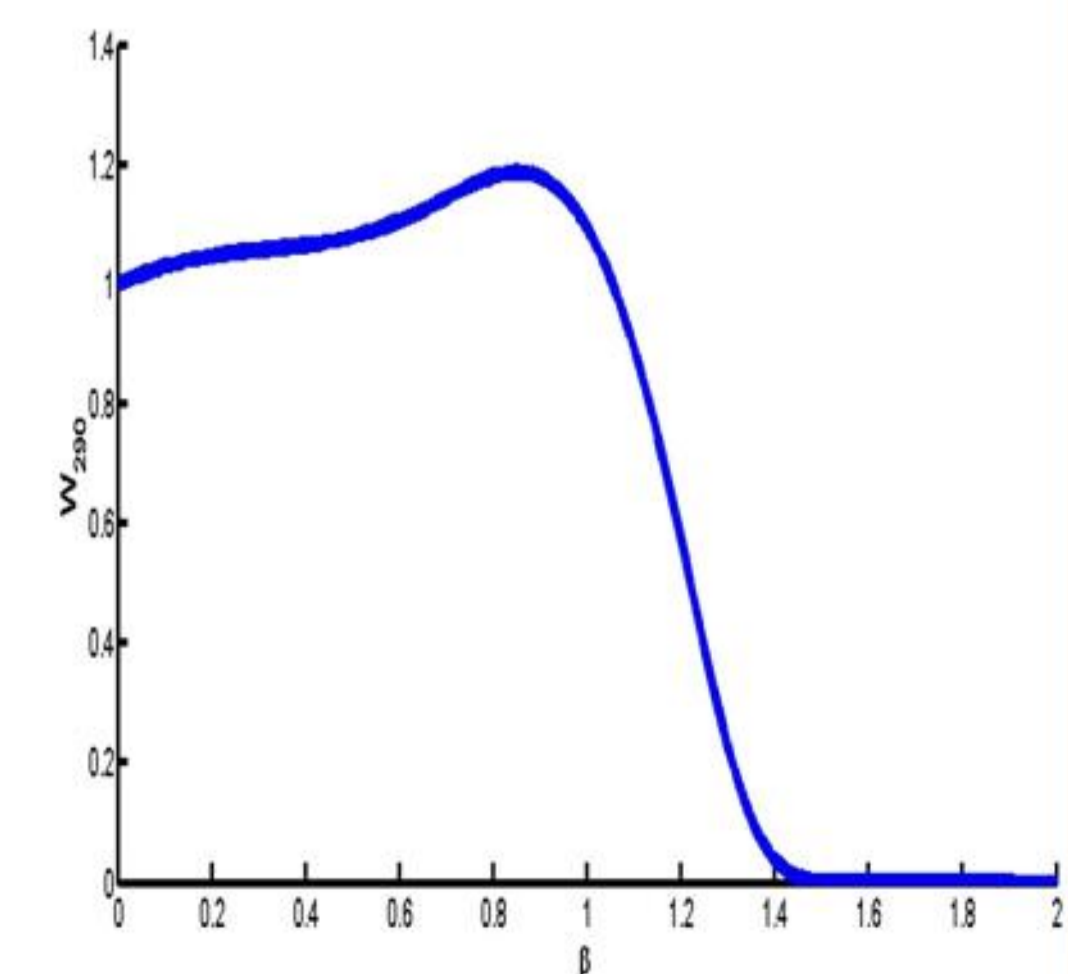


fig. 3