

Some Results and Conjectures for Tree Polymers under Weak/Strong Disorder

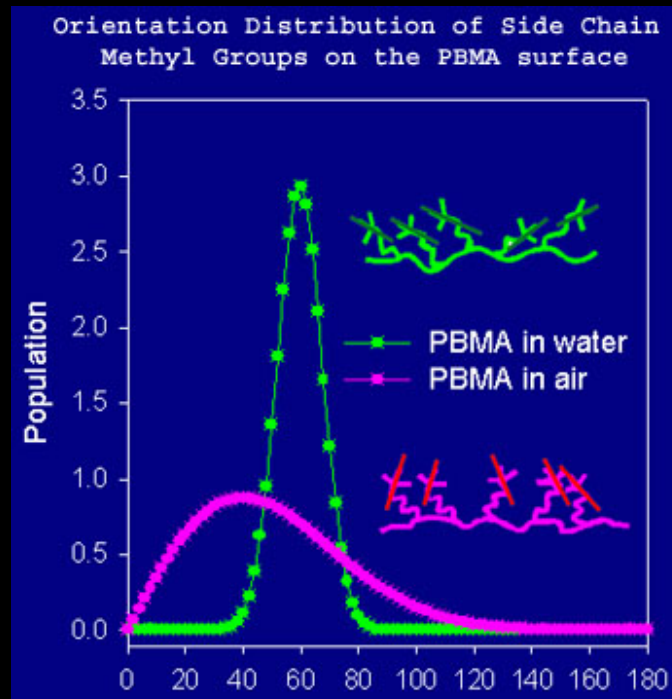
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3RD FRONTIER PROBABILITY DAYS

SALT LAKE CITY, UT MARCH 7-22, 2011

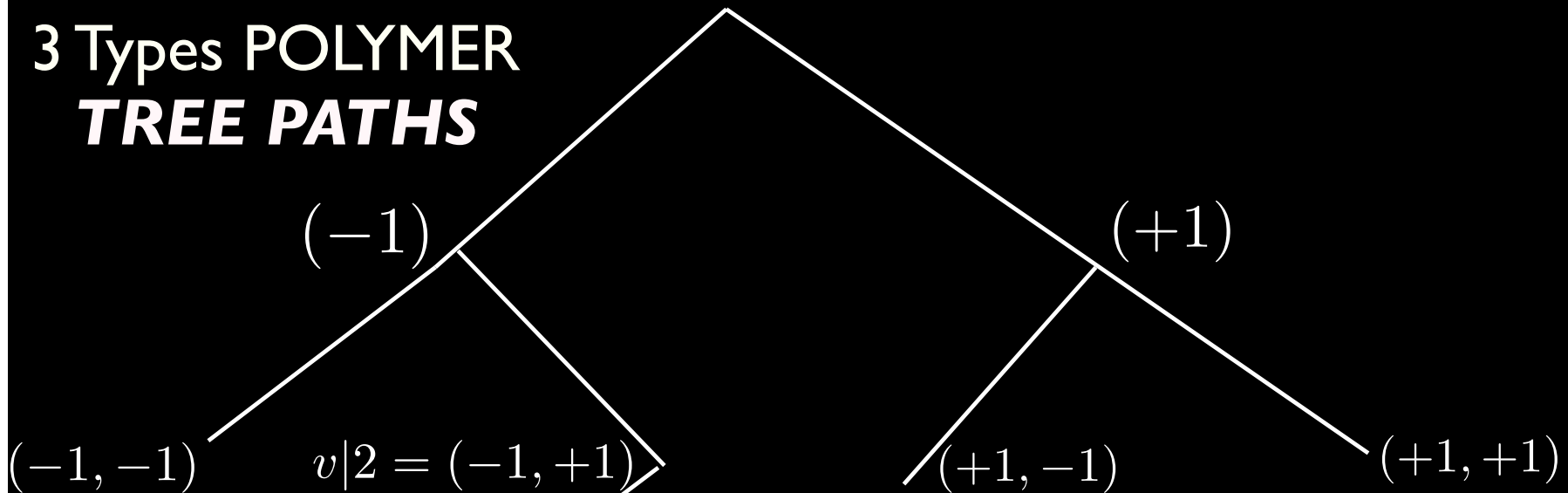


Based on joint work with Stanley C. Williams, Torrey Johnson

[W&W,2010] Tree polymers and multiplicative cascades,
in: *Fractals and Related Fields*, ed Julien Barrel, Birkhauser

[J&W,2011] Tree polymers in the infinite volume limit at critical
strong disorder, (PREPRINT)

3 Types POLYMER TREE PATHS



(FINITE TREE PATH): $v = (S|n) = (s_1, s_2, \dots, s_n) \in \{-1, +1\}^n$

(INFINITE TREE PATH): $S = (s_1, s_2, \dots) \in \partial T = \{-1, +1\}^\infty$

(POLYGONAL TREE PATH): $n \rightarrow (S)_n = \sum_{j=1}^n s_j$

Path Probability
Weights

IID POSITIVE
 $EX = 1$

$X(-1)$ (Ω, \mathcal{F}, P)

$X(+1)$

$X(-1, -1)$

$X(-1, +1)$

$X(+1, -1)$

$X(+1, +1)$

$X(v)$

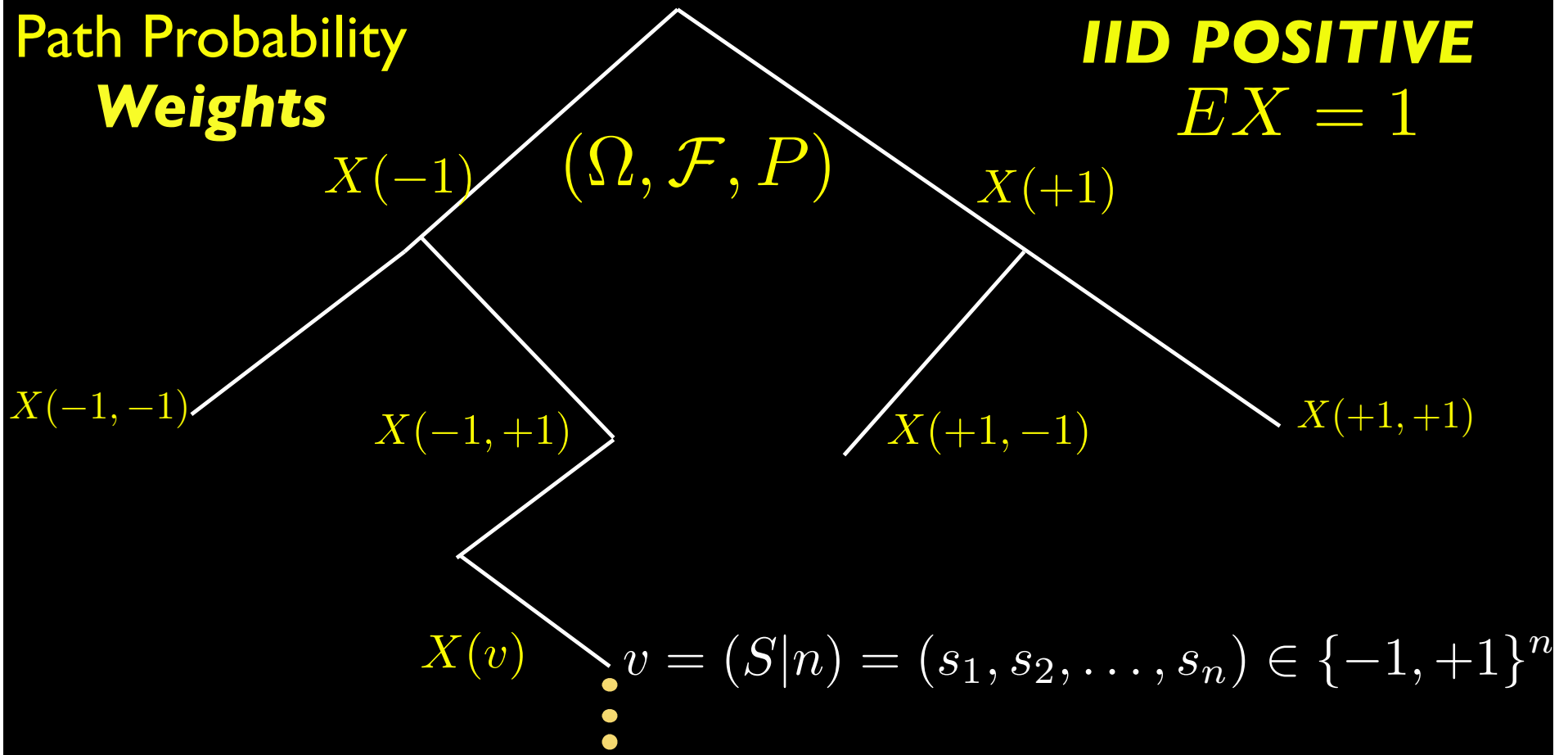
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$$n \rightarrow (S)_n = \sum_{j=1}^n s_j$$

**Path Probability
Weights**

**IID POSITIVE
 $EX = 1$**



(PATH PROBABILITIES): $S = (s_1, s_2, \dots) \in \partial T = \{-1, +1\}^\infty$

$$prob_n(\Delta_n(S)) = Z_n^{-1} \prod_{j=1}^n X(S|j) 2^{-n}$$

$$\Delta_n(S) = \{s \in \partial T : s|n = S|n\}$$

Normalized Haar measure: $\lambda(ds)$ on $\{-1, 1\}^\infty$

$$\frac{dprob_n}{d\lambda}(S) = Z_n^{-1} \prod_{j=1}^n X(S|j)$$

$X(v)$

$$S = (s_1, s_2, \dots) \in \partial T = \{-1, +1\}^\infty \quad (S)_n = \sum_{j=1}^n s_j$$

SIMPLE
SYMMETRIC RW $X \equiv 1 \Rightarrow prob_n(ds) = \lambda(ds)$

PROBABILITY LAWS OF $n \rightarrow (S)_n$ WELL UNDERSTOOD !

Bolthausen's Disorder Parameterization --

Weak Disorder: $Z_\infty = \lim_{n \rightarrow \infty} Z_n > 0 \text{ a.s.}$

Strong Disorder: $Z_\infty = \lim_{n \rightarrow \infty} Z_n = 0 \text{ a.s.}$

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Theorem (Kahane-Peyriere '76, Biggins '76, W., Williams, '94)

Weak Disorder if and only if $EX \ln X < \ln 2$

Strong Disorder if and only if $EX \ln X \geq \ln 2$

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THEOREM (Kahane-Peyriere 1976 "Live/Die" Criteria)

Weak Disorder if and only if $EX \ln X < \ln 2$

Strong Disorder if and only if $EX \ln X \geq \ln 2$

(Disorder Strength vs Branching Rate)

TYPICAL POLYMER PROBLEMS FOR:

$$(S)_n = \sum_{j=1}^n s_j$$

A. a.s. Non-ballistic ? (a.s. LLN---Weak/Strong ?)

B. Infinite Volume Limit ? $prob_{\infty}(ds) = \lim_{n \rightarrow \infty} prob_n(ds) \text{ a.s.}?$

C. a.s. Diffusive (sub/super) ? (a.s. CLT---Universal/Variance ?)

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$$prob_n(\{s \in \partial T : \frac{(s)_n}{\sqrt{n}} \leq x\}) \rightarrow \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{1}{2}z^2} dz \text{ a.s. ?}$$

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SIMPLE SYMMETRIC RW BENCHMARK: $prob_n(ds) = \lambda(ds)$

ANSWER A: Always non-ballistic (in a weak sense)
regardless of disorder type.

$$E_{prob_n} \left| \frac{(S)_n}{n} \right| \rightarrow 0 \quad a.s. \quad \text{as } n \rightarrow \infty$$

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NOTE: a..s. Strong Law requires answer to Question B
Existence of $prob_\infty$

e.g. $prob_\infty(s \in \partial T : \frac{(s)_n}{n} \rightarrow 0) = 1 \quad a.s.$

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e.g., $prob_\infty(s \in \partial T : \frac{(s)_n}{n} \rightarrow 0) = 1 \quad a.s.$

Similarly, law of iterated logarithm, and other a.s., a.s. laws

ANSWER B. SPECIAL CASES

PROPOSITION B2 (Johnson, W.) In the case of weak disorder

a. $prob_{\infty}(ds) = \lim_{n \rightarrow \infty} prob_n(ds)$ exists a..s.

In the case of critical strong disorder, for each finite $F \subset N$

b. $\widehat{prob}_{\infty}(F) = \lim_{n \rightarrow \infty} \widehat{prob}_n(F)$ in probability

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WD: Martingale Convergence Theorem

$$M_n(f) = Z_n \int_{\partial T} f(s) prob_n(ds), \quad n \geq 1.$$

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SD: Seneta-Heyde scaling & Derivative Martingale for BRW.
Aidekon and Shi (2011), Biggins and Kyprianou (2004)

If μ is a probability on group ∂T , $\hat{\mu}(F) = \int_{\partial T} \prod_{j \in F} (s_j) \mu(ds)$
is Fourier transform.

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Conventional Wisdom: General Nonexistence for SD.

Remark: Size-Biased Existence for SD (W. & Williams 1996)

Long-chain Diffusivity

THEOREM (Bolthausen, 1989)

Assume lognormally distributed weights: $X = e^{\beta W}$

Then for $\beta < \sqrt{2} < \beta_c = \sqrt{2 \ln 2}$ one has the diffusive scaling

$$\int_{\partial T} \left(\frac{(S)_n}{\sqrt{n}} \right)^2 \text{prob}_n(dS) \rightarrow 1 \quad a.s.$$

Also a.s. asymptotically normal distribution

$$M_n\left(\frac{r}{\sqrt{n}}\right) \equiv \int_{\partial T} e^{\frac{(S)_n}{\sqrt{n}} r} \text{prob}_n(dS) \rightarrow e^{\frac{r^2}{2}} \quad -\delta < r < \delta$$

NOTE: The full weak disorder regime is $\beta < \beta_c = \sqrt{2 \ln 2}$
i.e., since weak disorder is:

$$\frac{\beta^2}{2} = E \frac{X}{EX} \ln \frac{X}{EX} < \ln 2$$

PROPOSITION C1 (Weak Disorder) W.&Williams (2010)

Under weak disorder the following limit exists a.s.:

$$\lim_{n \rightarrow \infty} \frac{\ln M_n(r)}{n} = \ln \cosh(r) \quad a.s.$$

THEOREM C1. (W&W 2010): Assume $EX^{1+\epsilon} < \infty$

For the tree polymer model **weak disorder** implies **diffusive scaling** and **asymptotic normality** almost surely.

That is, a.s.

$$M_n\left(\frac{r}{\sqrt{n}}\right) \equiv \int_{\partial T} e^{\frac{(S)_n}{\sqrt{n}} r} \text{prob}_n(dS) \rightarrow e^{\frac{r^2}{2}} \quad -\delta < r < \delta$$

REMARK: Comets and Yoshida (2006), AoP, Proved CLT $\frac{1}{d}I_d$ for $d+1$ -dimensional lattice polymers, $d \geq 3$. Cov $\Sigma = \frac{1}{d}I_d$.

Proposition C3. (Strong Disorder) W.&Williams (2010)

Under strong disorder the following limit exists a.s. :

$$F(r) = \lim_{n \rightarrow \infty} \frac{\ln M_n(r)}{n \ln EX^{h(r)} + \ln \left(p_r^{h(r)}(+)+ p_r^{h(r)}(-) \right)}$$
$$= \ln \cosh(r) + \frac{\ln \cosh(r)}{h(r)}$$

where $h = h(r)$ solves

$$E \left\{ \frac{X^h}{EX^h} \ln \frac{X^h}{EX^h} \right\} = \epsilon \left(\bar{p}_r^h(+), \bar{p}_r^h(-) \right)$$

$$\epsilon(p, q) = -p \ln p - q \ln q ; \quad \bar{p}_r^h(\pm) = \frac{p_r^h(\pm)}{p_r^h(+)+ p_r^h(-)}$$

SPECIAL CASE: $X = e^{\beta W - \frac{\beta^2}{2}}$

$$F(r) = r \tanh(rh(r)) + \beta^2 h(r) - \beta_c \beta$$

where

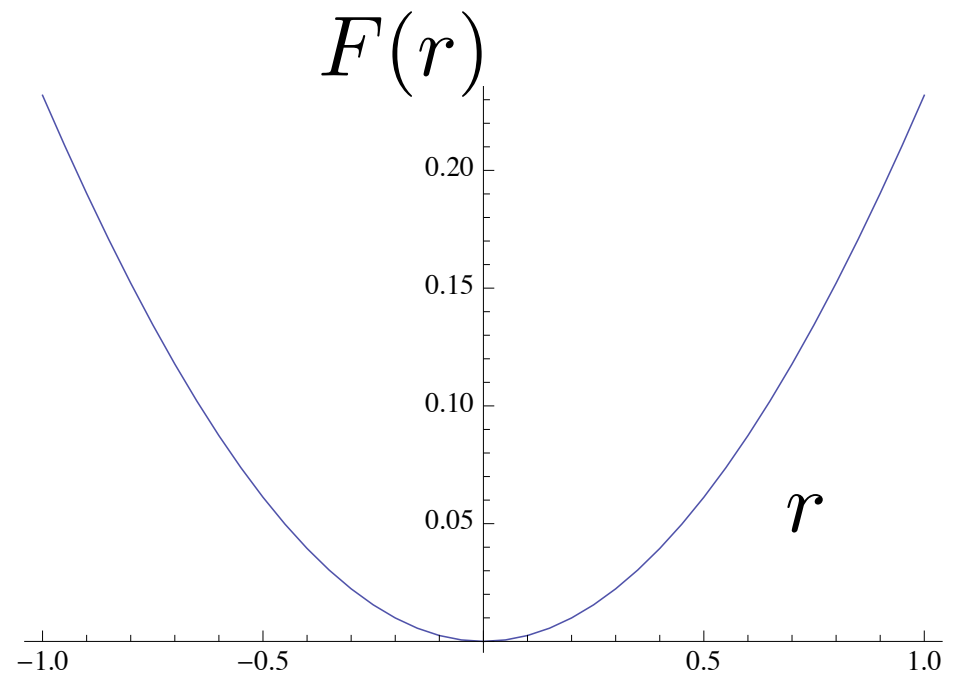
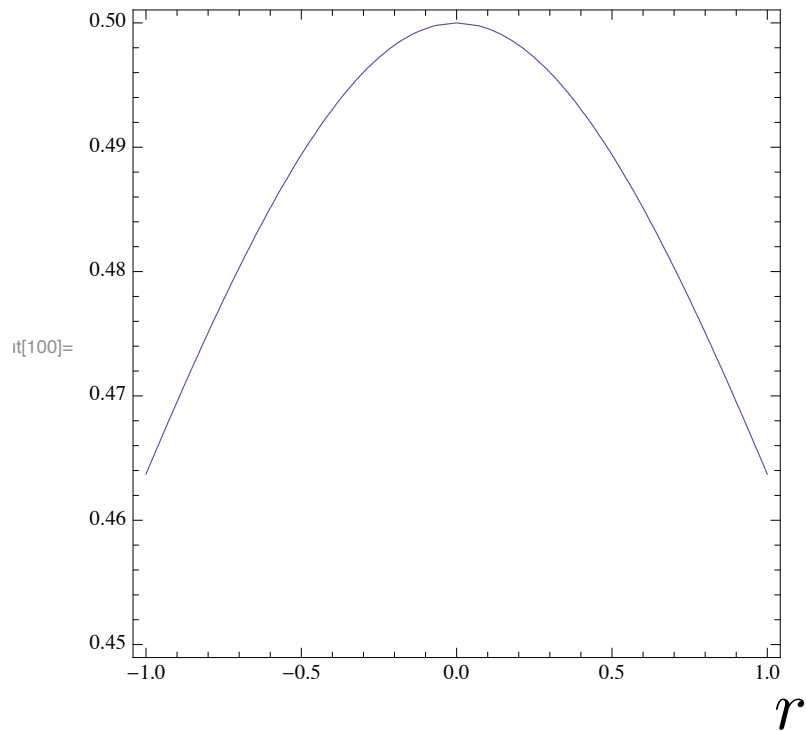
$$\beta^2 h^2(r) + 2rh(r) \tanh(rh(r)) - 2 \ln \cosh(rh(r)) = \beta_c^2$$

$$F(r) = r \tanh(rh(r)) + \beta^2 h(r) - \beta\beta_c$$

$$\beta^2 h^2(r) + 2rh(r) \tanh(rh(r)) - 2 \ln \cosh(rh(r)) = \beta_c^2$$

$$\beta = 2\sqrt{2 \ln 2} > \beta_c$$

$h(r)$



CONJECTURE (J., W. 2011) Diffusive scaling and variance
 $\sigma^2 = \sigma^2(\beta)$ under strong disorder

where

$$\sigma^2(\beta) = \frac{2\beta\beta_c - \beta_c^2}{\beta^2}, \quad \beta \geq \beta_c$$

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--SKETCHES OF PROOFS--

SIZE BIAS TOOL: (W., Williams 1994)

Define the **size-bias** probability on $\Omega \times \partial T$

$$Q(d\omega \times ds) = P_s(d\omega) \lambda(ds)$$

where on $\sigma(X_v : |v| \leq n)$

$$P_s(d\omega) = \prod_{j=1}^n X_{s|j}(\omega) P(d\omega)$$

SIZE BIAS TOOL: (W., Williams 1996 TAMS)

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$$P_\omega(ds) = \text{prob}_n(\omega, ds)$$

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$$Q_\Omega(d\omega) \equiv Q \circ \pi_\Omega^{-1}(d\omega) = Z_n(\omega)P(d\omega)$$

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$$Q_\Omega(d\omega) \equiv Q \circ \pi_\Omega^{-1}(d\omega) = Z_n(\omega) P(d\omega)$$

$$Q_{\partial T}(ds) \equiv Q \circ \pi_{\partial T}^{-1}(ds) = \lambda(ds)$$

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Define the **size-bias** probability on $\Omega \times \partial T$

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$$P_s(d\omega) = \prod_{j=1}^n X_{s|j}(\omega) P(d\omega) \quad \text{on } \sigma(X_v : |v| \leq n)$$

LEMMA (W,W, '94): On $\mathcal{F} = \sigma(X_v : v \in T)$

Weak disorder : $Q_\Omega(d\omega) \ll P(d\omega)$

Strong disorder: $Q_\Omega(d\omega) \perp P(d\omega)$

PROOF IDEAS:

WD: $\frac{dQ_{\Omega}}{dP}(\omega) = Z_{\infty}(\omega) < \infty$

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$$\frac{dQ_\Omega}{dP}(\omega) = Z_\infty(\omega) < \infty$$

SD:

$$\begin{aligned} Z_n &\geq \prod_{j=1}^n X_{s|j} 2^{-n} = \exp\left\{n\left(\frac{1}{n} \sum_{j=1}^n X_{s|j} - \ln 2\right)\right\} \\ &\sim \exp\left\{n(E X \ln X - \ln 2)\right\} \\ &\rightarrow \infty \quad \text{if} \quad EX \ln X > \ln 2 \end{aligned}$$

But, positive martingale property implies

$$P(Z_\infty < \infty) = 1$$

i.e. $Q_\Omega(d\omega) \perp P(d\omega)$

Calculate

$$F(r) = \lim_{n \rightarrow \infty} \frac{\ln M_n(r)}{n}$$

WD:

$$\frac{Z_n M_n(r)}{\cosh^n(r)} \quad -\delta < r < \delta \quad \text{Kahane's T-Martingale}$$

SD:

Via SIZE-BIASED Borel-Cantelli limsup and liminf calculations

OR

Via a vector cascade T-martingale coding (Displ., Weight)

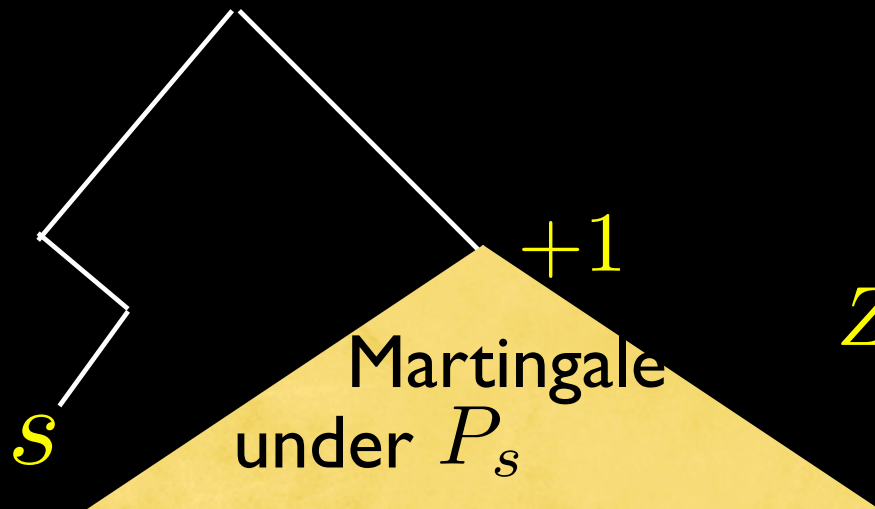
PROPOSITION B2 (W., Williams 1996 TAMS)

In the case of strong disorder there is a random path $\tau = \tau(\omega)$ such that

$$\text{prob}_\infty(ds) = \delta_\tau(ds) \quad Q_\Omega - a.s.$$

PROOF IDEA:

Strong disorder $Q_\Omega - a.s.$ makes $Z_\infty = \infty$



$$\Rightarrow Z_\infty^+(\omega) < \infty$$

1. Fix a path s say $s_1 = -1$

2. So now fix $\omega \in [Z_\infty = \infty]$

$$Z_\infty^+(\omega) < \infty \Rightarrow \tau_1(\omega) = -1$$

Theorem C1 (Proof Ideas)

For the tree polymer model weak disorder implies diffusive scaling and asymptotic normality almost surely.

That is

$$\lim_{n \rightarrow \infty} M_n\left(\frac{r}{\sqrt{n}}\right) = e^{\frac{r^2}{2}} \quad |r| < \delta \quad \text{a.s.}$$

Sketch of Proof:

$$EZ_n M_n(r) = \cosh^n(r)$$

$$\frac{Z_n M_n(r)}{\cosh^n(r)} \quad -\delta < r < \delta \quad \text{Kahane's T-MARTINGALE}$$

$$\frac{d}{dr} \frac{Z_n M_n(r)}{\cosh^n(r)} \quad -\delta < r < \delta \quad \text{(Signed) T-MARTINGALE}$$

$$m'_n(r) = \frac{d}{dr} \frac{Z_n M_n(r)}{\cosh^n(r)} = \sum_{j=1}^n m_{n,j}(r)$$

$$E|m'_n(r)|^q \leq C \sum_{j=1}^{\infty} \left(\frac{EX^q}{2^{q-1}} \right)^j$$

WEAK DISORDER \implies

$$\exists 1 < q < 2 \text{ such that } \frac{EX^q}{2^{q-1}} < 1$$

$$\implies \frac{Z_n M_n(r)}{\cosh^n(r)} \rightarrow Y(r) \quad |r| < \delta$$

$$M_n\left(\frac{r}{\sqrt{n}}\right) = Z_n^{-1} \frac{Z_n M_n\left(\frac{r}{\sqrt{n}}\right)}{\cosh^n\left(\frac{r}{\sqrt{n}}\right)} \cosh^n\left(\frac{r}{\sqrt{n}}\right) \rightarrow Z_{\infty}^{-1} Y(0) e^{\frac{r^2}{2}} = e^{\frac{r^2}{2}}$$



THANK YOU