

Thinning Invariant Sequences

NSF-CBMS Conference at UAHuntsville

Small Deviation Probabilities: Theory and Applications

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June 7, 2012

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Sherrington-Kirkpatrick model:

For each $N \in \mathbb{N}$, $H_N : \{+1, -1\}^N \rightarrow \mathbb{R}$ is a random function

$$H_N(\sigma) = \sum_{i,j=1}^N \frac{J_{ij}}{\sqrt{2N}} \sigma_i \sigma_j$$

where $\sigma = (\sigma_1, \dots, \sigma_N)$ and $(J_{ij})_{i,j=1}^N$ are i.i.d., $\mathcal{N}(0, 1)$.

Two key ideas.

- ▶ Gaussian inequalities: Slepian's lemma.
- ▶ Exchangeability.

1/10 Bernoulli- p thinning

Let $\mathcal{X} \stackrel{\text{def}}{=} \{0, 1\}$. Let $p \in (0, 1]$.

- Suppose

$$(X_1, X_2, \dots) \sim \mu \in \mathcal{M}(\mathcal{X}^{\mathbb{N}})$$

- Independently,

$$(B_1, B_2, \dots) \sim \text{i.i.d. Bernoulli}(p)$$

- Let $K_1 < K_2 < \dots =$ indices k s.t. $B_k = 1$
- $\theta_p(\mu) \in \mathcal{M}(\mathcal{X}^{\mathbb{N}}) =$ marginal distribution $(X_{K_1}, X_{K_2}, \dots)$.

	X_1	X_2	X_3	X_4	X_5	\dots
$B_k:$	1	1	0	1	0	\dots
	\downarrow	\downarrow	\swarrow			
	X_{K_1}	X_{K_2}	X_{K_3}	\dots		

$$\begin{array}{cccccc}
 & X_1 & X_2 & X_3 & X_4 & X_5 & \cdots \\
 B_k: & 1 & 1 & 0 & 1 & 0 & \cdots \\
 & \downarrow & \downarrow & \swarrow & & & \\
 & Y_1 & Y_2 & Y_3 & \cdots & &
 \end{array}$$

Definition: *Thinning invariance:*

$$\theta_p(\mu) = \mu \quad \text{for all } p \in (0, 1].$$

Example: Let X_1, X_2, \dots be i.i.d.

Example 2. Let X_1, X_2, \dots be conditionally i.i.d.

E.g. $U \sim \text{Unif}([0, 1])$.

Conditional on U : $X_1, X_2, \dots \sim \text{i.i.d. Bernoulli}(U)$.

Definition: *Spreadability:*

- ▶ \forall non-random $k_1 < k_2 < \dots$,
- ▶ given $(X_1, X_2, \dots) \sim \mu$,

$$(X_{k_1}, X_{k_2}, \dots) \sim \mu.$$

Ryll-Nardzewski showed that spreadability equals exchangeability.

Definition: *Exchangeability*

- ▶ \forall “finite” permutation $\pi : \mathbb{N} \rightarrow \mathbb{N}$,
- ▶ given $(X_1, X_2, \dots) \sim \mu$,

$$(X_{\pi(1)}, X_{\pi(2)}, \dots) \sim \mu.$$

References: for thinning invariant point processes.

Olav Kallenberg, “Random measures.”

Mathes, Kerstan and Mecke, “Infinitely divisible point processes.”

$\mathcal{M}(\mathcal{X}) =$ all Borel measures μ on \mathcal{X} .

$\mathcal{M}(\mathcal{M}(\mathcal{X})) =$ all Borel measures Q on $\mathcal{M}(\mathcal{X})$.

For $Q \in \mathcal{M}(\mathcal{M}(\mathcal{X}))$:

- ▶ $\mu \sim Q$
- ▶ conditional on μ : $(X_1, X_2, \dots) \sim \text{i.i.d.}, \mu$
- ▶ $\mu_Q =$ marginal distribution of (X_1, X_2, \dots)

De Finetti's Theorem:

$$\{\mu \in \mathcal{M}(\mathcal{X}^{\mathbb{N}}) : \text{exchangeable}\} = \{\mu_Q : Q \in \mathcal{M}(\mathcal{M}(\mathcal{X}))\}$$

5/10 Counterexamples to exchangeability

1. Choose $0 < \xi_1 < \xi_2 < \dots \sim$ Poisson point process.
Independently, choose $U \sim \text{Unif}([0, 2])$.

$$X_n = \mathbf{1}_{\mathbb{Z}_+[0,1)}(\ln \xi_n + U) \quad \text{for } n \in \mathbb{N},$$



2. Choose $0 < \xi_1 < \xi_2 < \dots \sim$ Poisson point process.
Independently, let $W : \mathbb{R} \rightarrow [0, 1] \sim$ reflected Brownian motion.

$$U_n = W(\ln \xi_n) \in [0, 1] \quad \text{for each } n \in \mathbb{N}.$$

Conditional on (U_1, U_2, \dots) : X_1, X_2, \dots independent

$$X_n \sim \text{Bernoulli}(U_n).$$

- Note $\{(\ln \xi_n, X_n)\}_{n \in \mathbb{N}}$ is a mixed Poisson process on $\mathbb{R} \times \mathcal{X}$.

- ▶ Let $\mathcal{M}_{\text{Leb}}(\mathbb{R} \times \mathcal{X})$ denote the set of all Borel measures ρ on $\mathbb{R} \times \mathcal{X}$ such that $\rho(\cdot \times \mathcal{X}) = \text{Lebesgue measure}$.
- ▶ Let $\mathcal{M}_{\text{I}}(\mathcal{M}_{\text{Leb}}(\mathbb{R} \times \mathcal{X}))$ denote the set of all Borel probability measures Q on $\mathcal{M}_{\text{Leb}}(\mathbb{R} \times \mathcal{X})$ such that

$$Q(\{\rho : \rho \circ \tau_t^{-1} \in \cdot\}) = Q,$$

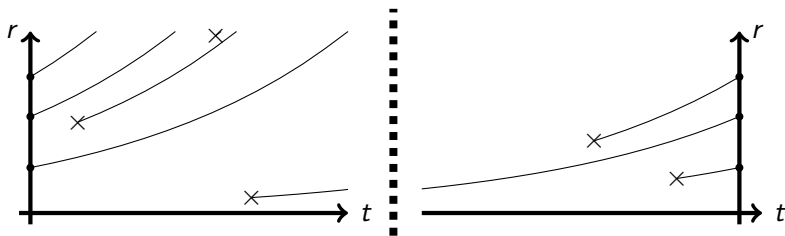
for all $t \in \mathbb{R}$, where $\tau_t(s, x) = (s + t, x)$.

Theorem [Wei, S] $\{\mu \in \mathcal{M}(\mathcal{X}^{\mathbb{N}}) : \forall p \in (0, 1], \theta_p(\mu) = \mu\}$ is isomorphic as a simplex to $\mathcal{M}_{\text{I}}(\mathcal{M}_{\text{Leb}}(\mathbb{R} \times \mathcal{X}))$.

- ▶ Given Q , let $\rho \in \mathcal{M}_{\text{Leb}}(\mathbb{R} \times \mathcal{X})$ be chosen according to Q .
- ▶ Let ρ' be a new measure: $\rho'(dt \times dx) = e^t \rho(dt \times dx)$.
- ▶ Let Ξ be a Poisson process on $\mathbb{R} \times \mathcal{X}$ with intensity ρ' .
- ▶ A.s., $\Xi = \sum_{n=1}^{\infty} \delta_{(\xi_n, X_n)}$ with $\xi_1 < \xi_2 < \dots$.
- ▶ $\mu_Q \in \mathcal{M}(\mathcal{X}^{\mathbb{N}}) = \text{the marginal distribution of } (X_1, X_2, \dots)$.

7/10 Key idea: Hoyle's super-creationist model

In one of his papers, Aldous describes the following as Hoyle's "steady state model."



- ▶ At any time, standard Poisson point process.
- ▶ Space dilates at a constant rate.
- ▶ New points added according to space-time Poisson process.

A random partition structure “is” a random sequence (ξ_1, ξ_2, \dots) :

- ▶ $\xi_1 > \xi_2 > \dots > 0$
- ▶ $\xi_1 + \xi_2 + \dots = 1$

Thinning here has two steps:

- ▶ $(\xi_{K_1}, \xi_{K_2}, \dots)$,
- ▶ $(\xi_{K_1}/Z, \xi_{K_2}/Z, \dots)$ where $Z = \xi_{K_1} + \xi_{K_2} + \dots$

Example. For $0 < m < 1$, let $\xi_1 > \xi_2 > \dots \sim \text{PPP}(m x^{-m-1} dx)$.

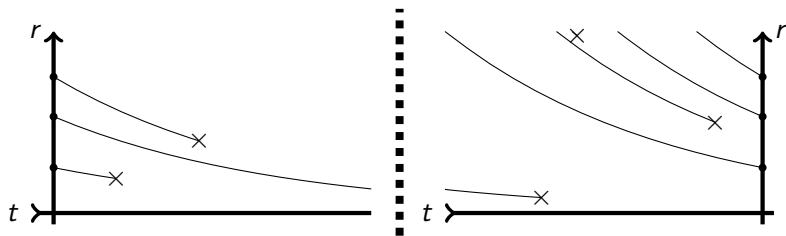
Let $Z = \xi_1 + \xi_2 + \dots$

$(\xi_1/Z, \xi_2/Z, \dots) \sim \text{Poisson-Dirichlet PD}(m, 0)$

References:

Pitman and Yor, The two parameter Poisson-Dirichlet process derived from a stable subordinator.

Pitman, Poisson-Kingman partitions.



$$-t^{-1} \ln \xi_1(t) \rightarrow 1/m$$

$$\Rightarrow \xi_n(0) \sim n^{-1/m}$$

Idea: *still only heuristic.* Multiply $\xi_n(0)$ by $(e^t N_n(-t))^{1/m}$ and then claim the result is a thinning invariant sequence.

* This problem arises in the simplest spin glass, Derrida's REM. Aizenman and Ruzmaikina solved the cavity step dynamics. But $N \rightarrow N + 1$ also doubles the configuration space.



Thank you!