

Free energy of a particle in high-dimensional Gaussian potentials with isotropic increments

Anton Klimovsky

EURANDOM
Eindhoven University of Technology

September 1, 2011, Prague

<http://arxiv.org/abs/1108.5300>

Gaussian fields with isotropic increments

- ▶ **Gaussian random field:** $X_N = \{X_N(u) : u \in \mathbb{R}^N\}$.
- ▶ $X_N(u)$ centred **Gaussian**, $u \in \mathbb{R}^N$.
- ▶ **Isotropic increments:**

$$\mathbb{E} [(X_N(u) - X_N(v))^2] = D \left(\frac{1}{N} \|u - v\|_2^2 \right) =: D_N(\|u - v\|_2^2), \quad u, v \in \mathbb{R}^N.$$

- ▶ **NB!** $N \gg 1$.
- ▶ **Any (admissible) $D : \mathbb{R}_+ \rightarrow \mathbb{R}_+$.**

Complete classification of the correlators

A.M. Yaglom (1957):

1. Isotropic field:

$$\mathbb{E} [X_N(u)X_N(v)] = B \left(\frac{1}{N} \|u - v\|_2^2 \right), \quad u, v \in \Sigma_N,$$

$$B(r) = c_0 + \int_0^{+\infty} \exp(-t^2 r) \nu(dt),$$

$c_0 \in \mathbb{R}_+$, $\nu \in \mathcal{M}_{\text{finite}}(\mathbb{R}_+)$.

$$D(r) = 2(B(0) - B(r)).$$

2. Non-isotropic field with isotropic increments:

$$D(r) = \int_0^{+\infty} [1 - \exp(-t^2 r)] \nu(dt) + A \cdot r, \quad r \in \mathbb{R}_+,$$

$A \in \mathbb{R}_+$, $\nu \in \mathcal{M}((0; +\infty))$

$$\int_0^{+\infty} \frac{t^2 \nu(dt)}{t^2 + 1} < \infty.$$

A particle subjected to a rugged potential:

- ▶ **Particle state space:**

$$S_N := S^N, \quad S \subset \mathbb{R}, \quad \text{or} \quad S_N := \{u \in \mathbb{R}^N : \|u\|_2 \leq L\sqrt{N}\}, \quad L > 0.$$

- ▶ **Partition function:**

$$Z_N(\beta) := \int_{S_N} \mu_N(du) \exp\left(\beta \sqrt{N} X_N(u)\right), \quad \beta \in \mathbb{R}_+.$$

- ▶ **Log-partition function:**

$$p_N(\beta) := \frac{1}{N} \log Z_N(\beta).$$

▶ **Q:**

$$\lim_{N \rightarrow +\infty} p_N(\beta) =: p(\beta) = ?$$

Parisi-type functional

- ▶ Regularised derivative:

$$D^{',M}(r) := \begin{cases} D'(r), & r \in [1/M; +\infty), \\ M, & r \in [0; 1/M). \end{cases}$$

- ▶ Parisi terminal value problem:

$$\begin{cases} \partial_q f(y, q) + \frac{1}{2} D^{',M}(2(r-q)) \left(\partial_{qq}^2 f(y, q) + \mathbf{x}(q) (\partial_y f(y, q))^2 \right) = 0, \\ f(y, 1) = h(y), \quad q \in (0, r), \quad y \in \mathbb{R}. \end{cases}$$

- ▶ Spin glass order parameter:

$$\mathbf{x} \in \mathcal{X}(r) := \{ \mathbf{x} : [0, r] \rightarrow [0, 1] \mid \text{càdlàg } \uparrow, \mathbf{x}(0) = 0, \mathbf{x}(r) = 1 \}.$$

- ▶ Boundary conditions (product state space)

$$h_\lambda(y) := \log \int_S \mu(du) \exp(\beta u y + \lambda u^2), \quad y \in \mathbb{R}, \quad \lambda \in \mathbb{R}.$$

- ▶ Parisi-type functional:

$$\mathcal{P}(\beta, r)[\mathbf{x}] := \lim_{M \uparrow +\infty} \left(\inf_{\lambda \in \mathbb{R}} \left[f_{r, \mathbf{x}, h_\lambda}^{(M)}(0, 0) - \lambda r \right] - \frac{\beta^2}{2} \int_0^1 \mathbf{x}(q) d\theta_r^{(M)}(q) \right),$$

$$\theta^{(M)}(q) := -q D^{',M}(q) - D(q), \quad q \in \mathbb{R}_+.$$

Variational formula

Effective size of the state space:

$$d := \sup_N \left(\frac{1}{N} \sup_{u \in S_N} \|u\|_2^2 \right).$$

Theorem

$$p(\beta) := \sup_{r \in [0; d]} \inf_{x \in \mathcal{X}(r)} \mathcal{P}(\beta, r)[x], \quad \text{almost surely and in } L^1.$$

Heuristics: "localisation"

► **Covariance structure:**

$$\mathbb{E} [X_N(u)X_N(v)] = \frac{1}{2} (D_N(\|u\|_2^2) + D_N(\|v\|_2^2) - D_N(\|u - v\|_2^2)), \quad u, v \in \mathbb{R}^N.$$

► **Overlap:**

$$\langle u, v \rangle_N := \frac{1}{N} \sum_{i=1}^N u_i v_i, \quad u, v \in \mathbb{R}^N.$$

► **Fix $r \in [0; d]$:**

$$\begin{aligned} \mathbb{E} [X_N(u)X_N(v)] &= D(r) - \frac{1}{2}D(2(r - \langle u, v \rangle_N)), \\ \|u\|_2^2 &= \|v\|_2^2 = rN. \end{aligned}$$

► \Rightarrow **Localisation.**

Particle in a rotationally invariant box

► Particle state space:

$$S_N := \{u \in \mathbb{R}^N : \|u\|_2 \leq L\sqrt{N}\}.$$

► A priori measure: $\mu_N \in \mathcal{M}_{\text{finite}}(S_N)$:

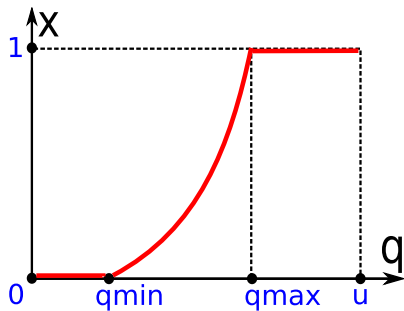
$$\frac{d\mu_N}{d\lambda_N}(u) := \exp\left(\sum_{i=1}^N f(u_i)\right), \quad u = (u_i)_{i=1}^N \in \mathbb{R}^N, \quad f: \mathbb{R} \rightarrow \mathbb{R},$$

$$f(u) := h_1 u - h_2 u^2, \quad h_1 \in \mathbb{R}, \quad h_2 \in \mathbb{R}_+.$$

► Fyodorov, Sommers (2007):

$$\begin{aligned} \mathcal{E}\mathcal{S}(\beta, r)[x] &:= \frac{1}{2} \left[\log(r - q_{\max}) + \int_0^{q_{\max}} \frac{dq}{\int_q^r x(s) ds} + h_1^2 \int_0^r x(q) dq - h_2 r \right] \\ &\quad + \frac{\beta^2}{2} \left(D'(2(r - q_{\max})) + \int_0^{q_{\max}} D'(2(r - q)) x(q) dq \right), \\ &x \in \mathcal{X}(r). \end{aligned}$$

A test function

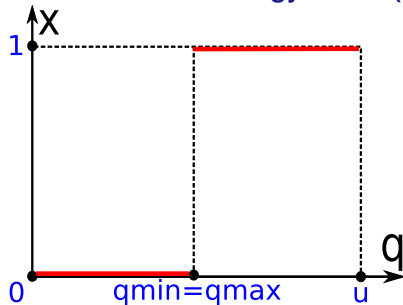


Short range: solution of the variational problem

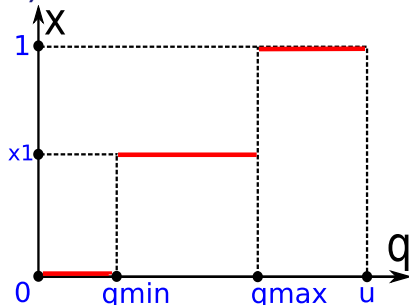
Short range: $D(r) := B(0) - B(r)$

$$3[D'''(r)]^2/2 - D''(r)D'''(r) =: S(r) > 0, \quad u \in \mathbb{R}_+,$$

Derrida's random energy model (REM) behaviour:



$\beta \in [0; \beta_c) \Rightarrow$ **RS optimiser**



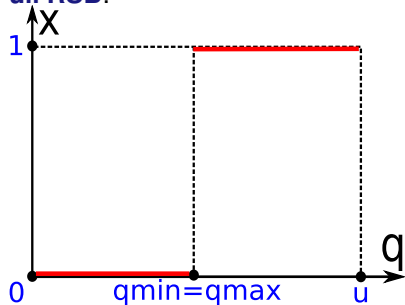
$\beta \in (\beta_c; +\infty] \Rightarrow$ **1-RSB optimiser**

Long range: solution of the variational problem

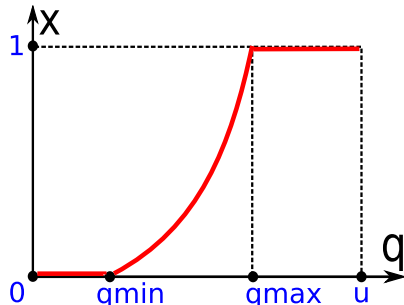
Long range: if \mathbf{D} satisfies

$$S(r) < 0, \quad u \in \mathbb{R}_+,$$

Full RSB:



$\beta \in [0; \beta_c) \Rightarrow$ **RS optimiser**



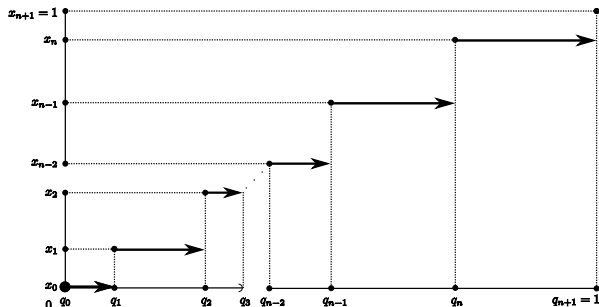
$\beta \in (\beta_c; +\infty] \Rightarrow$ **FRSB optimiser**

Critical range: logarithmic correlations

Long range: D satisfies

$$S(r) = 0, \quad u \in \mathbb{R}_+,$$

- ▶ $D(r) = \log(c + r)$, $c > 0 \Rightarrow$ **REM-behaviour** (at the level of \mathbb{E})
- ▶ $D(r) = \sum_{k=0}^{n+1} K_i \log(c_i + r)$ ($c_0 > c_1 > \dots > c_{n+1}$ and $K_i > 0$) \Rightarrow **generalised REM-behaviour** (n -RSB):



Sketch of proof

Compare with a class of hierarchically correlated fields.

- ▶ “Generalised random energy model”: $r \in [0; d]$:

$$\text{Cov} [a(\alpha^{(1)}), a(\alpha^{(2)})] = -D'(2(r - q(\alpha^{(1)}, \alpha^{(2)}))), \quad \alpha^{(1)}, \alpha^{(2)} \in \mathbb{N}^n,$$

where

$$0 = q_0 < q_1 < \dots < q_n < q_{n+1} = r,$$

ultrametric overlap:

$$q(\alpha^{(1)}, \alpha^{(2)}) := q_{\max\{k \in [1; n] \cap \mathbb{N} : [\alpha^{(1)}]_k = [\alpha^{(2)}]_k\}}.$$

- ▶ **Comparison process:**

$$A(u, \alpha) := \sum_{i=1}^N u_i a_i(\alpha), \quad u \in S_N, \quad \alpha \in \mathbb{N}^n.$$

Multiplicative probability cascades

- ▶ Let $\{g_i\}_{i=1}^{2^N}$ be i.i.d. r.v. of **Gumbel** extremal-type (say, Gaussian)

$$\sum_{i=1}^{2^N} \delta_{\exp(u_N(g_i)/x)} \xrightarrow{N \rightarrow +\infty} \xi(x), \quad x \in (0; 1).$$

where $u_N(y) = a_N y + b_N$ is the **linear extreme value normalisation**.

- ▶ **Ruelle (1987):**

$$\xi(x) := \{\delta_{\xi(x)_i}\}_{i \in \mathbb{N}} := \text{PPP}(\mathbb{R}_+ \ni t \mapsto xt^{-x-1}), \quad x \in (0; 1)$$

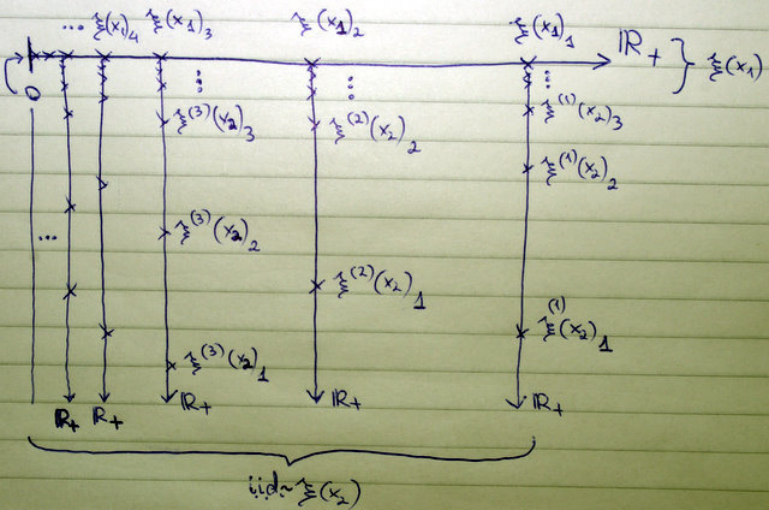
Cascades

Ruelle's probability cascades (RPC):

$$\begin{aligned} \{ \xi(x_1)_{i_1} \cdot \xi^{(i_1)}(x_2)_{i_2} \cdot \xi^{(i_1, i_2)}(x_3)_{i_3} \cdots \xi^{(i_1, i_2, \dots, i_{n-1})}(x_n)_{i_n} : i = (i_1, \dots, i_n) \in \mathbb{N}^n \} \\ =: \mathbf{RPC}(x_1, \dots, x_n), \end{aligned}$$

where $0 < x_1 < \dots < x_n < 1$, $\xi^{(i_1, i_2, \dots, i_{k-1})}(x_k)$ i.i.d. $\xi(x_k)$.

RPC construction (sketch)



Comparison

► **Interpolation:**

$$H_t(u, \alpha) := \sqrt{t}X(u) + \sqrt{1-t}A(u, \alpha), \quad t \in [0; 1], \quad u \in S_N, \quad \alpha \in \mathbb{N}^n.$$

► **Extended free energy functional:**

$$\Phi_N(x)[H_t] := \frac{1}{N} \mathbb{E} \left[\log \left(\int_{S_N} \mu_N(du) \sum_{\alpha \in \mathbb{N}^n} \text{RPC}(x)_\alpha \exp \left(\beta \sqrt{N} H_t(u, \alpha) \right) \right) \right]$$

► **Fundamental theorem of calculus:**

$$p_N(\beta) = \Phi_N(x)[H_1] = \Phi_N(x)[H_0] + \int_0^1 \frac{d}{dt} \Phi_N(x)[H_t] dt,$$

where

- **nonlinear summand** = $\Phi_N(x)[H_0]$.
- **linear summand** + (annoying) **remainder** = $\int_0^1 \frac{d}{dt} \Phi_N(x)[H_t]$.