

maximize $H_N(\sigma) : \sigma \in \{\pm 1\}^N$ generalized p-spin

$$H_N(\sigma) := \frac{1}{2} \langle \sigma, W \sigma \rangle, \quad W \sim \text{GOE}(n) : \text{SK model}$$

AMP $z^t \in \mathbb{R}^N, t \in \{0, \delta, 2\delta, \dots, 1-\delta, 1\}$

$$\underline{z^{t+\delta}} = W f_t(z^{st}) - \sum_{s=s}^t d_{t,s} f_{s-\delta}(z^{ss-\delta}) \quad \underline{z^0 \sim N(0, \delta I_N)}$$

Lem $\hat{\mathbb{E}}_N(\cdot) := \frac{1}{N} \sum_{i=1}^N (\cdot)$

$$p.\lim_{N \rightarrow \infty} \hat{\mathbb{E}}_N \psi(z_i^0, \dots, z_i^1) = \mathbb{E} \psi(z_0^\delta, z_1^\delta, \dots, z_1^\delta)$$

$$(z_0^\delta \dots z_1^\delta) \sim N(0, Q) \quad z_i^\delta \perp (z_0^\delta \dots z_{i-1}^\delta)$$

$$Q_{s+\delta, t+\delta} = \mathbb{E} \{ f_s(z_{ss}^\delta) f_t(z_{st}^\delta) \}$$

$m^t := f_t(z^{st})$, want

$$\frac{1}{N} \langle m^{t+\delta} - m^t, m^s \rangle \xrightarrow{N \rightarrow \infty} 0 \quad s \leq t$$

$$\mathcal{F}^{t,N} = \sigma(\{z^0, \dots, z^t\})$$

$$m^t := m^0 + \sum_{s=0}^{t-\delta} u^s (z^{s+\delta} - z^s)$$

$$m^0 = \sqrt{\delta} \mathbf{1}$$

$$u^s \in m \mathcal{F}^{s,N}$$

$$N < \infty$$

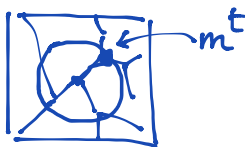
$$z^t, u^t, m^t$$

$$N = \infty$$

$$z_t^\delta, U_t^\delta, M_t^\delta$$

State evolution

$$\begin{cases} M_t^\delta = \sqrt{\delta} + \sum_{s=0}^{t-\delta} U_s^\delta (z_{s+\delta}^\delta - z_s^\delta) ; \quad U_s^\delta \in m \mathcal{F}^s \\ \mathbb{E}(z_{t+\delta}^\delta z_{s+\delta}^\delta) = \mathbb{E}(M_t^\delta M_s^\delta) \end{cases} \quad \square$$



Claim $(M_t^\delta)_{t \leq 1}$ $(Z_t^\delta)_{t \leq 1}$ are MC \square

Proof - Sufficient to prove for Z
 - Suff to check $\mathbb{E}(Z_t^\delta Z_s^\delta) = \vartheta_{(t \wedge s) - \delta}$

By induction over t . Assume true up to t

WTS $\mathbb{E}(Z_{t+\delta}^\delta Z_{s+\delta}^\delta) \stackrel{?}{=} \vartheta_s^\delta \quad \forall s \leq t$

$$= \delta + \sum_{\substack{t' \leq t-\delta \\ s' \leq s-\delta}} \mathbb{E}\{U_{s'}(Z_{s'+\delta}^\delta - Z_{s'}) U_{t'}(Z_{t'+\delta}^\delta - Z_{t'})\}$$

$$\vartheta_s^\delta = \delta \sum_{s' \leq s-\delta} \mathbb{E}[U_{s'}^2] (\vartheta_{t'+\delta}^\delta - \vartheta_{t'}) \quad \square$$

How to choose U_s^δ ?

$$U_s^\delta = \frac{\bar{U}_s^\delta}{\mathbb{E}[(\bar{U}_s^\delta)^2]^{1/2}}$$

$$\left\{ \begin{array}{l} \vartheta_s^\delta = \delta \sum_{t \leq s-\delta} (\vartheta_{t+\delta}^\delta - \vartheta_t^\delta) \\ \vartheta_0^\delta = 0 \end{array} \right.$$

$$\boxed{\vartheta_t^\delta = t}$$

$$\bar{U}_t^\delta = u(X_t^\delta; t)$$

$$\boxed{X_{t+\delta}^\delta = X_t^\delta + v(X_t^\delta; t) \cdot \delta + (Z_{t+\delta}^\delta - Z_t^\delta)}$$

$v, u: \mathbb{R} \times [0, 1] \rightarrow \mathbb{R}$.

$$\mathbb{E}(Z_t^\delta Z_s^\delta) = \vartheta_{t \wedge s}^\delta = \underline{t \wedge s} \quad \underline{Z_t^\delta = B_t} \quad \text{BM}$$

$$\boxed{M_t^\delta = \sqrt{\delta} + \sum_{s=0}^{t-\delta} \frac{u(X_s^\delta; s)}{\mathbb{E}[u(X_s^\delta; s)^2]^{1/2}} (B_{s+\delta} - B_s)}$$

$\delta \rightarrow 0 \quad (X_t^\delta) \rightarrow (X_t) \dots$

$$dX_t = v(X_t, t) dt + dB_t$$

$$M_t = \int_0^t \frac{u(X_s, s)}{\mathbb{E}[u(X_s, s)^2]^{1/2}} dB_s$$

to $\mathbb{E}(u(X_s, s)) = 1$

$$\mathbb{E}(u(X_s, s))^2 = 1 \quad \forall s \Leftrightarrow \mathbb{E}(M_t^2) = t$$

$$\begin{cases} M_t = \int_0^t \sqrt{\xi''(s)} u(s, X_s) dB_s \\ dX_t = v(t, X_t) dt + \sqrt{\xi''(t)} dB_t \end{cases}$$

$$m^t \in \mathbb{R}^N \quad m^1 \rightarrow \text{Thresh} \rightarrow \text{Round} \rightarrow \sigma^{\text{alg}}$$

Lem $\frac{1}{N} H_N(\sigma^{\text{alg}}) \geq \frac{1}{N} H_N(m^1) + o_N(1)$

Lem $\frac{1}{N} H_N(m^1) = \mathcal{E}(u, v) + o_N(1)$

$$\mathcal{E}(u, v) := \int_0^1 \xi''(t) \mathbb{E} u(t, X_t) dt \quad \square$$

Alg design

$$\begin{cases} \text{maximize} & \mathcal{E}(u, v) \\ \text{subj to} & \mathbb{E}(M_t^2) = t, \quad M_t \in (-1, 1) \text{ a.s.} \\ & \forall t \end{cases}$$

1) Relax

2) UB via duality

3) Dual \rightarrow Primal

$$\mathbb{E}(M_t^2) = 1 \quad (\|m^1\|^2 = N) \quad \xi(x) = x^2/2 \quad \text{VAL} = 1$$

REL $u_t \in m^{\mathcal{F}_t} \quad \mathcal{F}_t = \sigma(\{B_s\}_{s \leq t})$

$$\begin{aligned} \max & \int_0^1 \xi''(t) \mathbb{E} u_t dt \\ \text{s.t.} & M_t = \int_0^t \sqrt{\xi''(s)} u_s dB_s \quad M_t \in (-1, 1) \end{aligned}$$

$$\mathbb{E}(M_t^2) = t \quad \forall t$$

$$\gamma: [0,1] \rightarrow \mathbb{R}_{\geq 0}, \quad v(t) := \int_t^1 \xi''(s) \gamma(s) ds$$

$$J_\gamma := \sup_u \mathbb{E} \left\{ \int_0^1 \xi''(t) u_t^2 + \frac{1}{2} v(t) (u_t^2 \xi''(t) - 1) dt \right\}$$

st. $\int_0^1 \sqrt{\xi''(t)} u_t dB_t \in (-1,1)$

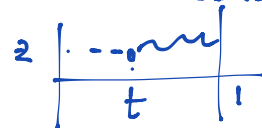
$$\text{REL} \leq J_\gamma$$

$$\begin{aligned} \mathbb{E} \int_0^1 v(t) (\xi''(t) u_t^2 - 1) dt &= \mathbb{E} \int_0^1 \int_0^1 \mathbb{1}_{s \geq t} \xi''(s) \gamma(s) (\xi''(t) u_t^2 - 1) ds dt \\ &= \int_0^1 \xi''(s) \gamma(s) (\mathbb{E} M_s^2 - s) ds = 0 \end{aligned}$$

Dynamic programming

$$J_\gamma(t, z) := \sup_{u \in D(t,1)} \mathbb{E} \left\{ \int_t^1 \left[\xi''(s) u_s^2 + \frac{1}{2} v(s) (\xi''(s) u_s^2 - 1) \right] ds \right\}$$

st. $z + \int_t^1 \sqrt{\xi''(s)} u_s dB_s \in (-1,1)$



$$J_\gamma(t, z) = \max_{\gamma \cdot u \in D(t, \theta)} J_\gamma(\theta, y) \quad J_\gamma = J_\gamma(0,0)$$

$$\theta > t$$

$\theta \rightarrow t$: HJB equation

PDE for J_γ

$$\text{REL} \leq J_\gamma(0,0)$$

Legendre fenchel dual of J_γ

$$\phi_\gamma(t, x) = \min_z [J_\gamma(t, z) - zx + \dots]$$

$\Rightarrow \phi$ satisfies Parisi's PDE

$$\text{REL} \leq \underbrace{P(\gamma)}_{\inf} \quad \underline{\gamma \in \mathcal{L}}$$

if inf is achieved at γ^*

$$v(t,x) = v_r(t,x) \quad u_f(t,x) \approx$$

$$\inf_{r \in \mathcal{L}} P(r) \quad , \quad \inf_{r \in \mathcal{U}} P(r)$$

Q1: Complexity of solving these within ϵ error.

$$\text{Q2: } d(\phi_{r_k}, \phi^\epsilon) \stackrel{?}{\leq} \epsilon$$

$$\dagger \quad m^1 \quad \text{TAP}(m^1) \approx 0$$

$$\frac{1}{N} \Phi(\mu_p \parallel \hat{\mu}) = o(1) \leftarrow$$

$$? \quad \|\mu_p - \hat{\mu}\|_{TV} = o(1) \leftarrow$$