



Algorithmic Obstructions in the Random Number Partitioning Problem

David Gamarnik¹ Eren C. Kızıldağ²

¹MIT ORC ²MIT EECS

Random Number Partitioning Problem (NPP)

Setup: Given n items $X_1, \dots, X_n \in \mathbb{R}$; partition them into two “bins” with total weights as close as possible: $\min_{\mathcal{A} \subset [n]} \left| \sum_{i \in \mathcal{A}} X_i - \sum_{i \in \mathcal{A}^c} X_i \right|$. Equivalently,

$$\min_{\sigma \in \mathcal{B}_n} \left| \langle \sigma, X \rangle \right|, \text{ where } \mathcal{B}_n = \{-1, 1\}^n \text{ and } \langle \sigma, X \rangle = \sum_{1 \leq i \leq n} \sigma_i X_i.$$

Our focus: X_i are i.i.d. standard normal: $X_i \stackrel{d}{=} \mathcal{N}(0, 1)$.

Applications

Randomized controlled trials: Gold standard for clinical trials (Krieger et al.'19; Harshaw et al.'19).

- n persons with covariate info (age, weight, height,...) $X_i \in \mathbb{R}^d, 1 \leq i \leq n$.
- Split into two groups (*treatment* and *control*) with similar “features”:

$$\min_{\sigma \in \mathcal{B}_n} \|X\sigma\|_{\infty}, \text{ where } X = (X_1, X_2, \dots, X_n) \in \mathbb{R}^{d \times n}.$$

- **Goal.** Accurate inference for a treatment effect.

Many more applications: *multiprocessor scheduling, VLSI design, cryptography...*

Available Guarantees

Existential: Let $X_i \stackrel{d}{=} \mathcal{N}(0, 1), 1 \leq i \leq n$ i.i.d. Then,

$$\min_{\sigma \in \mathcal{B}_n} |\langle \sigma, X \rangle| = \Theta(\sqrt{n}2^{-n}), \text{ w.h.p. as } n \rightarrow \infty.$$

Non-constructive. Extends to high dimensions: $\Theta(\sqrt{n}2^{-n/d})$ for $2 \leq d \leq o(n)$ (Turner et al.'20).

Algorithmic (Polynomial-Time): *Largest Differencing Method (LDM)* by Karmarkar and Karp'82.

For $d = 1$ and $X_i \stackrel{d}{=} \mathcal{N}(0, 1), 1 \leq i \leq n$ i.i.d.; returns a $\sigma_{\text{ALG}} \in \mathcal{B}_n$ such that

$$|\langle \sigma_{\text{ALG}}, X \rangle| = 2^{-\Theta(\log^2 n)} \text{ w.h.p. as } n \rightarrow \infty.$$

Extends to high dimensions: $\exp(-\Omega(\log^2 n/d))$ for $2 \leq d \leq O(\sqrt{\log n})$ (Turner et al.'20).

A Statistical-to-Computational Gap

Gap between **existential** guarantees and what **polynomial-time** algorithms can promise.

- **Our focus:** Dimension $d = 1$. For $X_i \stackrel{d}{=} \mathcal{N}(0, 1), 1 \leq i \leq n$ i.i.d.

$$\min_{\sigma \in \mathcal{B}_n} |\langle \sigma, X \rangle| = \Theta(\sqrt{n}2^{-n}) \text{ vs } |\langle \sigma_{\text{ALG}}, X \rangle| = 2^{-\Theta(\log^2 n)}.$$

- Ignoring \sqrt{n} , a striking gap: 2^{-n} vs $2^{-\Theta(\log^2 n)}$.

Source of this gap/hardness?

Study of Statistical-to-Computational Gap

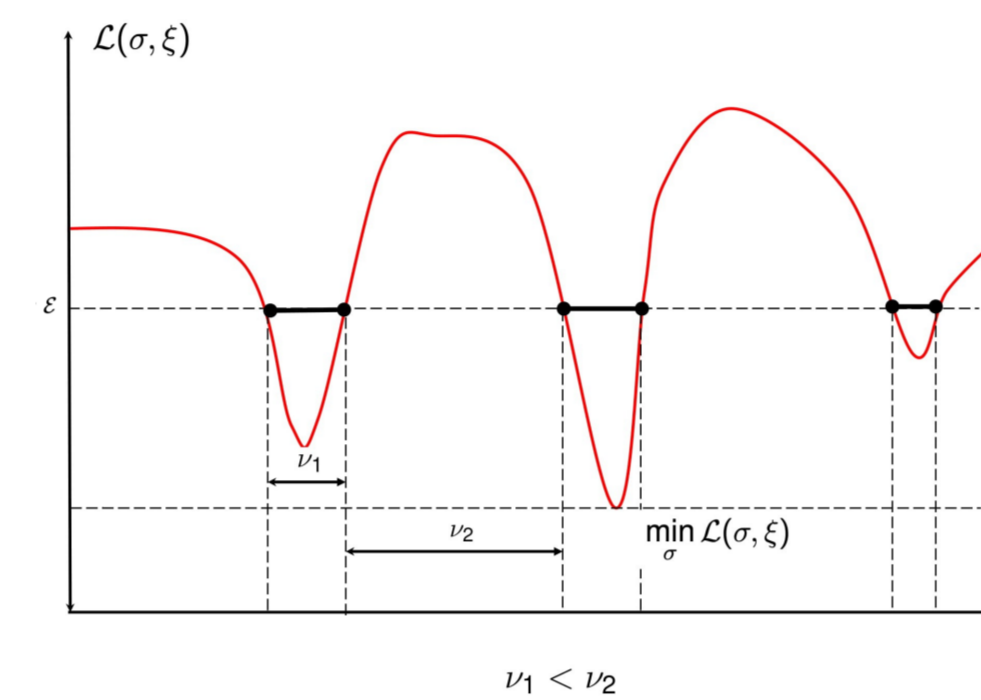
Common feature in many algorithmic problems in **high-dimensional statistics** & **random combinatorial structures**: *Random k-SAT, optimization over random graphs, p-spin model, planted clique, matrix PCA, linear regression, spiked tensor, largest submatrix problem...*

Average-Case Problems: No analogue of worst-case theory (such as $P \neq NP$). Various Forms of **Rigorous Evidences of Hardness**: *low-degree methods, reductions from the planted clique, failure of MCMC, failure of BP/AMP, SoS lower bounds...*

Overlap Gap Property (OGP)

Another approach from *spin glass theory*: **Overlap Gap Property (OGP)**.

- Generic optimization problem with random ξ : $\min_{\theta \in \Theta} \mathcal{L}(\sigma, \xi)$.
- (Informally) OGP for **energy** \mathcal{E} if $\exists 0 < \nu_1 < \nu_2$ s.t. w.h.p. over $\xi, \forall \sigma_1, \sigma_2 \in \Theta$,
 $\mathcal{L}(\sigma_j, \xi) \leq \mathcal{E} \implies \text{distance}(\sigma_1, \sigma_2) < \nu_1 \text{ or } \text{distance}(\sigma_1, \sigma_2) > \nu_2$.
- Any two **near optimal** σ_1, σ_2 are either *too similar* or *too dissimilar*.



First algorithmic implication: Maximum independent set in $\mathbb{G}_d(n)$ and $\mathbb{G}(n, \frac{d}{n})$ (Gamarnik and Sudan'13). **Many problems with OGP:** *random k-SAT, NAE-k-SAT, p-spin model, sparse PCA, largest submatrix problem, max-CUT, planted clique...*

OGP as a Provable Barrier to Algorithms: *WALKSAT, local algorithms, stable algorithms, low-degree polynomials, AMP, MCMC, low-depth circuits...*

Landscape Results: Presence of OGP

Theorem. $\forall \epsilon \in (1/2, 1), \exists \rho := \rho(\epsilon) \in (0, 1)$ such that if $\sigma, \sigma' \in \mathcal{B}_n$ achieve

$$|\langle \sigma, X \rangle| = O(\sqrt{n}2^{-\epsilon n}) \text{ and } |\langle \sigma', X \rangle| = O(\sqrt{n}2^{-\epsilon n})$$

then either $\sigma = \sigma'$ or $n^{-1}|\langle \sigma, \sigma' \rangle| \leq \rho$ w.h.p. That is, $n^{-1}|\langle \sigma, \sigma' \rangle| \notin (\rho, \frac{n-2}{n})$.

- Partitions achieving better than $2^{-\frac{n}{2}}$ are isolated vectors separated by $\Theta(n)$ distance.
- Yields existence of a *Free Energy Well (FEW)*: failure of **Glauber dynamics**.

Still large gap between $2^{-\frac{n}{2}}$ and $2^{-\Theta(\log^2 n)}$. **Idea:** Inspect m -tuples instead.

- Interpolate $Y_i(\tau) = \sqrt{1 - \tau^2}X_0 + \tau X_i$, where $X_0, \dots, X_m \stackrel{d}{=} \mathcal{N}(0, I_n)$ i.i.d.
- Study m -tuples $\sigma_i \in \mathcal{B}_n, 1 \leq i \leq m$, each near-optimal w.r.t. $Y_i(\tau_i)$ (Ensemble m -OGP).
- Reduce thresholds further, and rule out sufficiently stable algorithms.

Theorem. $\forall \epsilon > 0, \forall \mathcal{I} \subset [0, 1]$ with $|\mathcal{I}| = 2^{o(n)}, \exists m \in \mathbb{N}, \exists 1 > \beta > \eta > 0$ s.t. if

$$|\langle \sigma_i, Y_i(\tau_i) \rangle| = O(\sqrt{n}2^{-\epsilon n}), \tau_i \in \mathcal{I}, 1 \leq i \leq m$$

then w.h.p. $\exists 1 \leq i < j \leq m$ such that $n^{-1}|\langle \sigma_i, \sigma_j \rangle| \notin (\beta - \eta, \beta)$.

Still striking gap between $2^{-\epsilon n}$ and $2^{-\Theta(\log^2 n)}$. Unfortunately, m -OGP (with $m = O(1)$) absent for $2^{-o(n)}$. **New Idea:** Study m -tuples with $m = \omega_n(1)$.

Theorem. $\forall \omega(\sqrt{n \log n}) \leq E_n \leq o(n), \forall \mathcal{I} \subset [0, 1]$ with $|\mathcal{I}| = n^{O(1)}, \exists m_n \in \mathbb{N}, \exists 1 > \beta_n > \eta_n > 0$ s.t. if

$$|\langle \sigma_i, Y_i(\tau_i) \rangle| \leq \sqrt{n}2^{-E_n}, \tau_i \in \mathcal{I}, 1 \leq i \leq m_n$$

then w.h.p. $\exists 1 \leq i < j \leq m_n$ such that $n^{-1}|\langle \sigma_i, \sigma_j \rangle| \notin (\beta_n - \eta_n, \beta_n)$.

Algorithmic Hardness Results

Algorithm $\mathcal{A} : \mathbb{R}^n \rightarrow \mathcal{B}_n$, potentially randomized.

Stable Algorithms. Informally, \mathcal{A} is stable if small change in X yields small change in $\mathcal{A}(X)$.

Success:

$$\mathbb{P} \left(n^{-\frac{1}{2}} |\langle X, \mathcal{A}(X) \rangle| \leq E \right) \geq 1 - p_f.$$

Stability: $\exists \rho \in (0, 1), X, Y \stackrel{d}{=} \mathcal{N}(0, I_n)$ with $\text{Cov}(X, Y) = \rho I_n$;

$$\mathbb{P} \left(d_H(\mathcal{A}(X), \mathcal{A}(Y)) \leq f + L \|X - Y\|_2^2 \right) \geq 1 - p_{\text{st}}.$$

Stable algorithms include approximate message passing (AMP) type algorithms (Gamarnik and Jagannath'21) and the low-degree polynomials (Gamarnik, Jagannath, and Wein'20).

A Conjecture (verified by simulations): Largest differencing (LDM) algorithm is stable.

Theorem. Stable algorithms can't achieve value better than

$$\exp \left(-\omega \left(\frac{n}{\log^{1/5} n} \right) \right).$$

Semi-formally, $\forall \epsilon \in (0, 1/5), \forall \omega(n \log^{-1/5+\epsilon} n) \leq E_n \leq o(n)$, there is no stable \mathcal{A} that w.h.p. returns a σ with energy 2^{-E_n} (with appropriate $f, \rho', p_f, p_{\text{st}}$).

- For **extreme** case, $E_n = \Theta(n)$: rule out $p_f, p_{\text{st}} = O(1)$.
- **Proof Idea.** By contradiction. Suppose $\exists \mathcal{A}$.
 - m -OGP: a structure occurs with *vanishing probability*.
 - Run \mathcal{A} on correlated instances. Show that w.p. > 0 , *forbidden* structure occurs.
- Rate $2^{-\omega(n \log^{-1/5} n)}$: Via **Ramsey Theory**.

Failure of MCMC: Let $X \stackrel{d}{=} \mathcal{N}(0, I_n)$; define **Hamiltonian** $H(\sigma) \triangleq n^{-\frac{1}{2}} |\langle \sigma, X \rangle|$, and consider the **Gibbs distribution** at inverse temperature $\beta > 0$ on \mathcal{B}_n : $\pi_\beta(\sigma) \propto \exp(-\beta H(\sigma))$.

Construct $\mathbb{G} = (V, E)$ with $V = \mathcal{B}_n$ and $(\sigma, \sigma') \in E \iff d_H(\sigma, \sigma') = 1$; and consider any nearest neighbor MC $(X_t)_{t \geq 0}$ on \mathbb{G} reversible w.r.t. π_β .

Let $(\pm)\sigma^* = \min_{\sigma \in \mathcal{B}_n} H(\sigma)$, and for $\epsilon \in (1/2, 1], \rho := \rho(\epsilon)$ be the 2-OGP parameter. Set

$$I_1 = \left\{ \sigma : -\rho \leq \frac{1}{n} \langle \sigma, \sigma^* \rangle \leq \rho \right\}, \quad I_2 = \left\{ \sigma : \rho \leq \frac{1}{n} \langle \sigma, \sigma^* \rangle \leq \frac{n-2}{n} \right\}, \quad \text{and } I_3 = \{\sigma^*\}.$$

Theorem. For $\beta = \Omega(n2^{n\epsilon})$, w.h.p. (w.r.t. $X \stackrel{d}{=} \mathcal{N}(0, I_n)$), $\min \{ \pi_\beta(I_1), \pi_\beta(I_3) \} \geq e^{\Omega(n)} \pi_\beta(I_2)$.

I_2 is a FEW with exponentially small **Gibbs** mass separating I_3 and $I_1 \cup I_2 \cup I_3$. *Exit time* from well is **exponential**: **Slow mixing**.

Future Directions

- Formally verifying stability of **LDM**.
- Proving algorithmic hardness all the way to $2^{-\omega(\sqrt{n \log n})}$. Rate $2^{-\omega(n \log^{-1/5} n)}$ **unimprovable by Ramsey**.
- Still a significant gap $2^{-\omega(\sqrt{n \log n})}$ vs $2^{-\Theta(\log^2 n)}$.
 - Either prove **hardness** for $2^{-\omega(\log^2 n)}$: **OGP not applicable**.
 - Or devise a **better** (polynomial-time) **algorithm** achieving $2^{-\omega(\log^2 n)}$.
- **Slow mixing** for **higher** temperatures (smaller β); or for **different** initialization, e.g. *uniform* case.

Can OGP rule out **all** polynomial-time algorithms? Is there a problem **with** OGP **yet** admitting a polynomial-time algorithm?