

Noise Stability is Low-Dimensional

Anindya De, Elchanan Mossel, Joe Neeman

Gaussian noise stability

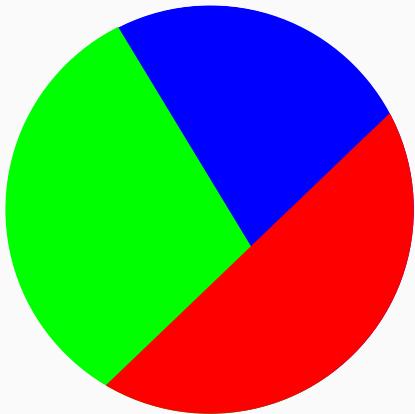
Take X and Y a pair of ρ -correlated Gaussians in \mathbb{R}^n ($0 < \rho < 1$).

For a partition $A = (A_1, \dots, A_k)$ of \mathbb{R}^n into k parts, define

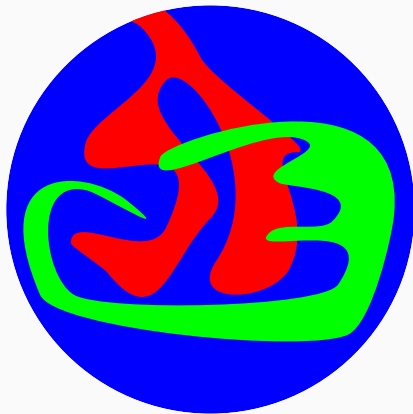
$$\text{Stab}_\rho(A) = \Pr(X \text{ and } Y \text{ land in the same part})$$

Gaussian noise stability

Noise stable



Not noise stable



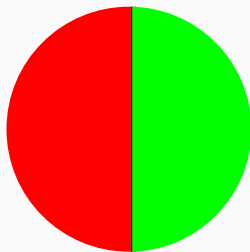
Gaussian noise stability

Theorem (Borell '85)

For a partition of \mathbb{R}^n into two parts of equal Gaussian measure,

$$\text{Stab}_\rho(A) \leq \frac{1}{2} + \frac{\sin^{-1} \rho}{\pi}.$$

Equality is attained for a partition into half-spaces.



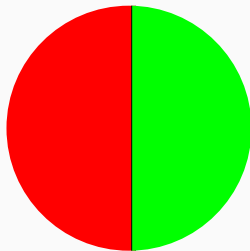
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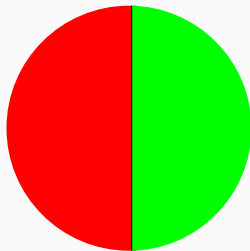
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- one-dimensional phenomenon

Gaussian noise stability

Theorem (???)

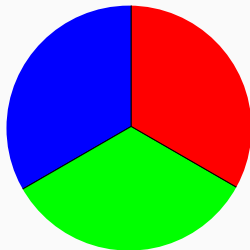
For a partition of \mathbb{R}^n into *three* parts of equal Gaussian measure,

???

Gaussian noise stability

Conjecture (Peace sign conjecture)

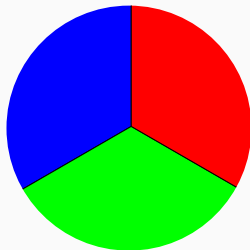
The optimal partition looks like a peace sign.



Gaussian noise stability

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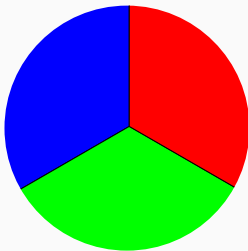
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(two-dimensional phenomenon)*



Gaussian noise stability

Conjecture (Peace sign conjecture)

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Conjecture (Multi-dimensional peace sign conjecture)

For partitions into k equal measures, the optimal partition occurs in \mathbb{R}^{k-1} . It looks like a multi-dimensional peace sign.

Theorem (De-Mossel-N.)

For any k and any $\epsilon > 0$, there is a computable $n_0 = n_0(k, \epsilon)$ such that an ϵ -approximately optimal partition occurs in \mathbb{R}^{n_0} .

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Corollary

The optimal value of k -part noise stability is computable.

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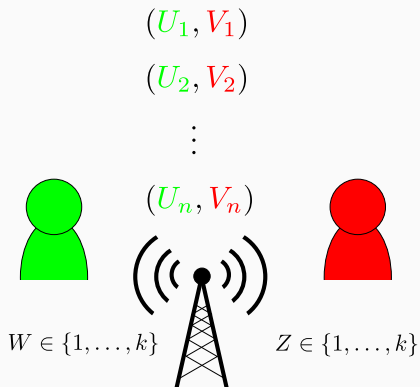
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Corollary (sort of)

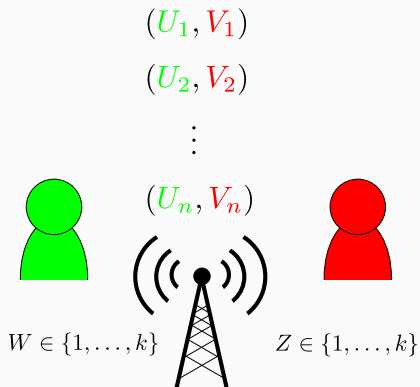
The non-interactive correlation distillation value with k -ary outputs is computable.

Correlation distillation



Goal: produce uniform output, agree with maximal probability.
What is the probability of agreement?

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Ghazi-Kamath-Sudan '16: reduction to correlated Gaussian signals

The main theorem

Theorem (De-Mossel-N.)

For any k and any $\epsilon > 0$, there is a computable $n_0 = n_0(k, \epsilon)$ such that an ϵ -approximately optimal partition occurs in \mathbb{R}^{n_0} .

Proof outline

Idea: take an optimal partition in \mathbb{R}^n (n huge) and try to “simulate” it in \mathbb{R}^{n_0} .

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1. An optimal partition is close to a bounded-degree polynomial threshold function (PTF)
2. A bounded-degree PTF can be approximately simulated by a bounded-degree PTF on a bounded number of variables

Step 1: approximation by polynomials

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Think of a partition as a function $f : \mathbb{R}^n \rightarrow \{e_1, \dots, e_k\} \subset \mathbb{R}^k$.

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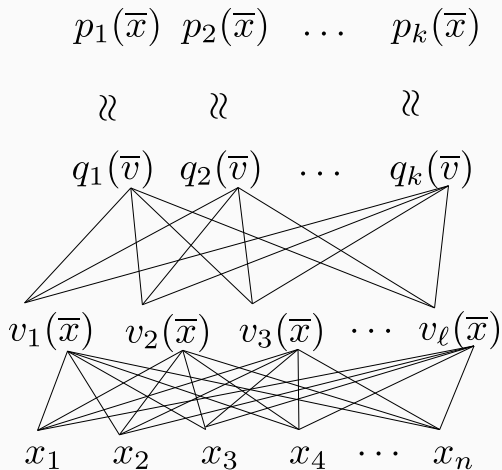
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Real proof goes through a smoothing/rounding procedure, and a connection with Gaussian surface area (KNOW '15).

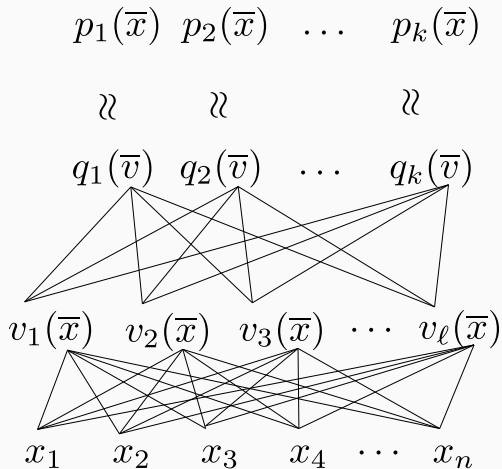
Step 2: dimension reduction

Polynomial structure theorem (De-Servedio)

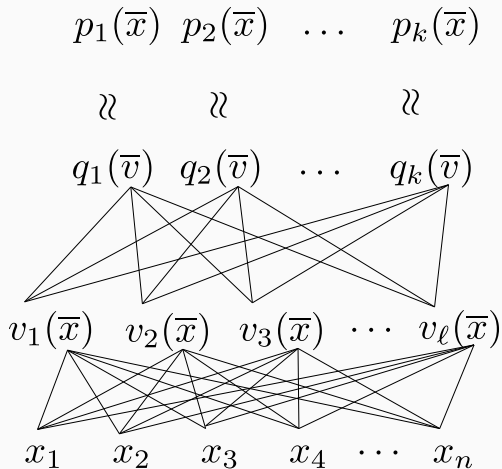


where ℓ is bounded and v_1, \dots, v_ℓ are “nice”

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“Nice” polynomials satisfy a CLT, so they may as well be linear functions of ℓ variables

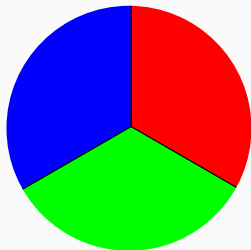
Polynomial Central Limit Theorem (De-Servedio)

$$\begin{array}{cccc} p_1(\bar{x}) & p_2(\bar{x}) & \dots & p_k(\bar{x}) \\ \Downarrow & \Downarrow & & \Downarrow \\ q_1(\bar{v}) & q_2(\bar{v}) & \dots & q_k(\bar{v}) \\ \swarrow & \swarrow & \swarrow & \swarrow \\ \sum_{i=1}^{\ell} a_{1,i}x_i & \sum_{i=1}^{\ell} a_{2,i}x_i & \sum_{i=1}^{\ell} a_{3,i}x_i & \sum_{i=1}^{\ell} a_{\ell,i}x_i \end{array}$$

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Thank you!
