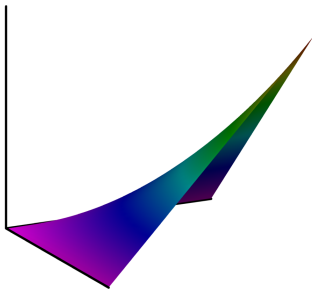


Gaussian noise stability

Elchanan Mossel¹ Joe Neeman²

¹UC Berkeley

²UT Austin



Gaussian noise stability

Fix a parameter $0 < \rho < 1$. Take

$$(X, Y) \sim \mathcal{N}\left(0, \begin{pmatrix} I_n & \rho I_n \\ \rho I_n & I_n \end{pmatrix}\right)$$

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Applications in

- ▶ approximability (e.g., optimal UGC hardness of MAX-CUT, KKMO '05)
- ▶ testing (e.g., testing half-spaces, MORS '09)

Borell's theorem

What sets have high noise stability?

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Half-spaces maximize the noise stability (among all sets of a given volume):

Theorem (Borell '85)

For any $A \subset \mathbb{R}^n$, if $A' \subset \mathbb{R}^n$ is a half-space with $\Pr(A') = \Pr(A)$ then

$$\Pr(X \in A, Y \in A) \leq \Pr(X \in A', Y \in A').$$

A *half-space* is a set of the form $\{x \in \mathbb{R}^n : x \cdot a \leq b\}$.

Borell's theorem

Define $\Phi(x) = \Pr(X_1 \leq x)$. Then $\{x \in \mathbb{R}^n : x_1 \leq \Phi^{-1}(a)\}$ is a half-space of volume a . Define

$$J(a, b) = \Pr(X_1 \leq \Phi^{-1}(a), Y_1 \leq \Phi^{-1}(b)).$$

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Since the Gaussian measure is rotationally invariant, Borell's theorem is equivalent to

$$\Pr(X \in A, Y \in A) \leq J(\Pr(A), \Pr(A)).$$

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If $\Pr(X, Y \in A) = J(\Pr(A), \Pr(A))$ then A is a.s. equal to a half-space.

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If $\Pr(X, Y \in A) \geq J(\Pr(A), \Pr(A)) - \delta$ then there is a half-space B with

$$\Pr(A \Delta B) \leq C(\rho, \Pr(A))\delta^{c(\rho)}.$$

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$$\Pr(A \Delta B) \leq \frac{C(\Pr(A))}{\sqrt{1 - \rho}} \sqrt{\delta \log(1/\delta)}.$$

Borell's theorem: previous proofs

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- ▶ Kindler-O'Donnell (when $\Pr(X \in A) = \frac{1}{2}$, and for certain values of ρ), using subadditivity.

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1. Sample $Z_1, \dots, Z_m \sim \mathcal{N}(0, I_n)$ and let $\hat{p} = \frac{\#\{Z_i \in A\}}{m}$.
2. Sample $(X_1, Y_1), \dots, (X_m, Y_m) \sim \text{Pr}_\rho$. Answer “yes” if

$$\frac{\#\{i : X_i \in A, Y_i \in A\}}{m} \geq J(\hat{p}, \hat{p}) - \tilde{O}(\epsilon^2)$$

and “no” otherwise.

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Theorem (Mossel, N. '12, Eldan '13)

If A is a half-space, then the algorithm above answers “yes” w.h.p.

If A is ϵ -far from a half-space and $m \geq \tilde{O}(\epsilon^{-4})$ then the algorithm answers “no” w.h.p.

MORS '09 showed that a similar algorithm works if $m \geq \epsilon^{-6}$.

Proof of Borell's theorem

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Recall $J(a, b) = \Pr(X_1 \leq \Phi^{-1}(a), Y_1 \leq \Phi^{-1}(b))$.

Theorem

For any $f : \mathbb{R}^n \rightarrow [0, 1]$,

$$\mathbb{E}J(f(X), f(Y)) \leq J(\mathbb{E}f, \mathbb{E}f).$$

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To get the original statement,

$$\Pr(X \in A, Y \in A) \leq J(\Pr(A), \Pr(A)),$$

set $f = 1_A$.

(Note that $J(1, 1) = 1$ and $J(0, 1) = J(1, 0) = J(0, 0) = 0$.)

Proof of Borell's theorem

Want to show $\mathbb{E}J(f(X), f(Y)) \leq J(\mathbb{E}f, \mathbb{E}f)$.

Define the operator P_t by

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The punchline: this is an increasing function of t .

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$$v_t = v_t(X) = \Phi^{-1}(P_t f(X))$$
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$$\geq 0 \quad \square$$

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Why consider $\mathbb{E}J(f(X), f(Y))$? Why does the proof work?

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and so Borell's theorem (in \mathbb{R}^{n+1}) applied to A_f gives

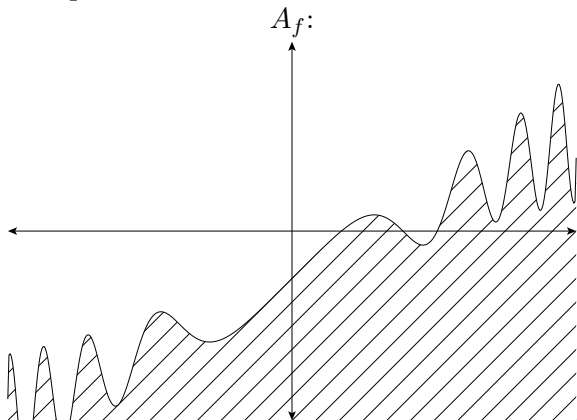
$$\begin{aligned} & \mathbb{E}J(f(X), f(Y)) \\ &= \Pr((X, X_{n+1}) \in A_f, (Y, Y_{n+1}) \in A_f) \\ &\leq J(\Pr(A_f), \Pr(A_f)) \\ &= J(\mathbb{E}f, \mathbb{E}f). \end{aligned}$$

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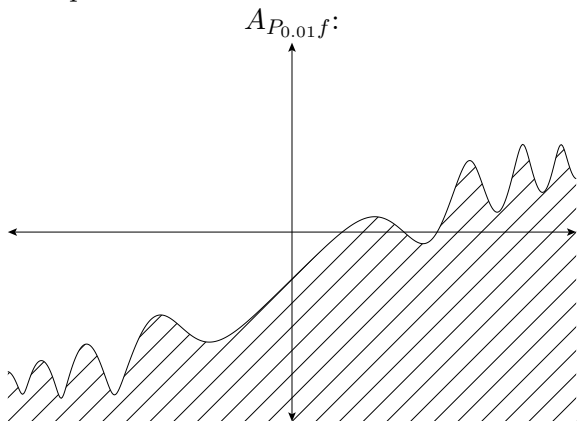
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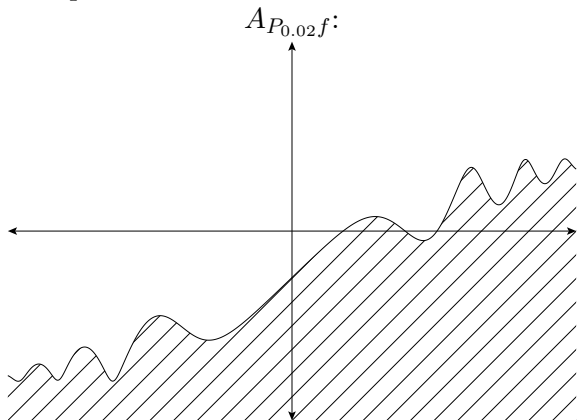
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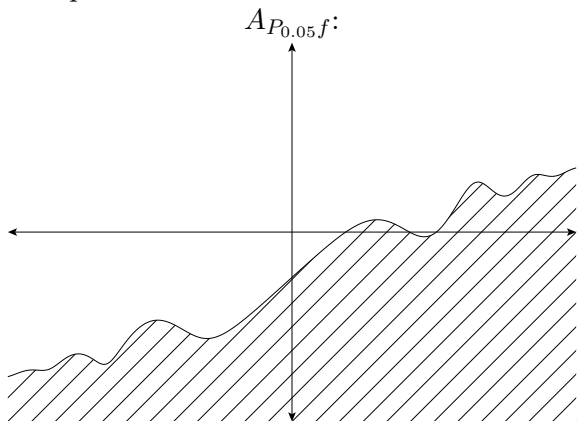
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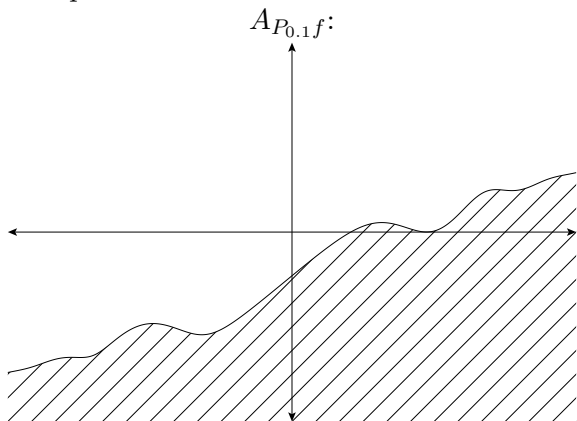
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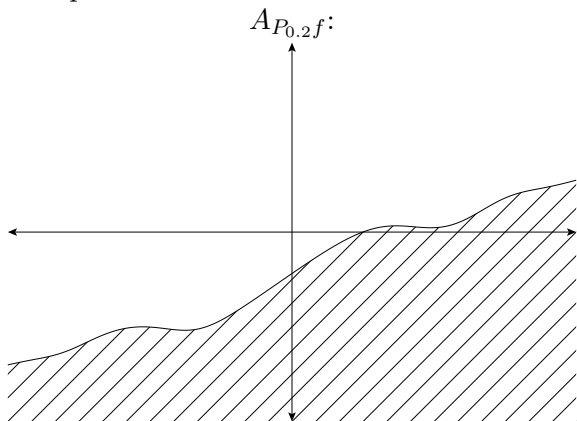
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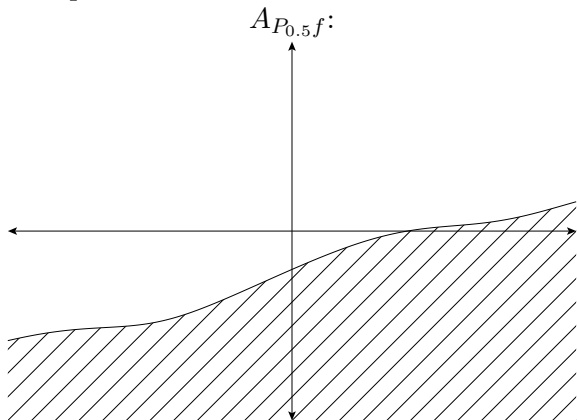
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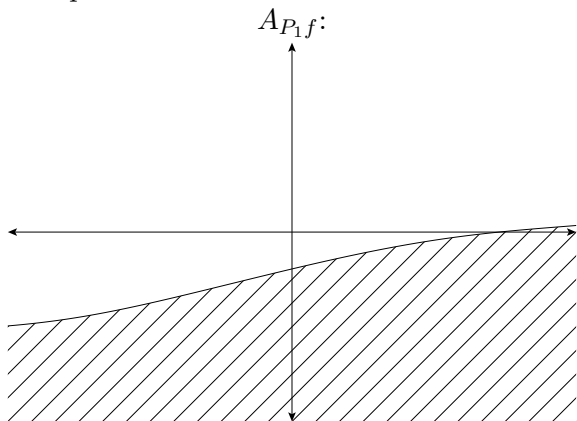
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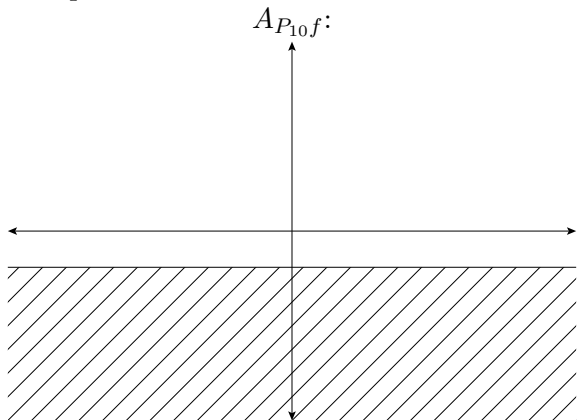
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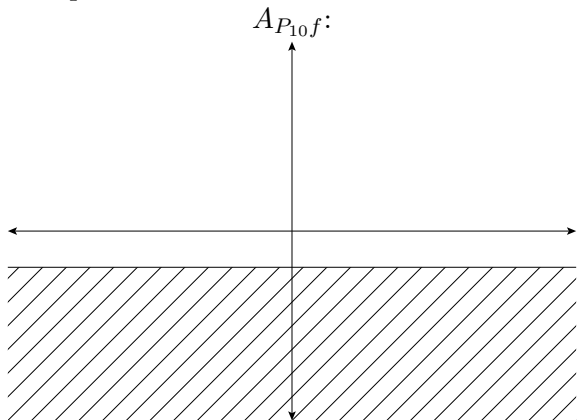
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We showed that this transformation only increases the noise stability.

This idea has been used before: Bakry and Ledoux '96 used it to prove the Gaussian isoperimetric inequality.

Borell's theorem vs. Jensen's inequality

Theorem (Mossel, N. '12)

If $J : [0, 1] \times [0, 1] \rightarrow \mathbb{R}$ satisfies $\begin{pmatrix} \frac{\partial^2 J(x,y)}{\partial x^2} & \rho \frac{\partial^2 J(x,y)}{\partial x \partial y} \\ \rho \frac{\partial^2 J(x,y)}{\partial x \partial y} & \frac{\partial^2 J(x,y)}{\partial y^2} \end{pmatrix} \leq 0$ then

$$\mathbb{E}J(f(X), f(Y)) \leq J(\mathbb{E}f, \mathbb{E}f)$$

whenever X and Y are ρ -correlated Gaussians.

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whenever X and Y are ρ -correlated Gaussians *any random variables.*

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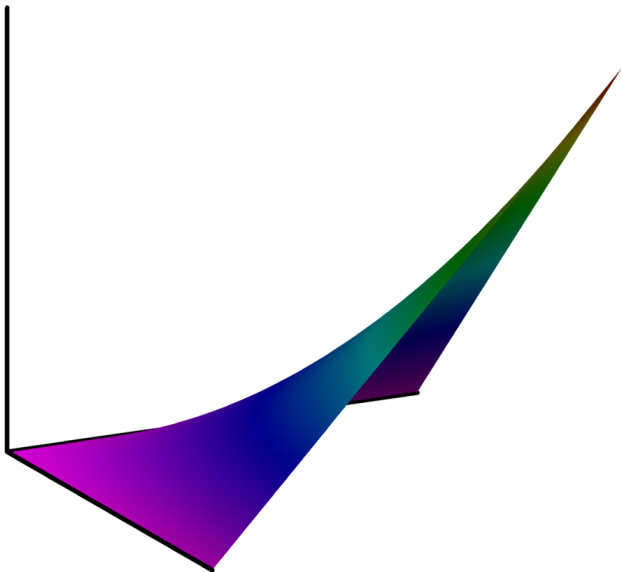
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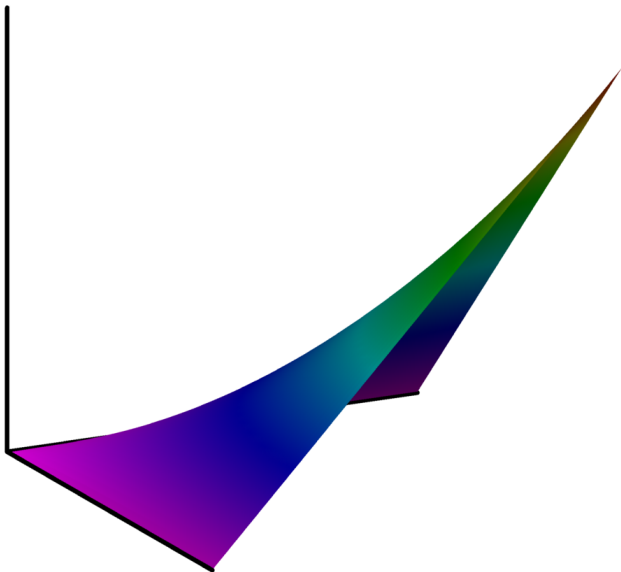
whenever X and Y are ρ -correlated Gaussians.

Does the condition mean anything? Our J is the smallest one satisfying it.

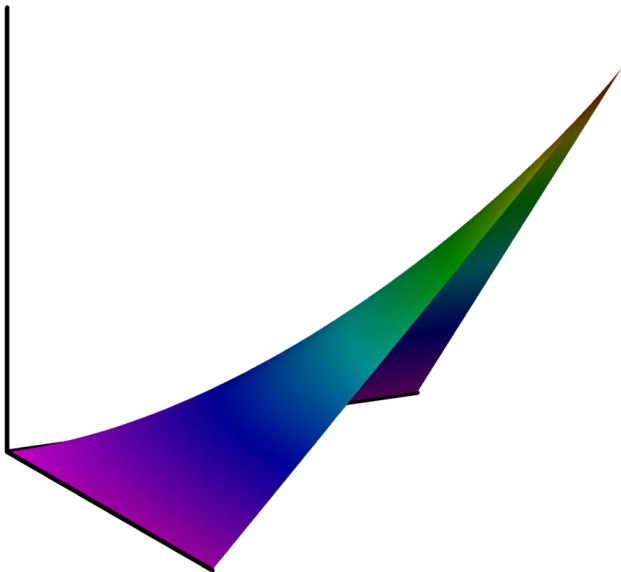
This is what J looks like ($\rho = 0.1$)



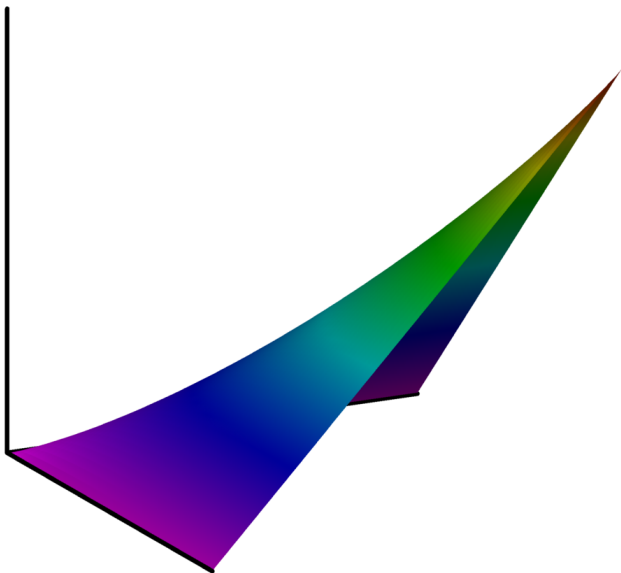
This is what J looks like ($\rho = 0.3$)



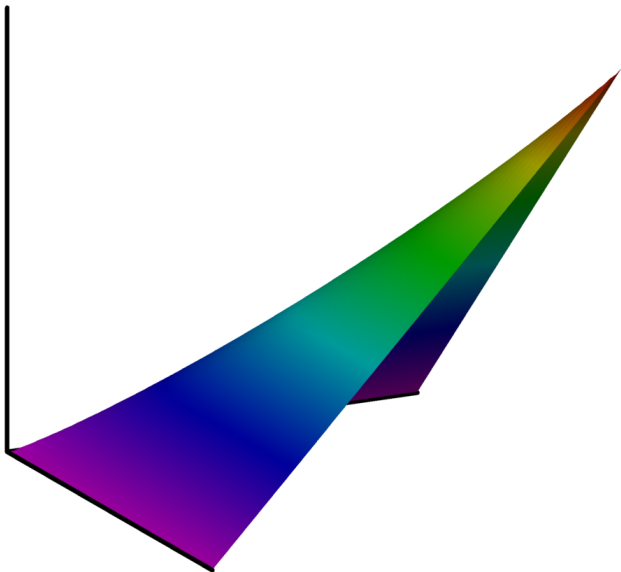
This is what J looks like ($\rho = 0.5$)



This is what J looks like ($\rho = 0.7$)



This is what J looks like ($\rho = 0.9$)



Proof: the equality case

Claim: if $f = 1_A$ and $\mathbb{E}J(f(X), f(Y)) = J(\mathbb{E}f, \mathbb{E}f)$ then A is a half-space.

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$$\mathbb{E}J(f, f) = J(\mathbb{E}f, \mathbb{E}f) \iff \forall t \nabla v_t(X) = \nabla w_t(Y) = \text{constant}$$

$$\iff P_t f(x) = \Phi(a(t) \cdot x + b(t))$$

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$$w_t = w_t(Y) = \Phi^{-1}(P_t f(Y))$$

$$\mathbb{E}J(f, f) = J(\mathbb{E}f, \mathbb{E}f) \iff \forall t \nabla v_t(X) = \nabla w_t(Y) = \text{constant}$$

$$\iff P_t f(x) = \Phi(a(t) \cdot x + b(t))$$

$$\iff \text{if } f = 1_A \text{ then } A \text{ is a half-space. } \square$$

Proof: robustness

Claim: if $f = 1_A$ and $\mathbb{E}J(f(X), f(Y)) \geq J(\mathbb{E}f, \mathbb{E}f) - \delta$ then A is almost a half-space.

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$$\begin{aligned} & J(\mathbb{E}f, \mathbb{E}f) - \mathbb{E}J(f(X), f(Y)) \\ &= \frac{\rho}{2\pi\sqrt{1-\rho^2}} \int_0^\infty \mathbb{E}e^{-(v_t^2+w_t^2-2\rho v_t w_t)} |\nabla v_t - \nabla w_t|^2 dt. \end{aligned}$$

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Lemma

For any $t > 0$, $P_t f$ is close to a function of the form $\Phi(a \cdot x + b)$.

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Lemma

If $P_t f$ is close to a function of the form $\Phi(a \cdot x + b)$ then f is also close to a function of the same form.

