

# Convex Poincaré inequality

(and weak transportation inequalities)

Based on joint work with Radosław Adamczak.

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# Setting

Standing assumption:

$\mu$  satisfies the *convex Poincaré inequality*, i.e.  $\forall$  convex  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ ,

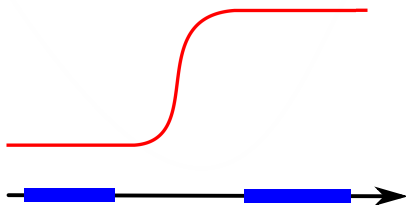
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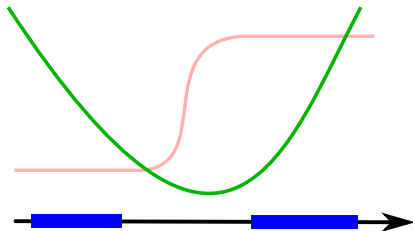


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**Goal:**  $\forall$  **convex** or **concave**  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  with  $|\nabla f| \leq c$ ,

$$\mathbb{E}_\mu f e^f - \mathbb{E}_\mu e^f \ln(\mathbb{E}_\mu e^f) =: \text{Ent}_\mu(e^f) \leq C \mathbb{E}_\mu |\nabla f|^2 e^f.$$

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- Implies two-level concentration:

$$\mathbb{P}_{\mu^{\otimes N}}(|f - \mathbb{E}_{\mu^{\otimes N}} f| \geq t) \leq \exp\left(-C_1 \frac{t^2}{L_2^2(f)} \wedge \frac{t}{L_1(f)}\right).$$

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- Convex case,  $n = 1$ : Feldheim-Marsiglietti-Nayar-Wang '15, Gozlan-Roberto-Samson-Shu-Tetali '15.



# Main result

## Theorem (Adamczak-St. '17)

$\forall$  convex or concave  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  with  $|\nabla f| \leq c < \sqrt{2\lambda}/e$ ,

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$$F(1) = \mathbb{E}_\mu f^2 e^f \lesssim \mathbb{E}_\mu |\nabla f|^2 e^f.$$

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## Question

If  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  is convex and 1-Lipschitz, then, for  $t \geq 0$ ,

$$\mathbb{P}_\mu(f \geq \text{Med}_\mu f + t) \leq 2 \exp(-C(\lambda)t).$$

Do we also have

$$\mathbb{P}_\mu(f \leq \text{Med}_\mu f - t) \stackrel{?}{\leq} 2 \exp(-C(\lambda)t) ?$$