

Efron-Stein, Poincaré inequalities, tensorization of entropy

$X_1, \dots, X_n$ independent r.v.'s. How to bound $\text{Var } f(X_1, \dots, X_n)$?

Then (Efron-Stein): If $X_1, \dots, X_n$ independent,

$Z = f(X_1, \dots, X_n)$ with $\text{Var } Z < \infty$, then

$$\text{Var } Z \leq \sum_{i=1}^n \mathbb{E} \text{Var}^{(i)} Z$$

where $\text{Var}^{(i)} Z = \text{Var} [Z \mid \underbrace{X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_n}_{\equiv X^{(i)}}]$

This is a "tensorization" property of variance, and reduces bounds for $\text{Var } Z$ to variance bounds for univariate functions.

Proof: Recall the martingale

$$M_i = \mathbb{E}[Z \mid X_1, \dots, X_i], \quad \Delta_i = M_i - M_{i-1}$$

Since $\mathbb{E} \Delta_i \Delta_j = 0 \quad \forall i \neq j$,

$$\text{Var } Z = \mathbb{E} (Z - \mathbb{E} Z)^2 = \mathbb{E} \left(\sum_{i=1}^n \Delta_i \right)^2 = \sum_{i=1}^n \mathbb{E} \Delta_i^2.$$

$$\begin{aligned} \text{Apply } \Delta_i^2 &= \left(\mathbb{E}[Z - \mathbb{E}^{(i)} Z \mid X_1, \dots, X_i] \right)^2 \\ &\leq \mathbb{E} \left[(Z - \mathbb{E}^{(i)} Z)^2 \mid X_1, \dots, X_i \right] \end{aligned}$$

$$\Rightarrow \text{Var } Z \leq \sum_{i=1}^n \mathbb{E} (Z - \mathbb{E}^{(i)} Z)^2 = \sum_{i=1}^n \mathbb{E} V_n^{(i)} Z.$$

Equivalent forms: Set $v = \sum_{i=1}^n \mathbb{E} V_n^{(i)} Z$.

① Recall $\text{Var } Y = \frac{1}{2} \mathbb{E} (Y - Y')^2$ for Y' an independent copy of Y .

$$\Rightarrow v = \frac{1}{2} \sum_{i=1}^n \mathbb{E} (Z - Z'_i)^2, \text{ where}$$

$$Z'_i = f(X_1, \dots, X_{i-1}, \underbrace{X'_i}_{\text{independent copy of } X_i}, X_{i+1}, \dots, X_n)$$

② By symmetry $Z - Z'_i \stackrel{d}{=} -(Z - Z'_i)$, so

$$v = \sum_{i=1}^n \mathbb{E} (Z - Z'_i)_+^2 = \sum_{i=1}^n \mathbb{E} (Z - Z'_i)_-^2.$$

Example: Suppose f has bounded differences property, $\|D_i f\|_\infty < \infty$.

Then $\text{Var } Z \leq \sum_{i=1}^n \|D_i f\|_\infty^2$. (Weaker than tail bound of Lecture 1.)

Example: Recall Rademacher complexity: $\varepsilon_1, \dots, \varepsilon_n \stackrel{\text{i.i.d.}}{\sim} \text{Rademacher}$,

$$Z = f(\varepsilon_1, \dots, \varepsilon_n) = \sup_{t \in T} \varepsilon^T t, \quad T \subseteq \mathbb{R}^n.$$

Let $t^* \equiv t^*(\varepsilon)$ be point where sup is attained, for ε .

$$Z - Z'_i \leq \varepsilon^T t^* - (\varepsilon_1, \varepsilon'_i, \dots, \varepsilon_n)^T t^* = (\varepsilon_i - \varepsilon'_i) t_i^*$$

$$\Rightarrow \mathbb{E} (Z - Z'_i)_+^2 \leq \mathbb{E} (\varepsilon_i - \varepsilon'_i)_+^2 t_i^{*2} \leq \mathbb{E} Z t_i^{*2}$$

$$\Rightarrow \text{Var } Z \leq \sum_{i=1}^n \mathbb{E} (Z - Z_i)_+^2 \leq 2 \mathbb{E} \|t^*\|_2^2 \leq 2 \sup_{t \in T} \|t\|_2^2.$$

(Better than bounded differences bound.)

Example (1st passage percolation): G graph w/ iid edge weights

$X_i \geq 0$, $\mathbb{E} X_i^2 = \sigma^2$. Fixing $u, v \in G$, let

$$Z = \inf_{\text{paths } P: u \rightarrow v} \sum_{e_i \in P} X_i.$$

Let P^* be path where inf is attained, for X_i .

$$(Z - Z_i')_- \leq (X_i' - X_i)_+ \mathbb{I}\{e_i \in P^*\} \leq X_i' \cdot \mathbb{I}\{e_i \in P^*\}$$

$$\Rightarrow \text{Var } Z \leq \sum_{i=1}^n \mathbb{E} (Z - Z_i')_-^2 \leq \sigma^2 \mathbb{E} [\text{length of } P^*].$$

[If e.g. $G = \mathbb{Z}^d$, $X_i \in [a, b]$ a.s., then

$$\text{length of } P^* \leq \frac{b}{a} \|u - v\|_1, \text{ while } \mathbb{E} Z \geq a \|u - v\|_1.]$$

Example (configuration functions): Call a property Π of

$(x_1, \dots, x_n)$ hereditary if

$(x_1, \dots, x_n)$ satisfies $\Pi \Rightarrow$ all subsequences of $(x_1, \dots, x_n)$ satisfy Π .

E.g. ① $x_1, \dots, x_n$ are distinct

② $x_1, \dots, x_n$ are increasing

③ $x_1, \dots, x_n$ are shattered by a collection of sets $\mathcal{A}$:

$$\forall S \subseteq \{x_1, \dots, x_n\}, \exists A \in \mathcal{A} \text{ s.t. } A \cap \{x_1, \dots, x_n\} = S.$$

Let $Z = f(x_1, \dots, x_n)$ be longest subsequence of $(x_1, \dots, x_n)$ satisfying Π .

① # of distinct elements

② longest increasing subsequence

③ VC dimension of $\mathcal{A}$ (restricted to $\{x_1, \dots, x_n\}$)

Any such f is self-bounding:

$$0 \leq f(x_1, \dots, x_n) - f(x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_n) \leq 1.$$

$$\sum_{i=1}^n f(x_1, \dots, x_n) - f(x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_n) \leq f(x_1, \dots, x_n).$$

$$\text{Note } \text{Var}^{(i)} Z \leq \mathbb{E}^{(i)} (f(x) - f(x^{(i)}))^2$$

$$\begin{aligned} \Rightarrow \text{Var } Z &\leq \mathbb{E} \sum_{i=1}^n (f(x) - f(x^{(i)}))^2 \\ &\leq \mathbb{E} \sum_{i=1}^n f(x) - f(x^{(i)}) \leq \mathbb{E} f(x) = \mathbb{E} Z. \end{aligned}$$

Poincaré inequalities

For certain distributions X_i and functions $f: \mathbb{R} \rightarrow \mathbb{R}$,

$$\text{Var } f(X_i) \leq C \cdot \mathbb{E}[\text{"squared gradient" of } f(X_i)]$$

Then: Let $X_1, \dots, X_n \stackrel{\text{iid}}{\sim}$ Rademacher, $f: \{\pm 1\}^n \rightarrow \mathbb{R}$. Define its discrete gradient $\text{grad } f(X_1, \dots, X_n) = (D_i f(X_1, \dots, X_n))_{i=1}^n$,

$$D_i f(X_1, \dots, X_n) = \frac{f(X_1, \dots, X_{i-1}, 1, X_{i+1}, \dots, X_n) - f(X_1, \dots, X_{i-1}, -1, X_{i+1}, \dots, X_n)}{2}$$

$$\text{Then } \text{Var } f(X_1, \dots, X_n) \leq \mathbb{E} \|\text{grad } f(X_1, \dots, X_n)\|_2^2$$

Proof: In 1-dimension ($n=1$):

$$\text{Var } f(X) = \left(\frac{f(+1) - f(-1)}{2} \right)^2 = \mathbb{E} (\text{grad } f(X))^2$$

For general n , by tensorization (Efron-Stein),

$$\begin{aligned} \text{Var } f(X_1, \dots, X_n) &\leq \sum_{i=1}^n \mathbb{E} \underbrace{\text{Var}^{(i)} f(X_1, \dots, X_n)}_{= \mathbb{E}^{(i)} D_i f(X_1, \dots, X_n)^2 \text{ by } n=1 \text{ case}} = \mathbb{E} \|\text{grad } f(X_1, \dots, X_n)\|_2^2 \quad \square \end{aligned}$$

Example: $\varepsilon_1, \dots, \varepsilon_n \stackrel{\text{iid}}{\sim}$ Rademacher, $Z = f(\varepsilon) = \sup_{t \in T} \varepsilon^T t$.

Previous argument shows $D_i f(\varepsilon)_+ \leq |t_{i+}^*(\varepsilon)|$

By symmetry in law, $E(D_i f(x))_+^2 = E(D_i f(x))_-^2 = \frac{1}{2} E D_i^2 f(x)^2$,

$$\text{so } \text{Var } f(x) \leq E \| \text{grad } f(x) \|_2^2 \leq 2 E \| \nabla f(x) \|_2^2 \leq 2 \sup_{x \in T} \| \nabla f \|_2^2$$

Then (Gaussian Poincaré inequality): Let $X_1, \dots, X_n \stackrel{\text{i.i.d.}}{\sim} \mathcal{N}(0, 1)$,

$f: \mathbb{R}^n \rightarrow \mathbb{R}$ weakly differentiable. Then

$$\text{Var } f(X_1, \dots, X_n) \leq E \| \nabla f(X_1, \dots, X_n) \|_2^2$$

Proof: In 1-dimension ($n=1$): Assume $f: \mathbb{R} \rightarrow \mathbb{R}$ has compact support, twice continuously-differentiable, so $\|f\|, \|f'\|, \|f''\| \leq K$.

Introduce $\varepsilon_1, \dots, \varepsilon_m \stackrel{\text{i.i.d.}}{\sim}$ Rademacher, $S_m = \frac{\varepsilon_1 + \dots + \varepsilon_m}{\sqrt{m}}$.

By previous Poincaré ineq. on hypercube,

$$\begin{aligned} \text{Var } [f(S_m)] &\leq \sum_{i=1}^m E D_i^2 f(S_m)^2 \\ &= \sum_{i=1}^m \frac{1}{4} E \left(f(S_m) - f\left(S_m - \frac{2\varepsilon_i}{\sqrt{m}}\right) \right)^2 \\ &\leq m \cdot \frac{1}{4} E \left[\left(\frac{2}{\sqrt{m}} |f'(S_m)| + \frac{2K}{m} \right)^2 \right] \end{aligned}$$

Since $S_m \xrightarrow{L} X$, by dominated convergence

$$\text{Var } f(X) = \lim_{m \rightarrow \infty} \text{Var } f(S_m) \leq \lim_{m \rightarrow \infty} E f'(S_m)^2 = E f'(X)^2$$

For any weakly differentiable (i.e. absolutely cont.) $f: \mathbb{R} \rightarrow \mathbb{R}$,
approximate by sequence of smooth, compactly supported functions.

For general n : By tensorization,

$$\begin{aligned} \text{Var } f(X_1, \dots, X_n) &\leq \sum_{i=1}^n \mathbb{E} \underbrace{\text{Var}^{(i)} f(X_1, \dots, X_n)} \leq \mathbb{E} \|\nabla f(X_1, \dots, X_n)\|_2^2 \\ &\leq \mathbb{E}^{(i)} \partial_i f(X_1, \dots, X_n)^2 \text{ by result for } n=1. \quad \square \end{aligned}$$

Cor: If f is L -Lipschitz (i.e. $|f(x) - f(x')| \leq L \cdot \|x - x'\|_2$)
and $X_1, \dots, X_n \stackrel{\text{i.i.d.}}{\sim} \mathcal{N}(0, 1)$, then $\text{Var } f(X_1, \dots, X_n) \leq L^2$.

Note: This is dimension free, i.e. there is no explicit dependence on n .

Example (Gaussian complexity): Let $g \sim \mathcal{N}(0, I) \in \mathbb{R}^n$,

$$f(g) = \sup_{t \in T} g^T t \text{ for some } T \subseteq \mathbb{R}^n.$$

For each fixed t : $|g^T t - g'^T t| \leq \|g - g'\|_2 \|t\|_2$, so

$g \mapsto g^T t$ is $\|t\|_2$ -Lipschitz

$$\Rightarrow \left| \sup_{t \in T} g^T t - \sup_{t \in T} g'^T t \right| \leq \sup_{t \in T} |g^T t - g'^T t| \leq \|g - g'\|_2 \cdot \sup_{t \in T} \|t\|_2$$

$$\text{So } \text{Var } f(g) \leq \sup_{t \in T} \|t\|_2^2.$$

Thm (convex Poincaré ineq.): Let $X_1, \dots, X_n$ be independent,
 $X_i \in [0, 1]$ a.s. Let $f: \mathbb{R}^n \rightarrow \mathbb{R}$ be weakly differentiable
 and separately convex (i.e. convex in each coordinate $1 \rightarrow n$).

$$\text{Thm } \text{Var } f(X_1, \dots, X_n) \leq \mathbb{E} \|\nabla f(X_1, \dots, X_n)\|_2^2.$$

Proof: For $n=1$: Let $x^* = \arg\min_{x \in [0, 1]} f(x)$.

$$\begin{aligned} \text{Var } f(x) &\leq \mathbb{E} (f(x) - f(x^*))^2 \\ &\stackrel{(\text{convexity})}{\leq} \mathbb{E} [f'(x)^2 (x - x^*)^2] \leq \mathbb{E} f'(x)^2. \end{aligned}$$

Then by tensorization,

$$\begin{aligned} \text{Var } f(X_1, \dots, X_n) &\leq \sum_{i=1}^n \mathbb{E} \underbrace{\text{Var}^{(i)} f(X_1, \dots, X_n)}_{\leq \mathbb{E}^{(i)} \partial_i f(X_1, \dots, X_n)^2} \leq \mathbb{E} \|\nabla f(X_1, \dots, X_n)\|_2^2. \end{aligned}$$

Example: Let $X \in \mathbb{R}^{m \times n}$ have independent entries, $X_{ij} \in [a, b]$.

$$\text{Let } f(X) = \sigma_{\max}(X) = \sup_{u, v: \|u\|_2 = \|v\|_2 = 1} u^T X v.$$

$$\bullet |u^T X v - u^T X' v| \leq \|u\|_2 \|X - X'\|_{\text{op}} \|v\|_2 \leq \|X - X'\|_F$$

$\Rightarrow X \mapsto u^T X v$ is 1-Lipschitz $\Rightarrow \sigma_{\max}(X)$ is 1-Lipschitz

$\bullet \sigma_{\max}(X)$ is supremum of linear functions, hence convex.

Then $\sigma_{\max}(X)$ is a convex, $(b-a)$ -Lipschitz function of $\{z_{ij}\}$ where $z_{ij} = \frac{X_{ij}-a}{b-a} \in [0,1]$

$$\Rightarrow \forall x \quad \sigma(X) \leq (b-a)^2.$$

Tensorization of Entropy

Def: For a nonnegative r.v. Z w/ $\mathbb{E} Z \log Z < \infty$, its entropy is

$$\text{Ent } Z = \mathbb{E} Z \log Z - (\mathbb{E} Z) \log (\mathbb{E} Z) \quad (\geq 0 \text{ by Jensen's})$$

Interpretation: Suppose Z is defined on (Ω, P) w/ $Z \geq 0$. If

$\mathbb{E} Z = 1$, then can interpret $Z(\omega) = \frac{dQ}{dP}(\omega)$ where

$$Q(A) = \mathbb{E}[Z(\omega) \mathbb{1}\{\omega \in A\}]. \quad \text{In this case}$$

$$\text{Ent } Z = \mathbb{E} Z \log Z$$

$$= \int \left[\frac{dQ}{dP}(\omega) \log \frac{dQ}{dP}(\omega) \right] dP(\omega)$$

$$= \int dQ(\omega) \log \frac{dQ}{dP}(\omega) = D_{KL}(Q \| P).$$

For $\mathbb{E} Z = c \neq 1$, let $Z = c \tilde{Z}$ where $\mathbb{E} \tilde{Z} = 1$. Then

$$\text{Ent } Z = \mathbb{E} c \tilde{Z} \log c \tilde{Z} - \mathbb{E} c \tilde{Z} \log \mathbb{E} c \tilde{Z} = c \cdot \text{Ent } \tilde{Z}.$$

So Ent is a homogeneous extension of KL-divergence.

$$[\text{Comp: } V_{\mathcal{N}} Z = \mathbb{E} Z^2 - (\mathbb{E} Z)^2]$$

If $Z = \frac{dQ}{dP}(\omega)$, then $\mathbb{E} Z = 1$ and

$$V_{\mathcal{N}} Z = \int \left(\frac{dQ}{dP}(\omega) \right)^2 dP(\omega) - 1 = \chi^2(Q \| P).$$

So $V_{\mathcal{N}}$ is a homogeneous extension of χ^2 -divergence.]

Thm: If $X_1, \dots, X_n$ independent, $Z = f(X_1, \dots, X_n)$ nonnegative w/ $\mathbb{E} Z \log Z < \infty$, then

$$\text{Ent } Z \leq \sum_{i=1}^n \mathbb{E} \text{Ent}^{(i)} Z$$

$$\begin{aligned} \text{where } \text{Ent}^{(i)}(Z) &= \text{Ent}[Z | X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_n] \\ &= \mathbb{E}^{(i)} Z \log Z - \mathbb{E}^{(i)} Z \log \mathbb{E}^{(i)} Z \end{aligned}$$

Lemma (duality): Let $Z \geq 0$ be a r.v. (Ω, P) , $\mathbb{E} Z \log Z < \infty$. Then

$$\begin{aligned} \text{Ent } Z &= \sup_{U: \Omega \rightarrow [-\infty, \infty)} \mathbb{E} U Z \\ &\quad \text{s.t. } \mathbb{E} e^{U-1} = 1 \end{aligned}$$

Proof: By homogeneity, assume $\mathbb{E} Z = 1$.

Consider any $U: \Omega \mapsto [-\infty, \infty)$ s.t. $\mathbb{E} e^U = 1$.

Let $Q(A) = \mathbb{E}[e^{U(\omega)} \mathbb{I}\{\omega \in A\}]$, so $\frac{dQ}{dP} = e^U$.

Then $1 = \mathbb{E}_P Z = \mathbb{E}_Q e^{-U} Z$, and

$$\begin{aligned} \text{Ent } Z &= \mathbb{E}_P Z \log Z = \mathbb{E}_Q e^{-U} Z \log Z \\ &= \mathbb{E}_Q e^{-U} Z \log e^{-U} Z + \mathbb{E}_Q e^{-U} Z U \\ &= \text{Ent}_Q [e^{-U} Z] + \mathbb{E}_P U Z \end{aligned}$$

$$\Rightarrow \text{Ent } Z - \mathbb{E} U Z = \text{Ent}_Q [e^{-U} Z] \geq 0.$$

$$\Rightarrow \text{Ent } Z \geq \sup_{U: \mathbb{E} e^U = 1} \mathbb{E} U Z. \text{ Equality holds for } e^U = Z. \quad \square$$

Remark: The supremum can be restricted to $U = f(Z)$ since equality holds for such U .

Cor (Donsker-Varadhan): For any P, Q on Ω w/ $Q \ll P$,

$$D_{KL}(Q \| P) = \sup_{U: \Omega \mapsto [-\infty, \infty) \text{ s.t. } \mathbb{E}_P e^U < \infty} \mathbb{E}_Q U - \log \mathbb{E}_P e^U$$

Proof: Let $Z = \frac{dQ}{dP}(\omega)$. Then

$$D_{KL}(Q \| P) = \text{Ent}_P Z = \sup_{U: \mathbb{E}_P e^U = 1} \mathbb{E}_P U Z$$

$$= \sup_{U': \mathbb{E}_P e^{U'} < \infty} \mathbb{E}_P Z(U' - \log \mathbb{E}_P e^{U'}) \quad (e^{U'} = e^{U'} / \mathbb{E}_P e^{U'})$$

$$= \sup_{U': \mathbb{E}_P e^{U'} < \infty} \mathbb{E}_P U' - \log \mathbb{E}_P e^{U'}.$$

□

Proof of Thm: Let $Z = f(X_1, \dots, X_n)$, $M_i = \mathbb{E}[Z | X_1, \dots, X_i]$.

$$\text{Ent } Z = \mathbb{E} Z (\log Z - \log \mathbb{E} Z)$$

$$= \mathbb{E} \sum_{i=1}^n Z (\log M_i - \log M_{i-1})$$

$$\equiv \mathbb{E}^{(i)} e^{U_i} = \mathbb{E}^{(i)} \frac{M_i}{\mathbb{E}^{(i)} M_i} = 1.$$

By duality, $\text{Ent}^{(i)} Z \geq \mathbb{E}^{(i)} Z (\log M_i - \log \mathbb{E}^{(i)} M_i)$

$$= \mathbb{E}^{(i)} Z (\log M_i - \log M_{i-1}).$$

$$\Rightarrow \text{Ent } Z \leq \mathbb{E} \sum_{i=1}^n \text{Ent}^{(i)} Z.$$

□