

Lecture 20

(1)

$$\alpha(t) = \sup_{A: P(A) \geq 1/2} P(A_t^c)$$

Levy's inequalities: For a Lipschitz function f and $Z=f(X)$,

$$P(Z \geq \mathbb{E}Z + t) \leq \alpha(t)$$

$$P(Z \leq \mathbb{E}Z - t) \leq \alpha(t)$$

We saw that exact isoperimetry results yield bounds for $\alpha(t)$. (e.g., Uniform distribution over S^{n-1} : $\alpha(t) \leq e^{-(n-1)t^2/2}$
Gaussian over $\mathbb{R}^n$: $\alpha(t) \leq e^{-t^2/2}$)

→ We shall see that often it is easier to get bounds on $\alpha(t)$ without an exact isoperimetry result.

[C] Bounded difference property and isoperimetry

Consider $f: \mathcal{X}^n \rightarrow \mathbb{R}$ s.t.

$$f(x) - f(y) \leq \sum_{i=1}^n c_i \mathbb{1}_{\{x_i \neq y_i\}} =: d_c(x, y)$$

Thus, f is Lipschitz under $d_c(x, y)$.

$$\begin{aligned} \text{For this metric, } \alpha(t) &= \sup_{A: P(A) \geq 1/2} P(A_t^c) \\ &= \sup_{A: P(A) \geq 1/2} P(d_c(X, A) > t) \end{aligned}$$

Theorem. For every $c = (c_1, \dots, c_n)$ and $A \subseteq \mathcal{X}^n$
 $P(A) \geq 1/2$, $P(d_c(X, A) \geq t) \leq e^{-t^2/2 \|c\|_2^2}$.

Corollary 1. For $d_c(x, y)$, $\alpha(t) \leq 2 \cdot e^{-t^2/2\|c\|_2^2}$. (2)

Thus, for a function f satisfying BDP with ϵ ,

$$\mathbb{P}(Z > \mathbb{E}Z + t) \leq 2 e^{-t^2/2\|c\|_2^2}$$

$$\mathbb{P}(Z < \mathbb{E}Z - t) \leq 2 e^{-t^2/2\|c\|_2^2}$$

Corollary 2. (Blowing-up Lemma)

For Hamming distance $d_H(x, y)$,

$$\mathbb{P}(d_H(X, A) > nt) \leq e^{-n \left(\frac{t^2}{2} - \frac{1}{n} \log \frac{1}{P(A)} \right)}.$$

In particular, for every $\beta_n \rightarrow 0$, there exists $t_n, \delta_n \downarrow 0$ s.t.

$$\mathbb{P}(X \in A) \geq \exp(-n\beta_n) \quad \forall n \text{ large}$$

$$\Rightarrow \mathbb{P}(d_H(X, A) > nt_n) \leq \delta_n.$$

Proof. Choose $t^2 = 2(\beta_n + \log n/n)$. Then,

$$\mathbb{P}(d_H(X, A) > n\sqrt{2(\beta_n + \log n/n)}) \leq \frac{1}{n}. \quad \square$$

Proof of Theorem

We shall show:

$$\mathbb{P} \left(d_c(X, A) \geq t + \sqrt{\frac{\|c\|_2^2}{2} \log \frac{1}{P(A)}} \right) \leq e^{-2t^2/\|c\|_2^2} \quad (\#)$$

Then, denoting $u = \sqrt{\frac{\|c\|_2^2}{2} \log \frac{1}{P(A)}}$, for every $t \geq u$,

$$\mathbb{P}(d_c(X, A) \geq t) \leq e^{-2(t-u)^2/\|c\|_2^2}.$$

Since $(t-u)^2 \geq t^2/4$ for every $t \geq 2u$,

$$P(d_c(X, A) \geq t) \leq e^{-t^2/2\|c\|_2^2}, \quad \forall t \geq 2u. \quad (3)$$

On the other hand, $t < 2u \Leftrightarrow P(A) \leq e^{-t^2/2\|c\|_2^2}$.

Thus, $\forall t > 0$, $P(A) P(d_c(X, A) \geq t) \leq e^{-t^2/2\|c\|_2^2}$.

We now complete the proof of (#). The proof strategy follows Talagrand and uses concentration around mean.

Thus, interestingly, conc. around mean yields conc. around median.

Indeed, noting that $g(x) = d_c(x, A)$ satisfies BDP with constants c ,

$$P(d_c(X, A) \leq \mathbb{E} d_c(X, A) - t) \leq e^{-2t^2/\|c\|_2^2}.$$

Thus, $P(d_c(X, A) \leq 0) = P(A) \leq e^{-2(\mathbb{E} d_c(X, A))^2/\|c\|_2^2}$.

Using the other direction,

$$P(d_c(X, A) > \mathbb{E} d_c(X, A) + t) \leq e^{-2t^2/\|c\|_2^2}$$

we get

$$P(d_c(X, A) > \sqrt{\frac{\|c\|_2^2}{2} \log \frac{1}{P(A)}} + t) \leq e^{-2t^2/\|c\|_2^2}. \quad \blacksquare$$

[D] Going beyond BDP: Talagrand's Convex Distance Ineq.

An examination of the proof of Levy's inequalities reveals that it does not require d to be a metric. We now greatly extend Corollary 1 above to functions which are Lipschitz in a more "relaxed" notion of distance than

$d_c(x, y)$. This distance is known as Talagrand's convex distance and is given by (4)

$$d_T(x, A) = \sup_{C: \|C\|_2 = 1} \inf_{y \in A} d_c(x, y)$$

To obtain the counterpart of Corollary 1 for $d_T(x, y)$, we need the counterpart of the Theorem above for it. This is precisely the Talagrand's convex distance inequality.

Note that Theorem above says that

$$\mathbb{P}(A) \sup_{C: \|C\|_2 = 1} \mathbb{P}(d_c(X, A) > t) \leq e^{-t^2/2}.$$

The result below says that the supremum can be taken inside $\mathbb{P}$.

Theorem (Talagrand's Convex Distance Inequality)

$$\mathbb{P}(A) \mathbb{P}(d_T(X, A) \geq t) \leq e^{-t^2/4}.$$

We can now extend Corollary 1 to functions that are Lipschitz in d_T . However, d_T is not a metric: It only defines distance of a point to a set and

$$\sup_{y \in A} d_T(x, y) \neq d_T(x, A).$$

But a simple generalization of the definition of ⑤
 Lipchitz will enable the required extension. Specifically,
 we say f is Lipchitz in a "set distance" d
 if $f(x) \leq \sup_{y \in A} f(y) + d(x, A) \quad \forall A \subseteq \mathcal{X}^n$.

Then, since for set $A = \{x : f(x) \leq M f(x)\}$,
 $A_t = \{y : f(y) \leq M f(x) + t\}$, Levy's inequalities
 extend to this definition of Lipchitz. We can therefore
 obtain the following extension of Corollary 1.

Corollary 3: Let $f : \mathcal{X}^n \rightarrow \mathbb{R}$ be Lipchitz in d_T .

Then, for $X = X_1, \dots, X_n$ indep.,

$$P(f(X) > M f(x) + t) \leq 2 \cdot e^{-t^2/4}$$

and $P(f(X) < M f(x) - t) \leq 2 \cdot e^{-t^2/4}.$

Lemma. If for all $x \quad \exists c \text{ s.t. } \|c\|_2 \leq 1, c_i \geq 0,$
 s.t. $f(x) \leq f(y) + d_c(x, y) \quad \forall y,$
 then f is Lipchitz in d_T .