

Lecture 08: Applications of Talagrand Inequality

Convex Distance

- Last lecture we considered $d_H(x, y) := |\{i: x_i \neq y_i\}|$
- We generalize this notion of distance to $\alpha \in \mathbb{R}^n$ such that each of its coordinates are ≥ 0 and $\|\alpha\| = 1$, i.e., $\sum_{i=1}^n \alpha_i^2 = 1$

Definition

$$d_\alpha(x, y) := \sum_{\substack{1 \leq i \leq n \\ x_i \neq y_i}} \alpha_i$$

- We define the convex distance as

Definition (Convex Distance)

$$d_T(x, y) := \sup_{\alpha: \|\alpha\|=1} \sum_{\substack{1 \leq i \leq n \\ x_i \neq y_i}} \alpha_i$$

Talagrand Inequality

We will not prove this inequality. We simply state it and shall use it to show the concentrated behavior of the longest increasing subsequence.

Theorem (Talagrand Inequality)

$$\mathbb{P}[\mathbb{X} \in A] \cdot \mathbb{P}[d_T(\mathbb{X}, A) \geq t] \leq \exp(-t^2/4)$$

- Suppose $x = (x_1, \dots, x_n)$ and each coordinate has been sampled independently and uniformly at random over \mathbb{R}
- Let $f(x)$ represent the longest increasing subsequence of x
- Suppose $f(x) = k$
- Note that there exists $K_x \subseteq \{1, \dots, n\}$ such that the longest increasing subsequence is formed by

$$(x_{i_1}, x_{i_2}, \dots, x_{i_k})$$

- Note that if y matches x everywhere in K_x , then we must have $f(y) \geq f(x)$
- Similarly, if y matches x everywhere in K_x except at ℓ positions, then $f(y) \geq f(x) - \ell$
- In particular, we can write the following bound:

$$f(y) \geq f(x) - |\{i: i \in K_x, x_i \neq y_i\}|$$

- Let α be a vector such that $\alpha_i = 0$ if $i \notin K_x$, otherwise $\alpha_i = 1/\sqrt{f(x)}$. Note that $\|\alpha\| = 1$ and the above inequality can equivalently be written as:

$$f(y) \geq f(x) - \sqrt{f(x)}d_\alpha(x, y)$$

- Rearranging, we get $f(x) \leq f(y) + \sqrt{f(x)}d_\alpha(x, y)$
- By the definition of $d_T(\cdot, \cdot)$, we can also write that:

$$f(x) \leq f(y) + \sqrt{f(x)}d_T(x, y)$$

- Let $A_a = \{y : f(y) \leq a\}$
- Consider $y \in A_a$. The above inequality becomes:

$$f(x) \leq a + \sqrt{f(x)}d_T(x, y)$$

- Rearranging, we get:

$$d_T(x, y) \geq \frac{f(x) - a}{\sqrt{f(x)}}$$

- Suppose $f(x) \geq a + t$. Then, it implies that

$$d_T(x, y) \geq \frac{t}{\sqrt{a+t}}$$

- Therefore, we can conclude that

$$\mathbb{P} [f(\mathbb{X}) \geq a + t] \leq \mathbb{P} \left[d_T(\mathbb{X}, y) \geq \frac{t}{\sqrt{a+t}} \right]$$

- Since, this statement is true for all $y \in A_a$, the following statement is also true

$$\mathbb{P} [f(\mathbb{X}) \geq a + t] \leq \mathbb{P} \left[d_T(\mathbb{X}, A_a) \geq \frac{t}{\sqrt{a+t}} \right]$$

- Multiplying both sides by $\mathbb{P}[\mathbb{X} \in A_a]$, we get

$$\begin{aligned} & \mathbb{P}[\mathbb{X} \in A_a] \cdot \mathbb{P}[f(\mathbb{X}) \geq a + t] \\ & \leq \mathbb{P}[\mathbb{X} \in A_a] \cdot \mathbb{P}\left[d_T(\mathbb{X}, A_a) \geq \frac{t}{\sqrt{a+t}}\right] \\ & \leq \exp\left(-t^2/4(a+t)\right) \end{aligned}$$

- Suppose we choose $a = m$, where m is the median of the distribution $f(\mathbb{X})$. So, we have $\mathbb{P}[f(\mathbb{X}) \leq m] \geq 1/2$. We get:

$$\mathbb{P}[f(\mathbb{X}) \geq m + t] \leq 2 \exp\left(-t^2/4(m + t)\right)$$

- Suppose we choose $a = m - t$. Then $\mathbb{P}[f(\mathbb{X}) \geq a + t] \geq 1/2$. Now we have:

$$\mathbb{P}[\mathbb{X} \in A_a] = \mathbb{P}[f(\mathbb{X}) \leq m - t] \leq 2 \exp\left(-t^2/4m\right)$$

Definition (Configuration Function)

A function f is a c -configuration function, if for every x, y , there exists α such that the following holds.

$$f(y) \geq f(x) - \sqrt{c \cdot f(x)} d_\alpha(x, y)$$

Note that the longest increasing subsequence function is an 1-configuration function. The derivation of the previous concentration bound on the longest increasing subsequence naturally generalizes to c -configuration functions.