

Lecture 06: Martingale Inequalities and Talagrand Inequality

This lecture note closely follows the presentation of Chapter 3 and 4 of “Concentration,” by Colin McDiarmid ([link](#))

The lecture assumes basic familiarity with probability spaces, σ -fields and martingales. See, for example, Section 3.3 of “Concentration,” by Colin McDiarmid ([link](#))

A Useful Lemma

All results are with respect to an implicit filtration

$$\mathcal{F}_0 \subseteq \mathcal{F}_1 \subseteq \dots \subseteq \mathcal{F}_n.$$

Lemma

Let Y_1, \dots, Y_n be a martingale difference sequence with $-a_k \leq Y_k \leq 1 - a_k$ for each k . Let $a = \sum a_k/n$ and $\bar{a} = 1 - a$. Then for any $0 \leq t \leq \bar{a}$,

$$\Pr\left(\sum Y_k \geq nt\right) \leq \left(\left(\frac{a}{a+t}\right)^{a+t} \left(\frac{\bar{a}}{\bar{a}-t}\right)^{\bar{a}-t} \right)^n$$

Let $S_n = S_{n-1} + Y_n$. Use the previous lecture to complete the proof from the following result:

$$\begin{aligned} \mathbb{E}[\exp(hS_n)] &= \mathbb{E}[\exp(hS_{n-1}) \cdot \mathbb{E}[\exp(hY_n | \mathcal{F}_{n-1})]] \\ &\leq \mathbb{E}[\exp(hS_{n-1}) ((1 - a_n) \exp(-ha_n) + a_n \exp(h(1 - a_n)))] \\ &\leq \exp\left(-\sum ha_k\right) \cdot \prod (1 - a_k + a_k \exp(h)) \end{aligned}$$

Hoeffding–Azuma inequality

Theorem (Hoeffding–Azuma inequality)

Let Y_1, \dots, Y_n be a martingale difference sequence with $|Y_k| \leq c_k$ for each k . For any $t \geq 0$, we have:

$$\Pr\left[\sum Y_k \geq nt\right] \leq 2 \exp\left(-n^2 t^2 / 2 \sum c_k^2\right)$$

Think: How to prove this?

Theorem

Let Y_1, \dots, Y_n be a martingale difference sequence with $-a_k \leq Y_k \leq 1 - a_k$ for each k . Let $a = \sum a_k/n$.

① For any $t \geq 0$

$$\Pr\left[\left|\sum Y_k\right| \geq t\right] \leq 2 \exp(-2t^2/n)$$

② For any $\varepsilon > 0$

$$\Pr\left[\sum Y_k \geq \varepsilon an\right] \leq \exp(-\varepsilon^2 an/2(1 + \varepsilon/3))$$

③ For any $\varepsilon > 0$

$$\Pr\left[\sum Y_k \leq -\varepsilon an\right] \leq \exp(-\varepsilon^2 an/2)$$

Use the useful lemma to prove this

Talagrand's Convex Distance

- Recall definition of $d_\alpha(\cdot, \cdot)$ from the previous lecture
- $d_T(\mathbf{x}, A) = \sup\{d_\alpha(\mathbf{x}, A) : \alpha \geq 0, \|\alpha\|_2 = 1\}$

Theorem (Talagrand Inequality)

Let $\mathbf{X} = (X_1, \dots, X_n)$ be a family of independent random variables and let A be a subset of the product space. Then for any $t \geq 0$,

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

Read the proof from the literature.

Application of Talagrand Inequality

Let X_k be a uniform random number. Let $f(\mathbf{x})$ represent the length of the longest increasing subsequence of (x_1, \dots, x_n) .

- For every \mathbf{x} there exists a subset $K(\mathbf{x}) \subseteq [n]$ such that $f(\mathbf{x}) = |K(\mathbf{x})|$
- Consider any \mathbf{y}
- Note that:
$$f(\mathbf{y}) \geq |\{i \in K(\mathbf{x}) : y_i = x_i\}| = f(\mathbf{x}) - |\{i \in K(\mathbf{x}) : y_i \neq x_i\}|$$
- Let α be a $1/\sqrt{f(\mathbf{x})}$ at all indices in $K(\mathbf{x})$ and 0 elsewhere
- We have $f(\mathbf{y}) \geq f(\mathbf{x}) - \sqrt{f(\mathbf{x})}d_\alpha(\mathbf{x}, \mathbf{y})$

Definition (c -Configuration Function)

A c -configuration function f satisfies: For any \mathbf{x} and \mathbf{y} , there exists α such that

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

Concentration of c -Configuration Functions

Theorem

Let f be a c -configuration function and let m be the median for $f(\mathbf{X})$. Then for any $t \geq 0$,

$$\Pr[f(\mathbf{X}) \geq m + t] \leq 2 \exp(-t^2/4c(m + t))$$

We will prove this using Talagrand inequality

- By definition $f(\mathbf{x}) \leq f(\mathbf{y}) + \sqrt{cf(\mathbf{x})}d_\alpha(\mathbf{x}, \mathbf{y})$
- Let $A_a = \{\mathbf{y}: f(\mathbf{y}) \leq a\}$
- So, $f(\mathbf{x}) \leq a + \sqrt{cf(\mathbf{x})}d_\alpha(\mathbf{x}, \mathbf{y})$ for any $\mathbf{y} \in A_a$
- So, $f(\mathbf{x}) \leq a + \sqrt{cf(\mathbf{x})}d_\alpha(\mathbf{x}, A_a) \leq a + \sqrt{cf(\mathbf{x})}d_T(\mathbf{x}, A_a)$
- If $f(\mathbf{x}) \geq a + t$ then $d_T(\mathbf{x}, A_a) \geq \frac{f(\mathbf{x}) - a}{\sqrt{cf(\mathbf{x})}} \geq \frac{t}{\sqrt{c(a+t)}}$
- Therefore, $\Pr[f(\mathbf{X}) \geq a + t] \leq \Pr\left[d_T(\mathbf{X}, A_a) \geq \frac{t}{\sqrt{c(a+t)}}\right]$
- By Talagrand Inequality:

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

- Use $a = m$ and the fact that $\Pr[\mathbf{X} \leq m] \geq 1/2$