

Concentration ineq. Hoeffding's Ineq

Theorem 1. Let X_1, \dots, X_n be independent random variables on \mathbb{R} such that X_i is bounded in the interval $[a_i, b_i]$. Let $S_n = \sum_{i=1}^n X_i$. Then for all $t > 0$, *bad event*.

$$\Pr[S_n - \mathbb{E}[S_n] \geq t] \leq e^{-2t^2 / \sum_{i=1}^n (b_i - a_i)^2} \quad (1)$$

$$\Pr[S_n - \mathbb{E}[S_n] \leq -t] \leq e^{-2t^2 / \sum_{i=1}^n (b_i - a_i)^2} \quad (2)$$

(1) $\Pr[(S_n - \mathbb{E}[S_n] \geq t) \cup (\dots \leq -t)]$

$$\leq \Pr[\dots \geq t] + \Pr[\dots \leq -t]$$

$$\leq 2 \cdot (\text{RHS of (1)}) \quad \delta$$

$$P(A \cup B) \leq P(A) + P(B)$$

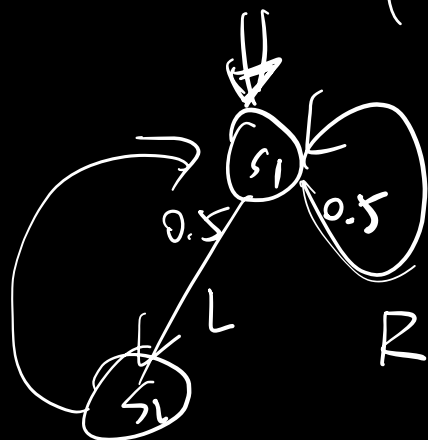
(2) Or, equivalently:

fix any $0 < \delta < 1$. *failure prob.* with prob. $\geq 1 - \delta$.

$$\left| \frac{S_n}{n} - \frac{\mathbb{E}[S_n]}{n} \right| \leq (b-a) \sqrt{\frac{1}{2n} \ln \frac{2}{\delta}}$$

(when $a_i \equiv a, b_i \equiv b$).

(3)



$s_1 \quad s_1 \quad s_2 \quad s_1 \quad s_1 \quad s_2 \quad s_1 \quad s_2 \quad \dots$

1000.
600.

$$1 \times \#R + 2 \times \#L = 1000.$$

$$\#R + \#L = 600$$

Scenario 2: run the chain until you hit s_1 600 times.

Application in bandits

For $t=1, 2, 3, \dots, T$

1. Learner chooses $i_t \in [K] := \{1, 2, \dots, K\}$.

2. Observes $r_t \sim R_{i_t}$

action space
 A

\hookrightarrow r.v. bounded in $[0, 1]$.

Alg: (1) pick each arm T/K times.

$$(2) \hat{\mu}_i$$

$$(3) \text{Output } \hat{i} = \operatorname{argmax}_{i \in [K]} \hat{\mu}_i.$$

Analysis: how good is \hat{i} ? $\mu^* - \mu_{\hat{i}} \leq ?$
 \parallel
 $\max_i \mu_i$

For any i , w.p. $\geq 1 - \delta$, $|\hat{\mu}_i - \mu_i| \leq \sqrt{\frac{1}{2T/k} \ln \frac{2}{\delta}}$

\Downarrow

w.p. $\geq 1 - \delta$, $\forall i$, $|\hat{\mu}_i - \mu_i| \leq \sqrt{\frac{k}{2T} \ln \frac{2k}{\delta}}$

$\Pr \left[\bigcup (|\hat{\mu}_i - \mu_i| > \text{value}) \right]$

$\leq \sum_{i=1}^k \Pr \left[|\hat{\mu}_i - \mu_i| > \sqrt{\frac{k}{2T} \ln \frac{2k}{\delta}} \right] = k\delta' = \delta$

$\delta' = \delta/k$

\Rightarrow w.p. $\geq 1 - \delta$, $\forall i$, $|\hat{\mu}_i - \mu_i| \leq \sqrt{\frac{k}{2T} \ln \frac{2k}{\delta}}$

$\mu^* - \mu_{\hat{i}} = \mu_{i^*} - \mu_{\hat{i}}$

\uparrow
argmax μ_i

$= \mu_{i^*} - \hat{\mu}_{i^*} + \hat{\mu}_{i^*} - \mu_{\hat{i}}$

$\leq \max_i |\mu_i - \hat{\mu}_i| + \hat{\mu}_{i^*} - \mu_{\hat{i}}$

$\max_i |\mu_i - \hat{\mu}_i|$

$$\leq 2 \cdot \sqrt{\frac{K}{2T} \ln \frac{2K}{\delta}} \quad \Delta$$

Learning theory "101". Assume \mathcal{F} is finite.

Given iid (X_i, Y_i) pairs. $Y_i \in \{0, 1\}$.

$$\mathcal{F} = \{f: X \rightarrow \{0, 1\}\}$$

$$\hat{f} = \underset{f \in \mathcal{F}}{\operatorname{argmin}} \frac{1}{n} \sum_{i=1}^n \mathbb{I}[f(X_i) \neq Y_i]$$

$$\mathbb{P}[\hat{f}(X_i) \neq Y_i] - \mathbb{P}[f^*(X_i) \neq Y_i]$$

where $f^* = \underset{f \in \mathcal{F}}{\operatorname{argmin}} \mathbb{P}[f(X_i) \neq Y_i]$.

Fix $f \in \mathcal{F}$. \leftarrow w.p. $\geq 1 - \delta$.

$$\left| \frac{1}{n} \sum_{i=1}^n \mathbb{I}[f(X_i) \neq Y_i] - \mathbb{P}[f(X_i) \neq Y_i] \right| \leq \sqrt{\frac{1}{2n} \ln \frac{2}{\delta}}$$

$$\text{w.p. } \geq 1 - \delta, \quad \forall f \in \mathcal{F}. \quad \left| \dots \right| \leq \sqrt{\frac{1}{2n} \ln \frac{2|\mathcal{F}|}{\delta}}$$

$$\begin{aligned}
& \mathbb{P}[\hat{f}(x_i) \neq y_i] - \mathbb{P}[f^*(x_i) \neq y_i] \\
&= \underbrace{\mathbb{P}[\hat{f} \dots]}_{\leq \sqrt{\frac{1}{2n} \ln \frac{2|F|}{\delta}}} - \underbrace{\hat{\mathbb{P}}[\hat{f} \dots]}_{\leq \frac{1}{n} \sum_{i=1}^n \mathbb{E}[\hat{f}(x_i) \neq y_i]} + \underbrace{\hat{\mathbb{P}}[f^* \dots]}_{\leq \frac{1}{n} \sum_{i=1}^n \mathbb{E}[f^*(x_i) \neq y_i]} - \underbrace{\mathbb{P}[f^* \dots]}_{\leq \frac{1}{n} \sum_{i=1}^n \mathbb{E}[f^*(x_i) \neq y_i]} \\
&\leq \sqrt{\frac{1}{2n} \ln \frac{2|F|}{\delta}} + \underbrace{\hat{\mathbb{P}}[f^* \dots] - \mathbb{P}[f^* \dots]}_{\leq \frac{1}{n} \sum_{i=1}^n \mathbb{E}[f^*(x_i) \neq y_i]} \\
&\leq \sqrt{\frac{2}{n} \ln \frac{2|F|}{\delta}} \dots
\end{aligned}$$

f_{θ}
 $\phi(x) \cdot \theta \geq 0 \rightarrow 1$
 $< 0 \rightarrow 0$
 $(\frac{1}{\epsilon})^d$