

Tail bounds / concentration inequalities

tell us how close r.v.'s are to their expectation

Markov Inequality

bounds right tail

very weak, but no assumptions (except ≥ 0)

Let X be a r.v. that assumes only nonnegative values

Then for all $a > 0$

$$\Pr(X \geq a) \leq \frac{E(X)}{a}$$

Proof:

$$\Pr(X \geq a) = \int_a^{\infty} f(x) dx \leq \int_a^{\infty} \frac{x}{a} f(x) dx \leq \frac{E(X)}{a}$$

$$\text{Var}(X) = E[(X - E(X))^2] = E(X^2) - [E(X)]^2$$

Chebyshev Bound

$$\forall a > 0 \quad \Pr(|X - E(X)| \geq a) \leq \frac{\text{Var}(X)}{a^2}$$

Proof:

$$\Pr(|X - E(X)| \geq a) = \Pr(\underbrace{(X - E(X))^2}_Y \geq a^2)$$

$$\leq \frac{E(Y)}{a^2} = \frac{\text{Var}(X)}{a^2}$$

Markov Inequality

Chernoff - Hoeffding Bounds for tails of sums of indep r.v.'s.

proof technique can be applied in many settings

Version 1.0

Let $X_1, \dots, X_n$ be indep r.v.'s s.t. $X_i \in [0, 1]$

Let $X = \sum_{i=1}^n X_i$ and $\mu = E(X)$ [Therefore, if $E(X_i) = p_i \Rightarrow \mu = \sum_{i=1}^n p_i$]

Chernoff-Hoeffding Upper Tail

$\forall \delta > 0$

$$\Pr(X \geq (1+\delta)\mu) \leq e^{-\mu[(1+\delta)\ln(1+\delta) - \delta]} \quad (1)$$

Weaker but useful versions:

$$\forall \delta \in [0, 1] \quad \Pr(X \geq (1+\delta)\mu) \leq e^{-\frac{\mu\delta^2}{3}} \quad (2)$$

$$\forall \delta \geq 1 \quad \Pr(X \geq (1+\delta)\mu) \leq e^{-\frac{\mu\delta}{3}} \quad (3)$$

Remark: Sufficient to have $\mu \geq E(X)$

Chernoff - Hoeffding Lower Tail

$$\forall \delta \in [0,1] \quad \Pr(X \leq (1-\delta)\mu) \leq e^{-\frac{\mu\delta^2}{2}} \quad (4)$$

Remark: Sufficient to have $\mu \leq E(X)$

Proofs:

Key property of indep r.v.'s: $E(YZ) = E(Y)E(Z)$

but we have sums.

Idea: convert sums to products using exponential fn.

Define $Y_i = e^{+tX_i}$ (t to be determined)

$$Y = e^{+X} = e^{+2X_i} = \prod_{i=1}^n e^{+X_i} = \prod Y_i$$

X_i 's indep $\Rightarrow Y_i$'s indep

$$\Rightarrow E(Y) = \prod_{i=1}^n E(Y_i)$$

Why is this useful?

$$\Pr(X \geq a) = \Pr(e^{tX} \geq e^{ta}) \underset{\forall t > 0}{\leq} \frac{E(e^{tX})}{e^{ta}} = \frac{E(Y)}{e^{ta}} = \frac{\prod E(Y_i)}{e^{ta}}$$

↑
Markov Inequality

Choose t to get best possible upper bound.

$\Rightarrow$

$$\Pr(X \geq a) \leq \min_{t > 0} \frac{E(e^{tX})}{e^{ta}}$$

Similarly, $\forall t < 0$

$$\Pr(X \leq a) = \Pr(e^{tX} \geq e^{ta}) \leq \frac{E(e^{tX})}{e^{ta}}$$

$\Rightarrow$

$$\Pr(X \leq a) \leq \min_{t < 0} \frac{E(e^{tX})}{e^{ta}}$$

$$\text{Claim: } E(e^{+X_i}) \leq 1 + p_i(e^+ - 1) \leq e^{(e^+ - 1)p_i}$$

Proof:

$$1+x \leq e^x \quad \forall x$$

$$\text{Suppose } X_i = \begin{cases} 1 & \text{w.p. } p_i \\ 0 & \text{o.w.} \end{cases}$$

$$E(e^{+X_i}) = e^+ p_i + (1-p_i)$$

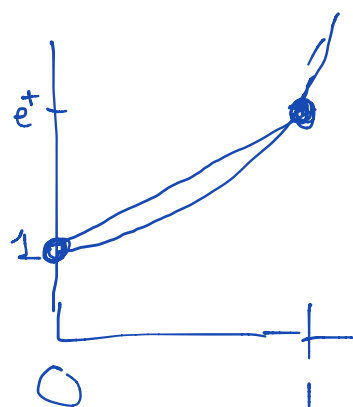
For $X_i \in [0,1]$, observe e^{+x} is convex fn of x [$2^{\text{nd}} \text{ deriv} > 0$]

$$\Rightarrow e^{+x} \leq 1 + (e^+ - 1)x \quad x \in [0,1]$$

$$\Rightarrow E(e^{+X_i}) \leq E(1 + (e^+ - 1)X_i)$$

$$= 1 + (e^+ - 1)E(X_i)$$

$$= 1 + (e^+ - 1)p_i$$



Putting it together

$$\Pr(X \geq (1+\delta)\mu) \leq \frac{E(e^{tX})}{e^{t(1+\delta)\mu}} \leq \frac{\prod_{i=1}^n e^{(e^t-1)p_i}}{e^{t(1+\delta)\mu}} \\ = e^{[(e^t-1)\sum p_i - t(1+\delta)\mu]}$$

We choose $t > 0$ to minimize $f(t) = e^{[(e^t-1) - t(1+\delta)]\mu}$

$$f'(t) = [e^t - (1+\delta)] e^{[(e^t-1) - t(1+\delta)]\mu}$$

$$f'(t) = 0 \quad \text{when } e^t = 1+\delta \quad \equiv$$

$$t = \ln(1+\delta)$$

$$f''(t) = [e^t - (1+\delta)]^2 e^{[\dots]} + e^t e^{[\dots]} > 0$$

$$\Pr(X \geq (1+\delta)\mu) \leq e^{[(e^t-1)\sum p_i - t(1+\delta)\mu]}$$

$$\text{with } t = \ln(1+\delta)$$

$$\begin{aligned}
 \Pr(X \geq (1+\delta)\mu) &\leq e^{[e^{\delta\mu(1+\delta)} - 1 - (1+\delta)\delta\mu]} \\
 &= e^{[1+\delta - 1 - (1+\delta)\delta\mu]} \\
 &= e^{-\mu[(1+\delta)\delta - \delta]}
 \end{aligned}$$



The Union Bound

Let $E_1, \dots, E_k$ be any collection of events in prob space

$$\text{Then } \Pr(E_1 \vee E_2 \vee \dots \vee E_k) \leq \sum_{i=1}^k \Pr(E_i)$$

Application #1 : Balls in Bins

n bins. Throw n balls, one at a time, independently at random into a bin (Prob $\rightarrow i^{\text{th}}$ bin $= \frac{1}{n}$)

Let $B_i = \#$ balls in i^{th} bin at end.

Theorem: $\Pr(\max_i B_i \geq e \frac{\ln n}{\ln \ln n}) \leq \frac{1}{n^c}$ for some $c > 0$

Proof: Outline: prove that prob that a particular bin so unlikely to exceed this bound that a union bound gives result.

Consider bin 1

Let $X_i = \begin{cases} 1 & i^{\text{th}} \text{ ball} \rightarrow \text{bin 1} \\ 0 & \text{o.w.} \end{cases}$

$$\Rightarrow B_1 = \sum_{i=1}^n X_i$$

$$E(X_i) = \frac{1}{n} \Rightarrow E(B_1) = 1$$

By Chernoff bound (1)

$$\Pr(B_i \geq \underbrace{(1+\delta)E(B_i)}_{1+\delta}) \leq e^{-E(B_i)[(1+\delta)\ln(1+\delta) - \delta]}$$

Set $\delta = \underbrace{\frac{e \ln n}{\ln \ln n}}_{\text{call this } C} - 1$

$$\Pr(B_i \geq C) \leq e^{-C \ln C + C - 1} \leq e^{-C(\ln C - 1)}$$

Claim: $\ln C = 1 + \ln \ln n - \ln \ln \ln n$
 $\geq 1 + \frac{\ln \ln n}{2}$

$$\frac{\ln \ln n}{2} \geq \ln \ln \ln n \quad \Leftrightarrow e^v \geq 2v \quad [v = \ln \ln \ln n]$$

$e^v = 1 + v + \frac{v^2}{2} + \dots$ trivial for $v \geq 1$

$$\Rightarrow \Pr(B_i \geq C) \leq e^{-C(\ln C - 1)}$$

$$\leq e^{-C(1 + \frac{\ln \ln n}{2} - 1)}$$

$$\leq e^{-C \frac{\ln \ln n}{2}} = e^{-\frac{e \ln n}{2}}$$

$$C = \frac{e \ln n}{\ln \ln n}$$

$$= n^{-\frac{e}{2}} \leq n^{-1.35}$$

$$\Pr(\max_i B_i \geq C) = \Pr(E_1 \vee \dots \vee E_n) \leq \sum_{i=1}^n \Pr(E_i) \leq \sum_{i=1}^n n^{-1.35}$$

E_j event that
 $B_j \geq C$

$$\leq n^{-0.35}$$

Application #2 : Congestion Minimization (randomized rounding)

Input: directed graph $G = (V, E)$

Set of pairs (s_i, t_i) $i=1, \dots, k$

Output: path P_i from $s_i \rightsquigarrow t_i$ $\forall i=1 \dots k$

such that congestion minimized

max # of paths that intersect any edge

NP-hard

We will develop an approx alg using randomized rounding

Approach:

① Solve fractional version of problem (multicommodity flow)

Vars $f(e)$: flow from $s_i \rightsquigarrow t_i$ on edge e
 $\forall e \in E$
and C

min C



$$\sum_{\substack{e \text{ s.t.} \\ e=(u \rightarrow v) \text{ for} \\ \text{some } u}} f_i(e) = \sum_{\substack{e \text{ s.t.} \\ e=(v \rightarrow w) \text{ for} \\ \text{some } w}} f_i(e) \quad \forall v \neq s_i, t_i$$

conservation of
flow

$$\sum_{\substack{e \text{ s.t.} \\ e=(s_i \rightarrow v) \\ \text{for some } v}} f_i(e) = 1$$

route 1 unit of
flow from s_i to t_i

$$\sum_i f_i(e) \leq C \quad \forall e$$

congestion bound

$$f_i(e) \geq 0 \quad \forall i, e$$

② Solve LP $\rightarrow f_i^*(e), C^*$

Observation: $C^* \leq \text{OPT}$ congestion for integer version

③ Decompose flows f^* into paths

$$\mathcal{P}_{s_i, t_i} = \{ P \mid P \text{ is a path from } s_i \text{ to } t_i \}$$

for each i , find set $\mathcal{P}_i^* \subseteq \mathcal{P}_{s_i, t_i}$ and $f_P^i \quad \forall P \in \mathcal{P}_i^*$

s.t. $|\mathcal{P}_i^*|$ polynomial

$$\sum_{P \in \mathcal{P}_i^*} f_P^i = f_i^*(e)$$

s.t. $e \in P$

$$\sum_{P \in \mathcal{P}_i^*} f_P^i = 1$$

Consider all edges with $f_i^*(e) > 0$

Find path P in graph from s_i to t_i

set $f_P^i = \min_{e \in P} f_i^*(e)$

remove f_P^i units of flow from all edges in P

repeat.

④ $\forall i \leq k$, pick path $P \in \mathcal{P}_i^*$ with prob f_P^i

call this path P_i

Output $P_1, \dots, P_k$

Theorem $\Pr(\text{any edge has congestion} \geq \underbrace{\frac{6 \log n}{\log \log n}}_{\Omega} C^*) \leq \frac{1}{n}$

Proof:

Fix $e = (u \rightarrow v)$

Let $X_i(e) = \begin{cases} 1 & \text{if } e \in P_i \\ 0 & \text{otherwise} \end{cases}$

$$E(X_i) = \sum_{P \in \mathcal{P}_i^* \text{ s.t. } e \in P} f_P^i = f_i^*(e)$$

$$\text{Let } X(e) = \sum_{i=1}^k X_i(e) \quad \Rightarrow \quad E(X(e)) = \sum_i f_i^*(e) \leq C^*$$

use Chernoff bound (1) with C^* upper bound on μ

$$\Pr(X(e) \geq (1+\delta)C^*) \leq e^{-C^*[(1+\delta)\ln(1+\delta) - \delta]}$$

$$\text{Take } (1+\delta)C^* = \alpha C^*$$

$$1 - \alpha = -\delta$$

$$\Rightarrow \Pr(X(e) \geq \alpha C^*) \leq e^{-C^*[\alpha \ln \alpha + 1 - \alpha]}$$

$$C^* \geq 1 \quad \leq e^{-\alpha(\ln \alpha - 1)}$$

$$\leq e^{-(\frac{6}{2})\ln n}$$

$$= \frac{1}{n^3}$$

$$\alpha = \frac{6 \ln n}{\ln \ln n}$$

$$\frac{6 \ln n}{\ln \ln n} \left[\ln 6 + \ln \ln n - \ln \ln \ln n - 1 \right] \geq \frac{\ln \ln n}{2} \text{ as before}$$

$$\Pr(\text{any edge has congestion} \geq \alpha C^*) \leq \sum_e \Pr(X(e) \geq \alpha C^*)$$

union bound

$$\leq n^2 \cdot \frac{1}{n^3} = \frac{1}{n}$$

□

Optimal !

$\exists$ graphs with $\frac{OPT}{C^*} = \Omega\left(\frac{\log n}{\log \log n}\right)$ (integrality gap)

so can't do better with such an approach

Hardness (directed graphs)

Every poly time alg has $\Omega\left(\frac{\log n}{\log \log n}\right)$ approx ratio

assuming $NP \not\subseteq BPTIME(n^{O(\log \log n)})$

If $C^* \geq c \ln n$ then w.h.p. congestion on

every edge $\leq C^* + \sqrt{c C^* \ln n}$

$$\Pr(X(e) \geq C^* + \sqrt{c C^* \ln n}) \leq e^{-\frac{C^* \delta^2}{3}} = e^{-\frac{c \ln n}{3}} = n^{-\frac{c}{3}}$$

$$(1+\delta)C^* \Rightarrow \delta = \sqrt{\frac{c \ln n}{C^*}}$$

Take $c=9$

$$\Pr(\text{edge } e \text{ bad}) \leq \frac{1}{n^3}$$

$$\Rightarrow \Pr(\exists \text{ bad edge}) \leq \frac{1}{n}$$

Observe that

$$C^* + \sqrt{c C_{ln}^*} = O(C^*) \quad \text{if } C^* = \mathcal{L}(lnn)$$