

1 Nov 2021

Chernoff Bound continued!

Chernoff Bound: If X_1, \dots, X_n are independent $[0, 1]$ -valued rand vars,

and if $X = X_1 + \dots + X_n$
 $\mu = \mathbb{E}[X]$.

then $\Pr(X > (1+\epsilon)\mu)$ and

$\Pr(X < (1-\epsilon)\mu)$

are both $< \exp(-\psi(\epsilon) \cdot \mu)$.

where $\psi(\epsilon) > 0$ for $\epsilon > 0$,

$\psi(\epsilon) > \frac{1}{3}\epsilon^2$ for $\epsilon < 1$.

Proof: We will focus on the quantity

$$\Phi_X(t) = \mathbb{E}[e^{tX}].$$

$$\begin{aligned}
&= \mathbb{E} \left[e^{tX_1 + tX_2 + \dots + tX_N} \right] \\
&= \mathbb{E} \left[\prod_{i=1}^N e^{tX_i} \right] \\
&= \prod_{i=1}^N \mathbb{E} \left[e^{tX_i} \right] = \prod_{i=1}^N \Phi_{X_i}(t).
\end{aligned}$$

Markov's Inequality: If Y is a rand var taking values in $[0, \infty)$

$$\Pr(Y > c \cdot \mathbb{E}Y) < \frac{1}{c}.$$

When $Y = e^{tX}$ then the event $X > (1+\varepsilon)x$ becomes

$$e^{tX} > e^{(1+\varepsilon)tX}$$

$$\begin{aligned}
\text{So } \Pr(X > (1+\varepsilon)x) &= \Pr(e^{tX} > e^{(1+\varepsilon)tX}) \\
&< \mathbb{E} \left[e^{tX} \right] \cdot e^{-(1+\varepsilon)tX}.
\end{aligned}$$

We have a battle of exponentials...
choose t s.t. the $e^{-(1+\varepsilon)tX}$ term wins!

Working out $\Phi_X(t) = \prod_{i=1}^N \mathbb{E}[e^{tX_i}]$.

For simplicity assume X_i is $\{0,1\}$ -valued.

Letting $x_i = \mathbb{E}[X_i]$, we have

w. prob. $x_i \dots X_i=1, e^{tX_i} = e^t$

w. prob. $1-x_i \dots X_i=0, e^{tX_i} = 1$

$$\begin{aligned} \int_0 \mathbb{E}[e^{tX_i}] &= 1-x_i + x_i e^t \\ &= 1 + x_i(e^t - 1) \\ &\leq e^{x_i(e^t - 1)} \end{aligned}$$

$$\begin{aligned} \prod_{i=1}^N \mathbb{E}[e^{tX_i}] &\leq e^{(\sum_{i=1}^N x_i)(e^t - 1)} \\ &= e^{x(e^t - 1)} \end{aligned}$$

$$\begin{aligned} \text{Pr}(X > (1+\epsilon)x) &\leq e^{(e^t - 1)x - (1+\epsilon)tx} \\ &= e^{[e^t - 1 - (1+\epsilon)t]x} \end{aligned}$$

Minimize $e^t - 1 - (1+\epsilon)t$ by choosing t ,

\Rightarrow set $t = \ln(1+\epsilon)$, so

$$e^t - 1 - (1+\epsilon)t = \epsilon - (1+\epsilon)\ln(1+\epsilon).$$

So we've proven Chernoff bd
with

$$\Psi(\varepsilon) = -\varepsilon + (1+\varepsilon) \ln(1+\varepsilon)$$

... for the event $X > (1+\varepsilon)\mu$.

For $\Pr(X < (1-\varepsilon)\mu)$ use $E[e^{-tX}]$,
go through same steps as above,
find out $\Psi(\varepsilon) = -\varepsilon + (1-\varepsilon) \ln(1-\varepsilon)$.

Is $\Psi(\varepsilon)$ positive? For small ε ,
use Taylor expansion...

$$\begin{aligned}\ln(1+\varepsilon) &= \varepsilon - \frac{1}{2}\varepsilon^2 + \frac{1}{3}\varepsilon^3 - \frac{1}{4}\varepsilon^4 + \dots \\ (1+\varepsilon)\ln(1+\varepsilon) &= \varepsilon + \frac{1}{2}\varepsilon^2 - \frac{1}{6}\varepsilon^3 + \frac{1}{12}\varepsilon^4 - \dots\end{aligned}$$

$$\begin{aligned}\Psi(\varepsilon) &= \frac{1}{2}\varepsilon^2 - \frac{1}{6}\varepsilon^3 + \frac{1}{12}\varepsilon^4 - \dots \\ &\geq \frac{1}{3}\varepsilon^2 \quad \text{for } 0 < \varepsilon \leq 1.\end{aligned}$$

For $\varepsilon \gg 1$, $(1+\varepsilon) \ln(1+\varepsilon)$ becomes
the leading order term in $\Psi(\varepsilon)$.

Apply to randomized rounding of disjoint paths...

k random paths, $N = k$

$$X_i = \begin{cases} 1 & \text{if edge } (u,v) \text{ in path } P_i \\ 0 & \text{o.w.} \end{cases}$$

$$X = \sum_{i=1}^k X_i = \text{load}(u,v)$$

By solving LP, we got

$$\mathbb{E}X = r \cdot c(u,v)$$

where r is as small as possible.

In original problem formulation assume $r \geq 1$. so $r \cdot c(u,v) \geq c(u,v)$.

$$\begin{aligned} \Pr(\text{load}(u,v) > (1+\epsilon)r c(u,v)) \\ < e^{-\psi(\epsilon)r c(u,v)} \leq e^{-\psi(\epsilon)c(u,v)}. \end{aligned}$$

$$\begin{aligned} \Pr(\exists \text{ an edge whose load is } > (1+\epsilon)r c(u,v)) \\ \leq m \cdot e^{-\psi(\epsilon) \min\{c(u,v)\}} \end{aligned}$$

Now choose ϵ such that

$$e^{-\Psi(\epsilon) \min\{c(u,v)\}} \leq \frac{1}{2m}.$$

Then $\Pr(\text{rounding also fails to achieve } (1+\epsilon)\text{-approximation})$

$$\leq m \cdot \frac{1}{2m} = \frac{1}{2}.$$

← Easy to test for this.

Repeat rounding step with fresh randomness as many times as necessary (≤ 2 times in expectation) and you achieve $1+\epsilon$ approx to optimum, provided

$$\Psi(\epsilon) \cdot \min\{c(u,v)\} \geq \ln(2m).$$

Two observations:

① High capacity: if $c(u,v) > \frac{3}{\epsilon^2} \ln(2m)$ for all (u,v) then

$$\Psi(\epsilon) \cdot \min\{c(u,v)\}$$

$$\geq \frac{1}{3\epsilon^2} \cdot \frac{3}{\epsilon^2} \cdot \ln(2m) = \ln(2m).$$

② Low capacity: if $c(u,v) = 1$ for

Some edge (u,v) , then we
need $\chi(\epsilon) \geq \ln(2m)$.

i.e.

$$(1+\epsilon) \ln(1+\epsilon) \geq 2 \ln(2m)$$

$$1+\epsilon \geq \frac{2 \ln(2m)}{\ln(\ln(2m))}.$$

i.e. even if edges have low
capacity, congestion $> O\left(\frac{\log m}{\log \log m}\right)$
is unlikely.