

13.9 Chernoff Bounds

Recall that the random variables X and Y are independent if the events $X=i$ and $Y=j$ are independent, that is $p(X=i \wedge Y=j) = p(X=i) \cdot p(Y=j)$

Consider a collection X_1, X_2, \dots, X_n of independent 0-1 valued (indicator) random variables.

Then with $X = \sum_{i=1}^n X_i$ we have

$$E(X) = \sum_{i=1}^n E(X_i) = \sum_{i=1}^n p_i$$

where $p_i = p(X_i=1)$

Intuition: If X_1, \dots, X_n are independent then fluctuations are likely to cancel out so that X should stay close to $E(X)$

Our goal: derive bounds on

$$p(X > E(X)) \text{ and } p(X < E(X))$$

Called Chernoff bounds after their inventor.

(13.42) Let X_1, X_2, \dots, X_n be independent 0-1 random variables, let $X = \sum X_i$ and let $p \geq E(X)$. Then $\forall \delta > 0$ we have $P[X > (1+\delta)p] < \left[\frac{e^\delta}{(1+\delta)^{1+\delta}} \right]^p$

proof: We use a sequence of transformations

$$(1) \quad \forall t > 0 \quad P[X > (1+\delta)p] = P[e^{tX} > e^{t(1+\delta)p}]$$

as e^{ty} is monotone increasing with y

(2) By Markov's inequality we have for every non-negative random variable Y and positive number γ

$$P[Y > \gamma] \leq \frac{E(Y)}{\gamma} \quad \text{so} \quad \gamma P[Y > \gamma] \leq E(Y) \quad (*)$$

Combining (1) and (*) we set

$$(3) \quad P[X > (1+\delta)p] = P[e^{tX} > e^{t(1+\delta)p}] \leq e^{t(1+\delta)p} E[e^{tX}]$$

So we need to bound $E[e^{tX}]$

$$E(e^{tX}) = E(e^{t\sum X_i}) = E(e^{\sum tX_i}) = E\left(\prod_{i=1}^n e^{tX_i}\right) = \prod_{i=1}^n E(e^{tX_i})$$

Here the last equality follows from the fact that X_1, X_2, \dots, X_n are independent

Recall that Y, Z independent $\Rightarrow E(Y \cdot Z) = E(Y) \cdot E(Z)$

$$E(e^{tx_i}) = p_i \cdot e^t + (1-p_i) \cdot e^{t \cdot 0} = p_i e^t + (1-p_i) = 1 + p_i(e^t - 1)$$

so $E(e^{tx_i}) \leq e^{p_i(e^t - 1)}$ as $1+x \leq e^x$ when $x \geq 0$

and we set

$$\begin{aligned} E(e^{tx}) &= \prod_{i=1}^n E(e^{tx_i}) \leq \prod_{i=1}^n e^{p_i(e^t - 1)} \\ &= e^{\sum p_i(e^t - 1)} \\ &= e^{(e^t - 1) \sum p_i} \\ &\leq e^{(e^t - 1)p} \text{ as } \sum p_i = E(X) \leq p \end{aligned}$$

Including this in $P[X > (1+\delta)p] \leq e^{-t(1+\delta)p} \cdot E(e^{tx})$

we set

$$P[X > (1+\delta)p] \leq e^{-t(1+\delta)p} \cdot e^{(e^t - 1)p}$$

This holds for all $t > 0$ so taking $t = \ln(1+\delta)$ we get

$$\begin{aligned} P[X > (1+\delta)p] &\leq e^{-\ln(1+\delta)(1+\delta)p} \cdot e^{(e^{\ln(1+\delta)} - 1)p} \\ &= (1+\delta)^{-1(1+\delta)p} \cdot e^{(1+\delta-1)p} \\ &= \left[\frac{e^\delta}{(1+\delta)^{(1+\delta)}} \right]^p \end{aligned}$$

□

Similarly one can show

13.43 let X_1, X_2, \dots, X_n be independent 0-1 variables

$$X = \sum_{i=1}^n X_i \text{ and let } p \geq E(X)$$

Then $\forall \delta$ with $0 < \delta < 1$ we have

$$P[X < (1-\delta)p] \leq e^{-\frac{1}{2}p\delta^2}$$

Easier formulas to use

$$P(X > (1+\delta)p) \leq e^{-\frac{\delta^2}{3}p} \quad \text{when } 0 < \delta$$

$$P(X < (1-\delta)p) \leq e^{-\frac{\delta^2}{2}p} \quad \text{when } 0 < \delta < 1$$

Example of application of Chernoff bounds

$X = \# \text{heads in } n \text{ flips of a fair coin}$

We have seen that

$$E(X) = \frac{n}{2} \quad \text{and} \quad V(X) = \frac{n}{4}$$

We want to bound the probability that
 $|X - \frac{n}{2}| \geq \frac{n}{4}$ (so $X \leq \frac{n}{4}$ or $X \geq \frac{3n}{4}$)

By Chebyshev:

$$P\left[|X - \frac{n}{2}| \geq \frac{n}{4}\right] \leq \frac{V(X)}{\left(\frac{n}{4}\right)^2} = \frac{\frac{n}{4}}{\left(\frac{n}{4}\right)^2} = \frac{4}{n}$$

By Chernoff:

$$\begin{aligned} P\left(X - \frac{n}{2} \geq \frac{n}{4}\right) &= P\left(X \geq \left(1 + \frac{1}{2}\right) \frac{n}{2}\right) \\ &\leq e^{-\left(\frac{1}{2}\right)^2 \cdot \frac{1}{3} \cdot \frac{n}{2}} = e^{-\frac{n}{24}} \end{aligned}$$

$$\begin{aligned} P\left(X - \frac{n}{2} \leq \frac{n}{4}\right) &= P\left(X \leq \left(1 - \frac{1}{2}\right) \cdot \frac{n}{2}\right) \\ &\leq e^{-\left(\frac{1}{2}\right)^2 \cdot \frac{1}{2} \cdot \frac{n}{2}} = e^{-\frac{n}{16}} \end{aligned}$$

$$\text{So } P\left[|X - \frac{n}{2}| \geq \frac{n}{4}\right] \leq e^{-\frac{n}{24}} + e^{-\frac{n}{16}} \leq 2 \cdot e^{-\frac{n}{24}}$$

Chebyshov $P\left[|X - \frac{n}{2}| \geq \frac{n}{4}\right] \leq \frac{4}{n}$

Chernoff $P\left[|X - \frac{n}{2}| \geq \frac{n}{4}\right] \leq 2 \cdot e^{-\frac{n}{24}}$

n	24	240	2400
Chebyshov	$\frac{1}{6}$	$\frac{1}{60}$	$\frac{1}{600}$
Chernoff	0,73	$9 \cdot 10^{-5}$	$7,4 \cdot 10^{-44}$

New calculation:

set $\sigma = \sqrt{\frac{6\ln n}{n}}$ then $\frac{n}{2} \cdot \sigma = \frac{1}{2} \sqrt{6\ln n}$

then by Chernoff bound

$$P\left(|X - \frac{n}{2}| \geq \frac{1}{2} \sqrt{6\ln n}\right) \leq 2 \cdot e^{-\frac{1}{3} \cdot \frac{n}{2} \cdot \frac{6\ln n}{n}}$$

$$= \frac{2}{n}$$

so very unlikely with deviations larger than $\sqrt{\frac{6\ln n}{n}}$ from $\frac{n}{2}$