

Thm: (Chernoff's bound) Let X_i be independent random variables such that $\mathbb{E}[X_i] = 0$ and $|X_i| \leq 1$ almost surely. Define $\sigma_i^2 = \mathbb{E}[X_i^2]$ and $\sigma^2 = \sum_{i=1}^n \sigma_i^2$. Then

$$\mathbb{P}\left(\sum_i X_i \geq t\right) \leq \max\left(e^{-t^2/4\sigma^2}, e^{t/\sigma^2}\right)$$

Note: Let $X = \sum_i X_i$. If applying Chebyshov's inequality:

$$\mathbb{P}(|X| > t) \leq \frac{\sigma^2}{t^2} \quad \text{"weak"}$$

Proof of Chernoff's bound:

$$\begin{aligned} \mathbb{P}\left(\sum_i X_i \geq t\right) &= \mathbb{P}\left(\lambda \sum_i X_i \geq \lambda t\right) \quad \text{for } \lambda \geq 0 \\ &= \mathbb{P}\left(e^{\lambda \sum_i X_i} \geq e^{\lambda t}\right) \\ &\leq \frac{\mathbb{E}[e^{\lambda \sum_i X_i}]}{e^{\lambda t}} \quad \text{because } e^x \text{ is monotone} \quad \text{by Markov's} \\ &= \frac{\prod_i \mathbb{E}[e^{\lambda X_i}]}{e^{\lambda t}} \end{aligned}$$

Now, for $x \in [0, 1]$, we have

$$e^x \leq 1 + x + x^2 \quad (\text{why?})$$

$$\begin{aligned} \rightarrow \mathbb{E}[e^{\lambda X_i}] &\leq 1 + \mathbb{E}[\lambda X_i] + \mathbb{E}[\lambda^2 X_i^2] \\ &= 1 + \lambda^2 \mathbb{E}[X_i^2] \\ &= 1 + \lambda^2 \sigma_i^2 \end{aligned}$$

Since $1+x \leq e^x$,
 $1 + \lambda^2 \sigma_i^2 \leq e^{\lambda^2 \sigma_i^2}$

$$\therefore \frac{\prod_i \mathbb{E}[e^{\lambda X_i}]}{e^{\lambda t}} \leq \frac{\prod_i \mathbb{E}[e^{\lambda^2 \sigma_i^2}]}{e^{\lambda t}} \\ = e^{\frac{\lambda^2 - \lambda t}{\sigma^2}}$$

\Rightarrow optimizing over $\lambda \in [0, 1]$, we get that
 $\lambda = \min(1, \frac{t}{2\sigma^2})$.

which completes the proof.

* Other useful forms:

Chebyshev's inequalities:

Let X_1, X_2, \dots, X_n be independent $\{0, 1\}$ -valued r.v.s.

Each $X_i = \begin{cases} 1 & \text{with prob. } p_i \\ 0 & \text{with prob. } 1-p_i \end{cases}$

Let $X = \sum_{i=1}^n X_i$ and $\mu = \mathbb{E}[X] = \sum_i p_i$. Then

$$P(X \geq (1+\epsilon)\mu) \leq e^{-\mu\epsilon^2/4}$$

$$P(X \leq (1-\epsilon)\mu) \leq e^{-\mu\epsilon^2/2}$$

$$\text{or } P(|X - \mu| \geq \epsilon\mu) \leq 2e^{-\mu\epsilon^2/4}.$$

• Example application: 1) coin tossing.

Suppose we have a fair coin.

Let S_n = the number of heads from the first n tosses

$$\Rightarrow \mathbb{E}[S_n] = \frac{n}{2} \text{ and } \text{Var}[S_n] = \frac{n}{4}.$$

Applying Chebyshov's inequality:

$$P\left(\left|S_n - \frac{n}{2}\right| \geq \varepsilon\right) \leq \frac{n}{4\varepsilon^2}$$

$$\Rightarrow P\left(\left|\frac{S_n}{n} - \frac{1}{2}\right| \geq \frac{\varepsilon}{n}\right) \leq \frac{n}{4\varepsilon^2}.$$

For example, take $\varepsilon = \frac{n}{4}$.

$$P\left(\left|\frac{S_n}{n} - \frac{1}{2}\right| \geq \frac{1}{4}\right) \leq \frac{4}{n}.$$

But how about Chernoff's bound?

$$P\left(\left|S_n - \frac{n}{2}\right| \geq \delta \frac{n}{2}\right) \leq 2e^{-\frac{n}{2}\delta^2/4}$$

$$\Rightarrow P\left(\left|\frac{S_n}{n} - \frac{1}{2}\right| \geq \frac{\delta}{2}\right) \leq 2e^{-\frac{n\delta^2}{8}}.$$

Take $\delta = \frac{1}{2}$,

$$P\left(\left|\frac{S_n}{n} - \frac{1}{2}\right| \geq \frac{1}{4}\right) \leq 2e^{-\frac{n}{32}}$$

Take $\delta = 2\sqrt{\ln n/n}$

$$P\left(\left|\frac{S_n}{n} - \frac{1}{2}\right| \geq \sqrt{\frac{\ln n}{n}}\right) \leq 2e^{-\frac{\ln n}{4}} = \frac{2}{n^{1/4}}$$

2) Set balancing.

Given an $n \times m$ matrix A with entries in $\{0, 1\}$, look for a vector \vec{b} with entries in $\{-1, 1\}$ that minimizes $\|A\vec{b}\|_\infty$.

In designing statistical experiments, each column of the matrix A represents a subject and each row represents a feature. Goal: the vector \vec{b} partitions the subject into two disjoint groups, so that each feature is roughly as balanced as possible between two groups.

randomized algo: choose entries of \vec{b} with
 $P(b_i = 1) = P(b_i = -1) = 1/2$
and b_i are independent.

Thm: For a random vector \vec{b} with entries chosen independently and equal prob. from the set $\{-1, 1\}$,
 $P(\|A\vec{b}\|_\infty \geq \sqrt{\log n}) \leq \frac{2}{n}$.

Pf: Exercise.