

# Theory of Computation Complexity

Tutorial 2

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# Outline

- Tail inequalities
  - Markov's inequality
  - Chebyshev inequality
  - Chernoff's bound
- Randomized communication complexity
  - About the success probability

# Markov inequality

- Let  $X$  be a nonnegative random variable. Then for all  $a > 0$ ,  $\Pr(X \geq a) \leq \frac{E[X]}{a}$ .

Proof: 
$$E[X] = \sum_i i \Pr(X=i)$$
$$= \sum_{i < a} i \Pr(X=i) + \sum_{i \geq a} i \Pr(X=i)$$
$$\geq a \sum_{i \geq a} \Pr(X=i) = a \Pr(X \geq a) \quad \blacksquare$$

# Chebyshev inequality

$$\Pr(|X - E(X)| \geq a) \leq \frac{\text{Var}[X]}{a^2}$$

Proof:

$$\Pr(|X - E(X)| \geq a) = \Pr\left(\left(X - E(X)\right)^2 \geq a^2\right)$$

$$\leq \frac{E\left[\left(X - E(X)\right)^2\right]}{a^2} \leftarrow \text{by Markov}$$

$$= \frac{\text{Var}[X]}{a^2} \blacksquare$$

Corollary  $\Pr(|X - E(X)| \geq t \cdot \sigma(X)) \leq \frac{1}{t^2}$

$$\Pr(|X - E(X)| \geq t \cdot E[X]) \leq \frac{\text{Var}[X]}{t^2 (E[X])^2}$$

# Chernoff bounds

- Let  $X_1, \dots, X_n$  be independent  $\{0,1\}$ -valued random variables, and  $p_i = \Pr[X_i = 1] = E[X_i]$ ,  $X = \sum_i^n X_i$ ,  $\mu = E[X]$ .
- Then for any  $\delta > 0$  :

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

$$\Pr[X < (1 - \delta)\mu] < \exp(-\mu\delta^2/2).$$

$$\begin{aligned}
\Pr[e^{tX} \geq e^{t(1+\delta)\mu}] &\leq \frac{1}{e^{t(1+\delta)\mu}} \cdot \mathbb{E}[e^{t\sum_i X_i}] && \text{Markov's inequality} \\
&= \frac{1}{e^{t(1+\delta)\mu}} \cdot \mathbb{E}[\prod_i e^{tX_i}] \\
&= \frac{1}{e^{t(1+\delta)\mu}} \cdot \prod_i \mathbb{E}[e^{tX_i}] && \text{independence of the } X_i \\
&= \frac{1}{e^{t(1+\delta)\mu}} \cdot \prod_i [1 + p_i(e^t - 1)] \\
&\leq \frac{1}{e^{t(1+\delta)\mu}} \cdot \prod_i e^{p_i(e^t - 1)} && 1 + x \leq e^x \quad \forall x \in \mathbb{R} \\
&= \frac{1}{e^{t(1+\delta)\mu}} \cdot e^{\mu(e^t - 1)} && \sum_i p_i = \mu \\
&= e^{-\mu(t(1+\delta) + 1 - e^t)}
\end{aligned}$$

We now want to choose  $t$  to maximize  $t(1+\delta) + 1 - e^t$ . Setting the derivative equal to 0 and solving for  $t$  gives  $t = \ln(1+\delta)$ . Plugging this value into the last line gives the stated bound.  $\square$

For  $0 \leq \delta \leq 1$ ,  $(1+\delta) \ln(1+\delta) - \delta \geq \delta^2/2$ .

So, the Chernoff bound can be restated as  $\Pr[X \geq (1+\delta)\mu] \leq e^{-\mu\delta^2/2}$  for  $\delta \in [0, 1]$ .

$$\Pr[X < (1-\delta)\mu] < \exp(-\mu\delta^2/2).$$

# Randomized communication complexity

- The randomized communication complexity actually depends on the required error probability, which we will use the subscript to specify.
- For example,  $R_\varepsilon(f)$  is the  $\varepsilon$ -error private-coin randomized communication complexity. When we don't specify the error probability, it's  $1/3$  by convention; that is,  $R(f) = R_{1/3}(f)$ .
- Like in randomized algorithms, the error probability can be decreased from a constant to an arbitrary  $\varepsilon$  by repeating the protocol  $O(\log 1/\varepsilon)$  times.

# Amplify the success probability

- Simply run the protocol (with error probability  $p$ )  $k$  times, and take a majority vote on all the  $k$  outcomes.
  - $p$  is a constant and  $p < 1/2$ .

- Then by Chernoff bound, it will give the wrong answer with probability at most

$$\Pr[X \geq k/2] = \Pr[X \geq (1 + \frac{1-2p}{2p})pk]$$

$$\leq e^{-pk(\frac{1-2p}{2p})^2/2} = \frac{1}{2^{\Omega(k)}}.$$

- Which is less than  $\varepsilon$ , when we repeat  $k = O(\log 1/\varepsilon)$  times.



# Summery

- Some tail inequalities.
- Amplify the success probability of randomized protocols.

Thanks