$\left\{ \operatorname{csci} 3|6907 \mid \operatorname{Lecture} 4 \right\}$ 

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#### ▶ Homework 2 is out, due next Wed

feel free to discuss in groups

homework must be written up individually

# PART I tail bounds

## Tail bounds, III

#### **Chernoff Bounds**

Let  $X_1, \ldots, X_n$  be *independent* r.v.'s assuming values in  $\{0, 1\}$ . Let  $X = X_1 + X_2 + \cdots + X_n$  and  $\mu = \mathbb{E}[X]$ . Then,

I. For all  $0 < \delta < 1$ ,

$$\Pr[|X - \mu| \ge \delta\mu] \le 2e^{-\mu\delta^2/3}$$

• **PROOF IDEA.** apply Markov's to non-negative r.v.  $e^{tX}$ .

$$\mathbf{E}[e^{tX}] = \prod_{i=1}^{n} \mathbf{E}[e^{tX_i}]$$

• EXAMPLE. toss *n* fair coins...

#### **Chernoff Bounds**

Let  $X_1, \ldots, X_n$  be *independent* r.v.'s assuming values in  $\{0, 1\}$ . Let  $X = X_1 + X_2 + \cdots + X_n$  and  $\mu = E[X]$ . Then,

I. For all  $0 < \delta < 1$ ,

$$\Pr[|X - \mu| \ge \delta\mu] \le 2e^{-\mu\delta^2/3}$$

**2**. For all  $0 < \delta < 1$ ,

$$\Pr[X \le (1-\delta)\mu] \le e^{-\mu\delta^2/2}$$

3. For all  $\delta > 0$ ,  $\Pr[X \ge (1+\delta)\mu] \le e^{-\frac{\mu\delta^2}{2+\delta}}$ 

- ▶ GENERALITY: Markov's ≫ Chebyshev's ≫ Chernoff (non-negative · bounded variance · independence)

## PART 2 | birthday paradox & balls-and-bins

QUESTION. What is the probability that amongst 30 people in a room, two share the same birthday? MODEL. Everyone's birthday is independently and uniformly chosen at random amongst 365 days.

ANALYSIS. Pr[all birthdays are distinct] is  $(1 - \frac{1}{365}) \cdot (1 - \frac{2}{365}) \cdot (1 - \frac{3}{365}) \cdots (1 - \frac{29}{365}) \approx 0.2937$ MORE GENERALLY... For *m* people and *n* "birthdays", it's

$$(1 - \frac{1}{n}) \cdot (1 - \frac{2}{n}) \cdot (1 - \frac{3}{n}) \cdots (1 - \frac{m-1}{n})$$
  
$$\approx \prod_{j=1}^{m-1} e^{-j/n} = e^{-m(m-1)/2n} \approx e^{-m^2/2n}$$

 $\Rightarrow$  constant prob of "collision" whenever  $m \gtrsim \sqrt{2n \ln 2}$ 

#### Chernoff Bound

For any events  $E_1, E_2$  not necessarily independent,

 $\Pr[E_1 \cup E_2] \le \Pr[E_1] + \Pr[E_2]$ 

- EXAMPLE. two types of errors: first w.p.  $\leq 0.1$ , second w.p.  $\leq 0.2$ .
- QUESTION.  $Pr[no \ errors] \ge ...$ ?
- GENERALIZATION.

 $\Pr[E_1 \cup E_2 \cup E_3 \cdots] \leq \Pr[E_1] + \Pr[E_2] + \Pr[E_3] + \cdots$ 

- *m* balls thrown into *n* bins
  - location of each ball independent and random
- ► Example: job scheduling
  - balls = tasks, bins = processors
- Quantities of interest
  - average load = expected number of balls in each bin
  - maximum load = number of balls in fullest bin
  - number of empty bins (= number of idle processors)
- $L_i$  be r.v. for # balls in Bin i

► 
$$L_i \sim B(m, \frac{1}{n})$$
, so  $E[L_i] = \frac{m}{n}$ ,  $Var[L_i] = \frac{m}{n}(1 - \frac{1}{n})$ 

#### Chernoff Bound

Let  $X_1, \ldots, X_n$  be *independent*  $\{0, 1\}$ -r.v.'s. Let  $X = X_1 + \cdots + X_n$  and  $\mu = \mathbb{E}[X]$ . Then, for all  $\delta > 0$ ,  $\Pr[X \ge (1 + \delta)\mu] \le e^{-\frac{\mu\delta^2}{2+\delta}}$ 

- Application. bounding  $\Pr[L_i \ge 2 \ln n + 1]$  for m = n
  - ► set  $\mu = 1, \delta = 2 \ln n$ , so  $\frac{\mu \delta^2}{2+\delta} \ge 2 \ln n$  $\Rightarrow \Pr[L_i \ge 2 \ln n + 1] \le e^{-2 \ln n} = \frac{1}{n^2}$
  - By union bound,  $\Pr[\bigvee_{i=1}^n (L_i \ge 2 \ln n + 1)] \le \frac{1}{n}$
  - Hence,  $\Pr[\text{maximum load} \le 2 \ln n + 1] \ge 1 \frac{1}{n}$ .
  - e.g. n = 1 million, max load is at most 30 w.h.p.

#### BETTER ANALYSIS.

$$\begin{aligned} \Pr[L_i \ge k] &= \Pr[\exists \text{ subset of } k \text{ balls all of which fall into bin } i] \\ &\leq \binom{n}{k} \cdot (1/n)^k \\ &\leq (ne/k)^k \cdot (1/n)^k = (e/k)^k \\ &\leq 1/n^2 \quad \text{ for } k \ge \frac{3\ln n}{\ln \ln n} \end{aligned}$$

BETTER BOUND.

- obtain a bound of  $O(\frac{\log n}{\log \log n})$  instead of  $O(\log n)$  for the maximum load.
- e.g. n = 1 million, max load is at most 16 w.h.p.

## **Empty Bins**

- Let X be random variable for # empty bins.
- Let  $X_i$  be r.v. indicating whether Bin *i* is empty.

▶ 
$$\Pr[X_i = 1] = (1 - \frac{1}{n})^m$$
 and  $\mathbb{E}[X] = n(1 - \frac{1}{n})^m$ .

▶ NOTE.  $X_i$  and  $X_j$  are *not* independent, e.g.  $\Pr[X_i = 1 \land X_j = 1] = (1 - \frac{2}{n})^m \neq \Pr[X_i = 1] \cdot \Pr[X_j = 1]$ 

### **Empty Bins: Variance**

- Recall  $\operatorname{Var}[X] = \operatorname{E}[X^2] \operatorname{E}[X]^2$ 
  - ►  $E[X^2] = E[(X_1 + \dots + X_n)^2] = \sum_{i=1}^n E[X_i^2] + \sum_{i \neq j} E[X_i X_j]$
  - If  $X_i \in \{0, 1\}$ , then  $E[X_i^2] = E[X_i]$
- Computing  $E[X_iX_j]$ 
  - $E[X_iX_j] = Pr[X_iX_j = 1] = Pr[X_i = 1 \land X_j = 1] = (1 \frac{2}{n})^m$
- ► Computing Var[X]
  - $E[X^2] = n(1 \frac{1}{n})^m + n(n-1)(1 \frac{2}{n})^m$
  - ► Var[X] =  $n(1 \frac{1}{n})^m + n(n-1)(1 \frac{2}{n})^m n^2(1 \frac{1}{n})^{2m}$

PART 4 | random graphs

## Random Graphs

- ▶ Random graph model  $G_{n,p}$ 
  - Distribution over undirected graphs on *n* vertices
  - Every edge occurs with probability *p*
  - Graph with given set of *m* edges has probability

$$p^m(1-p)^{\binom{n}{2}-m}$$

- Basic properties
  - Expected number of edges is p<sup>n</sup><sub>2</sub>
  - Each vertex has expected degree p(n-1)

## Threshold behavior for triangles

▶ NEXT WEEK: show that for random graph model  $G_{n,p}$ :

$$\Pr[G \text{ contains a triangle}] \xrightarrow{n \to \infty} \begin{cases} 1 & \text{if } p = \omega(\frac{1}{n}) \\ 0 & \text{if } p = o(\frac{1}{n}) \end{cases}$$

- If p grows faster than  $\frac{1}{n}$ , almost every graph contains a triangle
- If p grows slower than  $\frac{1}{n}$ , almost no graph contains a triangle
- THRESHOLD BEHAVIOR: holds for many properties, e.g. "is connected", "contains a clique of size 4", with difference choices of "1" n"