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## Lecture 22: Tail Bounds

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Slide Credit: Based on Stefano Tessaro's slides for 312 19au incorporating ideas from Alex Tsun's and Anna Karlin's slides for 312 20su and 20au

Music: Peach Pit

## Joint PMFs and Joint Range

**Definition.** Let *X* and *Y* be discrete random variables. The **Joint PMF** of *X* and *Y* is

$$p_{X,Y}(a,b) = \Pr(X = a, Y = b)$$

**Definition.** The **joint range** of  $p_{X,Y}$  is  $\Omega(X,Y) = \{(c,d) : p_{X,Y}(c,d) > 0\} \subseteq \Omega(X) \times \Omega(Y)$ 

Note that

$$\sum_{(s,t)\in\Omega(X,Y)}p_{X,Y}(s,t)=1$$

## Law of Total Expectation

**Law of Total Expectation (event version).** Let *X* be a random variable and let events  $A_1, \dots, A_n$  partition the sample space. Then,  $E[X] = \sum_{i=1}^{n} E[X|A_i] Pr(A_i)$ 

Law of Total Expectation (random variable version). Let X be a random variable and Y be a discrete random variable. Then,

$$E[X] = \sum_{y \in \Omega(Y)} E[X|Y = y] \Pr(Y = y)$$

#### **Example: Computer Failures**

Suppose your computer operates in a sequence of steps, and that at each step *i* your computer will fail with probability p (independently of other steps). Let X be the number of steps it takes your computer to fail. What is E[X]? Fails on first E[XIA]=1 X= # steps uptil first failure A = computer fails on step 1 E[XIA]= I+E[X] E[x] = E[x 1A] Pr(A) + E[x1A]Pr(A) X is memory lessness = | · P + (1+E[×])· (1-P) = 1 + (1-p)E[X] => E[X]= %

## Agenda

- Markov's Inequality
- Chebyshev's Inequality

# Tail Bounds (Idea)

Bounding the probability a random variable is far from its mean. Usually statements of the form:

 $\Pr(X \ge a) \le b$  $\Pr(|X - E[X]| \ge a) \le b$ 

Useful tool when

- An approximation that is easy to compute is sufficient
- The process is too complex to analyze exactly

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Markov's Inequality

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Example
XYO, ECXJ=4
                   P_{r}(x \ge 12) \le \frac{E[x]}{12} = \frac{4}{12} = \frac{1}{3}
```

**Theorem.** Let X be a random variable taking only non-negative values. Then, for any t > 0,



Incredibly simplistic – only requires that the random variable is non-negative and only needs you to know expectation. You don't need to know anything else about the distribution of X.

#### Markov's Inequality – Proof

**E**(

**Theorem.** Let *X* be a (discrete) random variable taking only non-negative values. Then, for any t > 0,

 $\mathbb{P}(X \ge t) \le \frac{\mathbb{E}(X)}{t}.$ 

$$X) = \sum_{x} x \cdot \mathbb{P}(X = x)$$
  
=  $\sum_{x \ge t} x \cdot \mathbb{P}(X = x) + \sum_{x < t} x \cdot \mathbb{P}(X = x)$   
 $\ge \sum_{x \ge t} x \cdot \mathbb{P}(X = x)$   
 $\ge \sum_{x \ge t} t \cdot \mathbb{P}(X = x) = t \cdot \mathbb{P}(X \ge t)$   
 $\rightarrow \mathbb{P}(X \ge t) \le \frac{\mathbb{E}\Sigma \times \mathbb{P}(X)}{t}$ 

 $\geq 0$  because  $x \geq 0$ whenever  $\mathbb{P}(X = x) \geq 0$ (takes only nonnegative values)

. . .

Follows by re-arranging terms

#### **Example – Geometric Random Variable**

#### Let *X* be geometric RV with parameter *p*

$$\mathbb{P}(X=i) = (1-p)^{i-1}p \qquad \qquad \mathbb{E}(X) = \frac{1}{p}$$

"How many times does Alice need to flip a biased coin until she sees heads, if heads occurs with probability p?

What is the probability that  $X \ge 2\mathbb{E}(X) = 2/p$ ?

Markov's inequality: 
$$\mathbb{P}(X \ge 2/p) \le \frac{\mathbb{E}(X)}{2/p} = \frac{1}{p} \cdot \frac{p}{2} = \frac{1}{2}$$

Can we do better?

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#### **Example**

(X20), E[X]=25

Suppose that the average number of ads you will see on a website is 25. Give an upper bound on the probability of seeing a website with 75 or more ads.  $P_{C}(x \ge 75) \le \frac{E \sum x}{25}$ 



$$(75) \leq \frac{EIX}{75}$$
$$= \frac{25}{75}$$
$$= \frac{1}{2}$$

# Example $\times 20$ , E[x] = 25

Suppose that the average number of ads you will see on a website is 25. Give an upper bound on the probability of seeing a website with 20 or more ads.



$$P_{r}(X \ge 20) \le \frac{E[X]}{20}$$
$$= \frac{25}{20}$$
$$= 1.25$$

Prob 5 1.25 is always true!

#### **Brain Break**



## Agenda

- Markov's Inequality
- Chebyshev's Inequality

# Chebyshev's Inequality $M_{arkov}: Pr(X \ge t) \le \frac{E[X]}{t}$

Theorem. Let X be a random variable. Then, for any 
$$t > 0$$
, $\mathbb{P}(|X - \mathbb{E}(X)| \ge t) \le \frac{\operatorname{Var}(X)}{t^2}$ . $\mathbb{P}(|X - \mathbb{E}(X)| \ge t) \le \frac{\operatorname{Var}(X)}{t^2}$ .

Proof: Define  $Z = X - \mathbb{E}(X)$   $\mathbb{E}[2^2] = \mathbb{E}[(X - \mathbb{E}[X])^2] = Var(X)$   $\mathbb{P}(|Z| \ge t) = \mathbb{P}(Z^2 \ge t^2) \le \frac{\mathbb{E}(Z^2)}{t^2} = \frac{Var(X)}{t^2}$   $|Z| \ge t \text{ iff } Z^2 \ge t^2$ Markov's inequality  $(Z^2 \ge 0)$  **Example – Geometric Random Variable** 

Let *X* be geometric RV with parameter *p* 

$$\mathbb{P}(X=i) = (1-p)^{i-1}p \qquad \mathbb{E}(X) = \frac{1}{p}$$

$$Var(X) = \frac{1-p}{p^2}$$

What is the probability that  $X \ge 2\mathbb{E}(X) = 2/p$ ?

Markov: 
$$\mathbb{P}(X \ge 2/p) \le \frac{\mathbb{E}(X)}{2/p} = \frac{1}{p} \cdot \frac{p}{2} = \frac{1}{2}$$
  
Chebyshev:  $\mathbb{P}(X \ge 2/p) \le \mathbb{P}\left(\left|X - \frac{1}{p}\right| \ge \frac{1}{p}\right) \le \frac{\operatorname{Var}(X)}{1/p^2} = 1-p$ 

Not better, unless  $p > 1/2 \otimes$ 

#### Example

# ()se Chehyshev's $V_{\alpha}(x)=\sigma^{2}=25$ Suppose that the average number of ads you will see on a website is 25 and the standard deviation of the number of ads is 5. Give an upper bound on the probability of seeing a website with 30 or more ads.

Pr(IX-E[X]]2t) < Var(X)/22

	Poll: pollev.com/hunter312	
С	а.	$0 \le p < 0.25$
_	<i>b.</i>	$0.25 \le p < 0.5$
	C.	$0.5 \le p < 0.75$
3	d.	$0.75 \le p$
	e.	Unable to compute



#### **Chebyshev's Inequality – Repeated Experiments**

"How many times does Alice need to flip a biased coin <u>until she sees heads n</u> times, if heads occurs with probability p?

$$X = #$$
 of flips until  $n$  times "heads"  
 $X_i = #$  of flips between  $(i - 1)$ -st and  $i$ -th "heads"

$$X = \sum_{i} X_{i}$$

Note:  $X_1, \ldots, X_n$  are independent and geometric with parameter p

$$\mathbb{E}(X) = \mathbb{E}\left(\sum_{i} X_{i}\right) = \sum_{i} \mathbb{E}(X_{i}) = \frac{n}{p} \qquad \text{Var}(X) = \sum_{i} \text{Var}(X_{i}) = \frac{n(1-p)}{p^{2}}$$

#### **Chebyshev's Inequality – Coin Flips**

"How many times does Alice need to flip a biased coin <u>until she sees heads n</u> times, if heads occurs with probability p?

$$\mathbb{E}(X) = \mathbb{E}\left(\sum_{i} X_{i}\right) = \sum_{i} \mathbb{E}(X_{i}) = \frac{n}{p} \quad \operatorname{Var}(X) = \sum_{i} \operatorname{Var}(X_{i}) = \frac{n(1-p)}{p^{2}}$$

What is the probability that  $X \ge 2\mathbb{E}(X) = 2n/p$ ?

Markov: 
$$\mathbb{P}(X \ge 2n/p) \le \frac{\mathbb{E}(X)}{2n/p} = \frac{n}{p} \cdot \frac{p}{2n} = \frac{1}{2}$$
  
Chebyshev:  $\mathbb{P}(X \ge 2n/p) \le \mathbb{P}\left(\left|X - \frac{n}{p}\right| \ge \frac{n}{p}\right) \le \frac{\operatorname{Var}(X)}{n^2/p^2} = \frac{1-p}{n}$   
Goes to zero as  $n \to \infty$   $\odot$ 

#### **Tail Bounds**

Useful for approximations of complex systems. How good the approximation is depends on the actual distribution and the context you are using it in.

 Usually loose upper-bounds are okay when designing for worstcase

Generally (but not always) making more assumptions about your random variable leads to a more accurate upper-bounder.