

# CS5314

# Randomized Algorithms

## Lecture 10: Chernoff Bounds (Introduction)

# Objectives

- Chernoff bounds
  - Based on all the moments for bounding tail distribution (such as  $\Pr(X \geq a)$ )
  - more powerful than Markov (based on  $E[X]$ ) and Chebyshev (based on  $E[X]$  and  $E[X^2]$ )
- Moment generating function

# Moment Generating Function

Definition: The **moment generating function (MGF)** of a random variable  $X$  is defined as

$$M_X(t) = E[e^{tX}]$$

Recall:  $E[X^k]$  is called the  $k$ th moment of  $X$

The MGF is a special function that "captures" all the moments of  $X$

# Moment Generating Function

Theorem:

Let  $X$  be a random variable with MGF  $M_X(t)$

and assume that it is legal to exchange expectation and differentiation operands in calculation \*\*.

Then, for all  $n > 1$ ,

$$E[X^n] = M_X^{(n)}(0),$$

where  $M_X^{(n)}(0) = n$ th derivative of  $M_X(t)$  evaluated at  $t = 0$

\*\* The above assumption is not generally true.

However, it is true for the random variables we study

# Proof

$$\begin{aligned} M_X^{(1)}(t) &= \frac{d}{dt} (E[e^{tX}]) = E\left[\frac{d}{dt} (e^{tX})\right] \\ &= E[Xe^{tX}] \end{aligned}$$

$$\begin{aligned} M_X^{(2)}(t) &= \frac{d}{dt} (E[Xe^{tX}]) = E\left[\frac{d}{dt} (Xe^{tX})\right] \\ &= E[X^2e^{tX}] \end{aligned}$$

...

$$M_X^{(n)}(t) = E[X^n e^{tX}] \Rightarrow M_X^{(n)}(0) = E[X^n]$$

# Example

Let  $X = \text{Geo}(p)$

Questions:

1. What is  $E[e^{tX}]$ ?
2. What is  $M_X^{(1)}(t)$ ? What is  $E[X]$ ?
3. What is  $M_X^{(2)}(t)$ ? What is  $E[X^2]$ ?
4. What is  $\text{Var}[X]$ ?

# Equality of MGF

Theorem:

Let  $X$  and  $Y$  be two random variables.

If there exists  $d > 0$  such that

$$M_X(t) = M_Y(t)$$

for all  $t$  in  $[-d, d]$ , then  $X$  and  $Y$  have the same distribution

Proof: omitted (out of scope) ...

# MGF for Sum of Independent RV

Theorem: If  $X$  and  $Y$  are two independent random variables, then

$$M_{X+Y}(t) = M_X(t) M_Y(t)$$

Proof:

$$\begin{aligned} M_{X+Y}(t) &= E[e^{t(X+Y)}] \\ &= E[e^{tX} e^{tY}] \\ &= E[e^{tX}] E[e^{tY}] \quad \dots \text{(why?)} \\ &= M_X(t) M_Y(t) \end{aligned}$$



# Remark

Suppose that we know  $X$  and  $Y$  are two independent random variables, and we have computed  $M_X(t)$  and  $M_Y(t)$

One day, we find that  $M_X(t) M_Y(t)$  is an MGF for some random variable  $Z$

What does that imply?

# Chernoff Bounds

Theorem: Let  $X$  be a random variable.  
Then,

$$(i) \quad \Pr(X \geq a) \leq \min_{t > 0} E[e^{tX}] / e^{ta}$$

$$(ii) \quad \Pr(X \leq a) \leq \min_{t < 0} E[e^{tX}] / e^{ta}$$

How to prove?

# Proof

(i) For  $t > 0$ ,

$$\begin{aligned}\Pr(X \geq a) &= \Pr(e^{tX} \geq e^{ta}) \quad \dots \text{(why?)} \\ &\leq E[e^{tX}] / e^{ta} \quad \dots \text{(why?)}\end{aligned}$$

Thus,  $\Pr(X \geq a) \leq \min_{t > 0} E[e^{tX}] / e^{ta} \quad \dots \text{(why?)}$

(ii) Similar proof