

① Markov's inequality

② Chebyshev's inequality

Reminder:

HW sessions 3 times a week (@39)

## Markov's Inequality Intro

Simple bound on the tail of a random variable that only uses the expected value (first moment) and the fact that the random variable is nonnegative.

## Markov's Inequality Definition

If  $X$  is a nonnegative r.v. with finite mean and  $a > 0$ , then the probability that  $X$  is at least  $a$  is at most the expectation of  $X$  divided by  $a$ .

## Markov's Inequality: Proof 1

WLOG, let  $X$  be a nonnegative continuous R.V.

$$E[X] =$$

## Markov's Inequality: Proof II

Let  $I$  be the indicator r.v. defined as:

$$I = \begin{cases} 1 & \text{if } X \geq a \\ 0 & \text{o.w.} \end{cases}$$

Then

Markov's Inequality: Proof II

$$E[X] =$$

## Example: Markov & Coin Flips

Let  $X \sim \text{Geom}(\frac{1}{2})$ . Use Markov's inequality to bound  $P(X > 10)$ .

# Generalized Markov's Inequality

If  $X$  is any random variable with finite mean and  $a > 0$ , then for an  $r > 0$ :

Proof left as an exercise.



# Chebyshev's Inequality Intro

Often times we can do better than Markov's Inequality if we use more information about the random variable.

For Chebyshev's Inequality, we use the first two moments  $E[X]$  and  $E[X^2]$ .

Note:

## Chebyshev's Inequality: Definition.

If  $X$  is a random variable with finite mean  $\mu$  and finite variance, and  $c > 0$ , then the probability that  $X$  is at least  $c$  away from  $\mu$  is at most  $\frac{\text{Var}(X)}{c^2}$ .

→ doesn't need to be nonnegative.

Note:

## Chebyshev's Inequality: Proof.

$X$  is a r.v.

$$E[X] = \mu$$

$$\text{Let } Y = (X - \mu)^2,$$

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Example: Chebyshev & Coin Flips

Let  $X \sim \text{Geom}(\frac{1}{2})$ . Use Chebyshev's Inequality to upper bound  $\mathbb{P}(X > 10)$ .

## Chebyshev Corollary

For any random variable  $X$  with finite expectation  $E[X] = \mu$  and finite standard deviation  $\sigma = \sqrt{\text{Var}(X)}$  :

## Example: Chebyshev Corollary

Let  $X \sim N(\mu, \sigma^2)$ . Find a bound on the probability that  $X$  is  $2\sigma$  or more away from its mean  $\mu$ .

## Law of Large Numbers: Preview

If we observe a random variable many times and average our observations, the average will converge to the average of the random variable.

Note: