

# COMPSCI 240: Reasoning Under Uncertainty

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## Lecture 18: Limit Theorems

# Overview

- Let  $X_1, X_2, \dots, X_n$  be a sequence of i.i.d. (either discrete or continuous) random variables with mean of  $\mu$  and variance of  $\sigma^2$ .
- Limit theorems are mostly concerned with the sum of these random variables (which forms another random variable):

$$S_n = X_1 + X_2 + \dots + X_n$$

especially when  $n$  is very large.

- Then, the mean and variance of  $S_n$  can be computed as

$$E[S_n] = E[X_1] + E[X_2] + \dots + E[X_n] = n\mu$$

$$\text{var}(S_n) = \text{var}(X_1) + \text{var}(X_2) + \dots + \text{var}(X_n) = n\sigma^2$$

# Overview

- Let us introduce a new RV:

$$Z_n = \frac{S_n - n\mu}{\sigma\sqrt{n}}$$

- The mean and variance of  $Z_n$  is

$$E[Z_n] = 0$$

$$\text{var}(Z_n) = 1$$

- The **central limit theorem** states that the distribution of  $Z_n$  becomes the **standard normal variable** as  $n$  increases.

## Markov and Chebyshev Bounds

We will learn these two bounds to prove the Central Limit Theorem. More specifically, Markov Inequality  $\rightarrow$  Chebyshev Inequality  $\rightarrow$  Central Limit Theorem.

- **Markov Bound:** For a non-negative random variable  $X$ ,

$$P(X \geq a) \leq \frac{E[X]}{a}$$

- **Chebyshev Bound:** For a random variable  $X$ ,

$$P(|X - E[X]| \geq c) \leq \frac{\text{var}(X)}{c^2}$$

# The Markov Bound

- **Markov Bound:** For any non-negative random variable,

$$P(X \geq a) \leq \frac{E[X]}{a}$$

- **Proof:** Introduce a new RV  $Y_a$  where

$$Y_a = \begin{cases} 0 & \text{if } X < a \\ a & \text{if } X \geq a \end{cases}$$

- Then, we know that  $Y_a \leq X$ , which yields

$$E[Y_a] \leq E[X]$$

- On the other hand,

$$E[Y_a] = a \cdot P(Y_a = a) = a \cdot P(X \geq a)$$

- Thus,

$$P(X \geq a) \leq \frac{E[X]}{a}$$

# The Markov Bound

- Let  $X$  be a continuous random variable with uniform density over  $[0,4]$ .
- Its mean can be computed as

$$E[X] = \int_0^4 x \frac{1}{4} dx = 2$$

- Then, the Markov inequality asserts that

$$P(X \geq 0) \leq 1 \text{ whereas } P(X \geq 0) = 1.$$

$$P(X \geq 1) \leq 1 \text{ whereas } P(X \geq 1) = \frac{3}{4}.$$

$$P(X \geq 2) \leq 1 \text{ whereas } P(X \geq 2) = 0.5.$$

These are uninformative...

$$P(X \geq 3) \leq \frac{2}{3} \text{ whereas } P(X \geq 3) = \frac{1}{4}$$

$$P(X \geq 4) \leq \frac{2}{4} \text{ whereas } P(X \geq 4) = 0$$

- The Markov inequality can be quite loose.

# The Chebyshev Bound

- **Chebyshev Bound:**

$$P(|X - E[X]| \geq c) \leq \frac{\sigma^2}{c^2}, \text{ for all } c > 0$$

- **Proof:** Let us introduce a non-negative RV  $(X - \mu)^2$ .
- We can apply the Markov bound on this RV:

$$P((X - \mu)^2 \geq a) \leq \frac{E[(X - \mu)^2]}{a} = \frac{\sigma^2}{a}$$

- We will only consider  $a > 0$  where  $a$  can be defined as  $a = c^2$ . Then, we have

$$P((X - \mu)^2 \geq c^2) \leq \frac{\sigma^2}{c^2}$$

- Since the event  $(X - \mu)^2 \geq c^2$  is identical to the event  $|X - \mu| \geq c$ ,

$$P((X - \mu)^2 \geq c^2) = P(|X - \mu| \geq c) \leq \frac{\sigma^2}{c^2}$$

- The Chebyshev bound is often more powerful than the Markov bound because it also uses information on the variance of  $X$ .

## An Alternate Form of The Chebyshev Bound

- Let  $c = k\sigma$ , then

$$P(|X - \mu| \geq k\sigma) = P\left(\left|\frac{X - \mu}{\sigma}\right| \geq k\right) = \frac{1}{k^2}$$

- The probability that a RV takes a value more than  $k$  standard deviations from its mean is at most  $1/k^2$ .

# The Chebyshev Bound

- Let  $X$  be a continuous random variable with uniform density over  $[0,4]$ .
- Its mean is

$$E[X] = \int_0^4 x \frac{1}{4} dx = 2.$$

- Its variance is

$$\text{var}(X) = E[X^2] - E[X]^2 = \frac{4}{3}$$

- Then, the Chebyshev inequality asserts that

$$P(|X - 2| \geq 0) \leq 1 \text{ so is not informative}$$

$$P(|X - 2| \geq 1) \leq 1 \text{ so is not informative}$$

$$P(|X - 2| \geq 2) \leq \frac{1}{3} \text{ or equivalently } P(X \geq 4 \cup X \leq 0) \leq \frac{1}{3}$$

- We can also derive that

$$P(0 < X < 4) \geq \frac{2}{3}$$

## Example

Suppose we know that the number of items produced in a factory during a week is a random variable with mean of 50 and variance of 25.

- What can be said about the probability that this week's production will exceed 75?

By Markov's inequality

$$P(X \geq 75) \leq \frac{50}{75} = \frac{2}{3}$$

## Example

Suppose we know that the number of items produced in a factory during a week is a random variable with mean of 50 and variance of 25.

- What can be said about the probability that this week's production will be between 40 and 60?

By Chebyshev's inequality

$$P(|X - 50| \geq 10) \leq \frac{25}{100} = \frac{1}{4}$$

Thus,

$$P(40 \leq X \leq 60) = 1 - P(|X - 50| \geq 10) \geq \frac{3}{4}$$