

# CS70 - Lecture 21 Notes

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## Distributions Review

1. **Uniform** ( $U[1, \dots, n]$ );  $m = 1, \dots, n$

- $\Pr[X = m] = \frac{1}{n}$
- $E[X] = \frac{n+1}{2}$

2. **Binomial or Bernoulli** ( $B(n, p)$ );  $m = 0, \dots, n$

- $\Pr[X = m] = \binom{n}{m} p^m (1-p)^{n-m}$
- $E[X] = np$

3. **Geometric** ( $G(p)$ );  $n = 1, 2, \dots$

- $\Pr[X = n] = (1-p)^{n-1} p$
- $E[X] = \frac{1}{p}$

4. **Poisson** ( $P(\lambda)$ );  $n \geq 0$

- $\Pr[X = n] = \frac{\lambda^n}{n!} e^{-\lambda}$
- $E[X] = \lambda$

5. **Indicator**  $X = 1$  or  $0$

- $\Pr[X = 1] = p, \Pr[X = 0] = (1-p)$
- $E[X] = p$

## Poisson and Queuing

(Derivation in previous notes)

1. Flip coin  $n$  times,  $\Pr[H] = \frac{\lambda}{n}$

2. RV  $X$  = no. of heads (Bernoulli indicator -when 1)

3.  $X = B(n, \frac{\lambda}{n})$

4. Distribution of  $X$  “for large  $n$ ”

- Distribution of the number of events in an interval
- The average value comes out to  $\lambda$
- Cut up situation into  $n \rightarrow \infty$  intervals described by Bernoulli indicators
- This means you can assume no two events occur in the same interval and there is a  $\frac{\lambda}{n}$  chance the indicator is 1 in any interval

## Independence Review

1.  $X, Y$  are independent if and only if:
  - $\Pr[X = x, Y = y] = \Pr[X = x]\Pr[Y = y], \forall x, y$
  - $\Pr[X \in A, Y \in B] = \Pr[X \in A]\Pr[Y \in B], \forall A, B$
2.  $X, Y, Z, \dots$  are mutually independent if and only if:
  - $\Pr[X = x, Y = y, Z = z, \dots] = \Pr[X = x]\Pr[Y = y]\Pr[Z = z] \cdots, \forall x, y, z, \dots$
  - $\Pr[X \in A, Y \in B, Z \in C, \dots] = \Pr[X \in A]\Pr[Y \in B]\Pr[Z \in C] \cdots, \forall A, B, C, \dots$
3. If  $U, V, W, X, Y, Z, \dots$  are all mutually independent then:
  - $f(U, V), g(W, X, Y), h(Z, \dots), \dots$  are mutually independent

## Variance

1. Measures deviation from the mean value (Standard deviation ( $\sigma(X)$ ) squared)
  2. Use squared distance as a continuous function that you can do derivatives and other operations on.
  3. To calculate intermediate value  $E[X^2]$  of expectations with infinite series:
    - Calculate  $E[X^2] - (1 - p)E[X^2] = pE[X^2]$
    - This gives you  $pE[X^2]$  in terms of  $E[X]$  and a known distribution (total dist.=1)
    - Divide both sides by  $p$
  4. Variance of the constant is  $c$  that constant squared  $\text{Var}[c] = c^2$
- $\text{Var}[X] = E[(X - E[X])^2]$
  - $= E[X^2 - 2XE[X] + E[X]^2]$
  - $= E[X^2] - 2E[X]E[X] + E[X]^2$
  - $= E[X^2] - E[X]^2$

### Variance of $X$

$$\sigma^2(X) := \text{Var}[X] = E[(X - E[X])^2] = E[X^2] - E[X]^2 \quad (1)$$

### Potential Final Question:

Give an example where this ratio  $\rightarrow \infty$

$$\frac{\sigma(X)}{E[|X - E[X]|]} \quad (2)$$

### Uniform Variance

Assume that  $\Pr[X = i] = \frac{1}{n}$  for  $i \in \{1, \dots, n\}$

$$E[X] = \frac{1}{n} \sum_{i=1}^n i = \frac{1}{n} \frac{n(n+1)}{2} = \frac{n+1}{2} \quad (3)$$

$$E[X^2] = \sum_{i=1}^n i^2 \Pr[X = i] = \frac{1}{n} \sum_{i=1}^n i^2 = \frac{1 + 3n + 2n^2}{6} \quad (4)$$

$$\text{Var}[X] = \frac{1 + 3n + 2n^2}{6} - \frac{(n+1)^2}{4} = \frac{n^2 - 1}{12} \quad (5)$$

**Geometric Distribution Variance**

$X = G(p)$ ,  $\Pr[X = n] = (1 - p)^{n-1}p$  for  $n \geq 1$

$$E[X] = \frac{1}{p} \quad (6)$$

$$\begin{aligned} E[X^2] &= p + 4p(1 - p) + 9p(1 - p)^2 + \dots \\ (1 - p)E[X^2] &= p(1 - p) + 4p(1 - p)^2 + 9p(1 - p)^3 + \dots \\ pE[X^2] &= 2(p + 2p(1 - p) + 3p(1 - p)^2 + \dots) \\ &\quad - (p + p(1 - p) + p(1 - p)^2 + \dots) \\ &= 2E[X] - \text{Distribution} \\ pE[X^2] &= (2E[X] - 1) = (2(\frac{1}{p}) - 1) = \frac{2 - p}{p} \end{aligned}$$

$$E[X^2] = \frac{2 - p}{p^2} \quad (7)$$

$$\text{Var}[X] = E[X^2] - E[X]^2 = \frac{2 - p}{p^2} - \frac{1}{p^2} = \frac{1 - p}{p^2} \quad (8)$$

$$\sigma(X) = \frac{\sqrt{1 - p}}{p} \quad (9)$$

**Fixed Points Variance**

Number of fixed points in a random permutation of  $n$  items

i.e. Number of students that get hw back with RV  $X = X_1 + X_2 + \dots + X_n$

$X_i$  = indicator for  $i$ th student getting hw back

$$E[X] = 1 \quad (10)$$

$$\begin{aligned} E[X^2] &= \sum_i E[X_i^2] + \sum_{i \neq j} E[X_i X_j] \\ &= n \times (1 \times \Pr[X_i = 1] + 0 \times \Pr[X_i = 0]) = \frac{1}{n} \\ &\quad + n(n - 1) \times (1 \times \Pr[X_i = 1 \cap X_j = 1] + 0 \times \Pr[\text{"anything else"}]) = \frac{1 \times 1 \times (n - 2)!}{n!} = \frac{1}{n(n - 1)} \end{aligned}$$

$$E[X^2] = 2 \quad (11)$$

$$\text{Var}[X] = E[X^2] - E[X]^2 = 2 - 1 = 1 \quad (12)$$

**Binomial**

Flip coin with  $\Pr[H] = p$

$X = X_1 + X_2 + \dots + X_n$  no. of heads ( $X_i, X_j$  are independent)

$X_i = 1$  if  $i$ th flip is heads,  $X_i = 0$  otherwise (indicator RV)

$$E[X] = np \quad (13)$$

$$E[X_i^2] = 1^2 \times p + 0^2 \times (1 - p) = p$$

$$\text{Var}[X_i] = p - (E[X_i])^2 = p - p^2 = p(1 - p)$$

$$\text{Var}[X] = \text{Var}[X_1 + X_2 + \dots + X_n] = n\text{Var}[X_i] = np(1 - p) \quad (14)$$

## Properties of Variance

1.  $\text{Var}[cX] = c^2 \text{Var}[X]$  where  $c$  is a constant
2.  $\text{Var}[X + c] = \text{Var}[X]$  where  $c$  is a constant (shifts center)

### Variance of the Sum of 2 Independent RVs

- **Thm:** If  $X$  and  $Y$  are independent, then  $\text{Var}[X + Y] = \text{Var}[X] + \text{Var}[Y]$ 
  - Assume  $E[X] = E[Y] = 0$  (You can just shift this to accountn for any other RVs)
  - $E[XY] = E[X]E[Y] = 0$  (By Independence)

$$\begin{aligned}\text{Var}[X + Y] &= E[(X + Y)^2] \\ &= E[X^2 + 2XY + Y^2] \\ &= E[X^2] + 2E[XY] + E[Y^2] \\ &= E[X^2] + EY^2 \\ &= \text{Var}[X] + \text{Var}[Y]\end{aligned}$$

### Variance of the Sum of Multiple Independent RVs

- **Thm:** If  $X, Y, Z, \dots$  are independent, then  $\text{Var}[X + Y + Z + \dots] = \text{Var}[X] + \text{Var}[Y] + \text{Var}[Z] + \dots$ 
  - Assume  $E[X] = E[Y] = E[Z] = \dots = 0$  (You can just shift this to accountn for any other RVs)
  - $E[XY] = E[XZ] = E[YZ] = \dots = 0$  (By Independence)

$$\begin{aligned}\text{Var}[X + Y + Z + \dots] &= E[(X + Y + Z + \dots)^2] \\ &= E[X^2 + Y^2 + Z^2 + \dots + 2XY + 2XZ + 2YZ + \dots] \\ &= E[X^2] + E[Y^2] + E[Z^2] + 0 + \dots + 0 \\ &= \text{Var}[X] + \text{Var}[Y] + \text{Var}[Z] + \dots\end{aligned}$$

**If  $X$  and  $Y$  are Independent**

$$\text{Var}[X + Y] = \text{Var}[X] + \text{Var}[Y] \quad (15)$$

**If  $X, Y, Z, \dots$  are Independent**

$$\text{Var}[X + Y + Z + \dots] = \text{Var}[X] + \text{Var}[Y] + \text{Var}[Z] + \dots \quad (16)$$

## Inequalities: Overview

1. Markov:  $\Pr[X \geq a] \leq \frac{E[f(X)]}{f(a)}$ , for all  $a$  such that  $f(a) > 0$  and  $f : \mathbb{R} \rightarrow [0, \infty)$  is non-decreasing
  - Bound the probability of being at least  $a$  away from the mean  $E[X]$
2. Chabyshev:  $\Pr[|X - E[X]| > a] \leq \frac{\text{Var}[X]}{a^2}$ 
  - Bound probability of getting at least  $a$  away from the mean  $E[X]$

### Markov's Inequality Proof

- Assume  $f : \mathbb{R} \rightarrow [0, \infty)$  is non-decreasing.
- $\Pr[X \geq a] \leq \frac{E[f(X)]}{f(a)}$ , for all  $a$  such that  $f(a) > 0$

- Observe that:  $1\{X \geq a\} \leq \frac{f(X)}{f(a)}$  because:
  - When  $X > a$ :
    - \* Left side = 1
    - \* Right side  $> 1$  because  $f : \mathbb{R} \rightarrow [0, \infty)$  is non-decreasing
  - When  $X = a$ :
    - \* Left side = 1
    - \* Right side = 1
  - When  $X < a$ :
    - \* Left side = 0
    - \* Right side  $\geq 0$  because  $f \geq 0$ .
- Take expectation of both sides (because expectation is monotone):
  - $E[1\{X \geq a\}] \leq E[\frac{f(X)}{f(a)}] \implies \Pr[X \geq a] \leq \frac{E[f(X)]}{f(a)}$

#### Markov's Inequality

$f : \mathbb{R} \rightarrow [0, \infty)$  is non-decreasing and for all  $a$  such that  $f(a) > 0$

$$\Pr[X \geq a] \leq \frac{E[f(X)]}{f(a)} \quad (17)$$

#### Markov's Inequality Example $G(p)$ :

- Let  $X = G(p)$ ,  $E[X] = \frac{1}{p}$ ,  $E[X^2] = \frac{2-p}{p^2}$
- Using  $f(x) = x$ :  $\Pr[X \geq a] \leq \frac{E[X]}{a} = \frac{1}{ap}$
- Using  $f(x) = x^2$ :  $\Pr[X \geq a] \leq \frac{E[X^2]}{a^2} = \frac{2-p}{a^2 p^2}$

#### Markov's Inequality Example $P(\lambda)$ :

- Let  $X = P(\lambda)$ ,  $E[X] = \lambda$ ,  $E[X^2] = \lambda + \lambda^2$
- Using  $f(x) = x$ :  $\Pr[X \geq a] \leq \frac{E[X]}{a} = \frac{\lambda}{a}$
- Using  $f(x) = x^2$ :  $\Pr[X \geq a] \leq \frac{E[X^2]}{a^2} = \frac{\lambda + \lambda^2}{a^2}$

#### Chebyshev's Inequality Proof

- Let  $Y = |X - E[X]|$ ,  $f(y) = y^2$
- Use Markov's Inequality:  $\Pr[Y \geq a] \leq \frac{E[f(Y)]}{f(a)} = \frac{\text{Var}[X]}{a^2}$
- Confirms that the variance measures the “deviations from the mean”

#### Chebyshev's Inequality

For all  $a > 0$

$$\Pr[|X - E[X]| > a] \leq \frac{\text{Var}[X]}{a^2} \quad (18)$$

#### Chebyshev's Inequality Example $P(\lambda)$ :

- Let  $X = P(\lambda)$ ,  $E[X] = \lambda$ ,  $\text{Var}[X] = \lambda$
- $\Pr[|X - \lambda| > n] \leq \frac{\lambda}{n^2}$

#### Use Markov to get Chebyshev Bounds

- Let  $X = P(\lambda)$ ,  $E[X] = \lambda$ ,  $E[X^2] = \lambda + \lambda^2$ ,  $\text{Var}[X] = \lambda$

- Using Markov's with  $f(x) = x^2$ :  $\Pr[X \geq a] \leq \frac{E[X^2]}{a^2} = \frac{\lambda + \lambda^2}{a^2}$
- If  $a > \lambda$ , then  $X \geq a \implies X - \lambda \geq a - \lambda > 0 \implies |X - \lambda| \geq a - \lambda$
- So, for  $a > \lambda$ ,  $\Pr[X \geq a] \leq \Pr[|X - \lambda| \geq a - \lambda] \leq \frac{\lambda}{(a - \lambda)^2}$

### Fraction of H's

- How likely is it that the fraction of  $H$ 's differs from 50%?
- Let  $X_m = 1$  if the  $m$ th flip of a fair coin is  $H$  and  $X_m = 0$  otherwise
  - Any type of polling system mimics this scenario
  - $\Pr[H] = p$  Yes to poll question
  - $\Pr[T] = 1 - p$  No to poll question
- Define  $Y_n = \frac{X_1 + \dots + X_n}{n}$ , for  $n \geq 1$ 
  - Ratio of people who said yes to question
- Estimate  $\Pr[|Y_n - 0.5| \geq 0.1] = \Pr[Y_n \leq 0.4 \text{ or } Y_n \geq 0.6]$ 
  - If poll result is  $\geq 10\%$  away from the mean  $\implies$  mistake
- By Chebyshev:  $\Pr[|Y_n - 0.5| \geq 0.1] \leq \frac{\text{Var}[Y_n]}{0.1^2} = 100\text{Var}[Y_n] = \frac{25}{n}$ 
  - $\text{Var}[Y_n] = \frac{1}{n^2}(\text{Var}[X_1] + \text{Var}[X_2] + \dots + \text{Var}[X_n]) = \frac{n\text{Var}[X_i]}{n^2} = \frac{\text{Var}[X_i]}{n} \leq \frac{1}{4n}$
  - Because all  $X_i$ 's are independent and the variance of a constant is its square.
  - As  $n \rightarrow \infty$ ,  $Y_n \rightarrow 0$
  - $\text{Var}[X_i] = p(1 - p) \leq (.5)(.5) = \frac{1}{4}$
- For  $n = 1000$ ,  $\Pr[|Y_n - 0.5| \geq 0.1] \leq 2.5\%$

### Law of Large Numbers

As  $n \rightarrow \infty$ ,  $\Pr[|X - 0.5| \geq \varepsilon] \rightarrow 0$

$$\Pr[|T_n - 0.5| \leq \varepsilon] \rightarrow 1 \quad (19)$$

### Weak Law of Large Numbers

$$\Pr\left[\left|\frac{X_1 + \dots + X_n}{n} - \mu\right| \geq \varepsilon\right] \rightarrow 0, \text{ as } n \rightarrow \infty \quad (20)$$

### Weak Law of Large Numbers

#### Proof

- Let  $Y_n = \frac{X_1 + \dots + X_n}{n}$
- $\Pr[|Y_n - \mu| \geq \varepsilon] \leq \frac{\text{Var}[Y_n]}{\mu^2} = \frac{\text{Var}[X_1 + \dots + X_n]}{n^2 \varepsilon^2} = \frac{\text{Var}[X_i]}{n \varepsilon^2} \rightarrow 0, \text{ as } n \rightarrow \infty$

### Summary

**Variance**

$$\text{Var}[X] = E[(X - E[X])^2] = E[X^2] - E[X]^2 \quad (21)$$

**Fact**

$$\text{Var}[aX + b] = a^2 \text{Var}[X] \quad (22)$$

**Sum**

$$X, Y, Z, \dots \text{ mutually independent} \implies \text{Var}[X + Y + Z + \dots] = \text{Var}[X] + \text{Var}[Y] + \text{Var}[Z] + \dots \quad (23)$$

**Markov**

$f : \mathbb{R} \rightarrow [0, \infty)$  is non-decreasing and for all  $a$  such that  $f(a) > 0$

$$\Pr[X \geq a] \leq \frac{E[f(X)]}{f(a)} \quad (24)$$

**Chebyshev**

For all  $a > 0$

$$\Pr[|X - E[X]| \geq a] \leq \frac{\text{Var}[X]}{a^2} \quad (25)$$

**Weak Law of Large Numbers**

$$X_m \text{ i.i.d} \implies \frac{X_1 + \dots + X_n}{n} \approx E[X] \quad (26)$$