

# Probability and Statistics: Lecture-17

Monsoon-2020

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## » Higher Order Moments...

### Define $n$ th moment

The  $n$ th moment about the mean or  $n$ th central moment of a real valued random variable  $X$  is defined as follows

$$\mu_n = E[(X - E[X])^n],$$

where  $E$  is the expectation operator.

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$$\mu_1 = E[X - E[X]] = 0$$

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*↑ about mean*

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### Generating Moments...

Is there a quick way to generate moments?



## » Moment Generating Function...

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$$g(x) = e^{tx}$$

### Moment Generating Function

The moment generating function  $M_X(t)$  is the expectation value

*put t* →  $M_X(t) = E[\underbrace{e^{tx}}] = \sum_x \underbrace{e^{tx}} \underbrace{p_X(x)}$

### Lemma

✓  $M_X(0) = 1$

✓  $E[X] = M'_X(0)$ , where ' is the derivative w.r.t.  $t$

*derivative  
w.r.t. t*

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$$\begin{aligned} M_X(t) &= \sum_{x=0}^n \binom{n}{x} p^x (1-p)^{n-x} e^{tx} \\ &= \sum_{x=0}^n \binom{n}{x} \underbrace{(e^t p)^x (1-p)^{n-x}}_{\text{PMF of } X} \\ &= (e^t p + 1 - p). \quad \leftarrow \end{aligned}$$

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$$\rightarrow M'_X(t) = n(e^t p + 1 - p)^{n-1} p e^t.$$



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Differentiating w.r.t.  $t$ , we have

$$M'_X(t) = n(e^t p + 1 - p)^{n-1} p e^t.$$

- \* Setting  $t = 0$ ,  $\underline{M'_X(0)} = \underline{np} = \underline{E[X]}$

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$$\begin{aligned} M_X(t) &= \sum_{x=0}^{\infty} e^{-\lambda} \frac{\lambda^x}{x!} e^{tx} \\ &= \sum_{x=0}^{\infty} e^{-\lambda} \frac{(\lambda e^t)^x}{x!} \quad \text{PMF of Poisson } (\lambda) \\ &= e^{\lambda(e^t - 1)} \quad \leftarrow \end{aligned}$$

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## » Variance Using Moment Generating Function...

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$$E[X^2]$$

$$E[X]^2$$

Variance Using Moment Generating Function

$$\text{Var}(X) = M''_X(0) - M'_X(0)^2$$

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

## » Computing Variance Using Moment Generating Function...

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### Computing Variance using moment generating function

Let  $X$  be a discrete random variable whose PMF is a binomial distribution with parameters  $n$  and  $p$ . It has mean  $\mu = np$  and the moment generating function is

$$\underline{M_X(t)} = (e^t p + 1 - p)^n$$

$$M_X'(t) = \underbrace{n(e^t p + 1 - p)^{n-1}}_{2^{nd}} \underbrace{p e^t}_{1^{st}}$$

$$M_X''(t) = \underbrace{p^2 e^{2t}}_{2^{nd}} \cdot \underbrace{n(n-1)(e^t p + 1 - p)^{n-2}}_{1^{st}} + \underbrace{n(e^t p + 1 - p)^{n-1}}_{1^{st}} \underbrace{p e^t}_{1^{st}}$$

$$M_X''(0) = n(n-1)p^2 + np$$

$$\begin{aligned} \text{Var}(X) &= M_X''(0) - \underbrace{(M_X'(0))^2}_{(np)^2} = n(n-1)p^2 + np - (np)^2 \\ &= \underline{\underline{np(1-p)}} \end{aligned}$$

## » Computing Variance Using Moment Generating Function...

### Computing Variance using moment generating function

Let  $X$  be a discrete random variable whose PMF is a binomial distribution with parameters  $n$  and  $p$ . It has mean  $\mu = np$  and the moment generating function is

$$M_X(t) = (e^t p + 1 - p)^n$$

Find the Variance using  $M_X(t)$ .

## » Computing Variance Using Moment Generating Function...

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Computing Variance using moment generating function

Let  $X$  be a discrete random variable whose PMF is a Poisson distribution with mean  $\lambda$ .  
The moment generating function is

$$M_X(t) = \underbrace{e^{\lambda(e^t - 1)}}_{f(t)} \quad \begin{matrix} u(t) \\ f'(t) = e^{u(t)} \cdot u'(t) \\ f'(t) = e \end{matrix}$$

$$M_X'(t) = e^{\lambda(e^t - 1)} \cdot e^t \cdot \lambda$$

$$M_X'(0) = \lambda \leftarrow \checkmark$$

$$M_X''(t) = e^{\lambda(e^t - 1)} \lambda e^t + e^{\lambda(e^t - 1)} (\lambda e^t)^2$$

At  $\underline{t=0}$   $M_X''(0) = \lambda + \lambda^2$  |  $\text{Var}(X) = M_X''(0) - \{M_X'(0)\}^2$

$$= \lambda + \lambda^2 - (\lambda)^2 = \lambda$$

## » Computing Variance Using Moment Generating Function...

### Computing Variance using moment generating function

Let  $X$  be a discrete random variable whose PMF is a Poisson distribution with mean  $\lambda$ .  
The moment generating function is

$$M_X(t) = e^{\lambda(e^t - 1)}$$

Find the Variance using  $M_X(t)$ .



## » Problem 1

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Let  $X$  be a discrete R.V. with PMF

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$$P_X(x) = \begin{cases} 0.1 & \text{for } x = 0.2 \\ 0.2 & \text{for } x = 0.4 \\ 0.2 & \text{for } x = 0.5 \\ 0.3 & \text{for } x = 0.8 \\ 0.2 & \text{for } x = 1 \\ 0 & \text{otherwise} \end{cases}$$

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Answer the following:

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2. Find  $P(X \leq 0.5)$

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Answer the following:

1. Find  $R_X$
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3. Find  $P(0.25 < X < 0.75)$

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Handwritten notes:  $x < 0.6$ ,  $x \in \{0.2, 0.4, 0.5\}$ ,  $x = 0.2$ ,  $x \in \{0.2\}$ ,  $x > 0.5$

Answer the following:

1. Find  $R_X$
2. Find  $P(X \leq 0.5)$  ✓
3. Find  $P(0.25 < X < 0.75)$
4. Find  $P(\underline{X} = 0.2 \mid X < 0.6)$



## » Answer to previous problem...

a) Find  $R_x = \{0.2, 0.4, 0.5, 0.8, 1\}$

b)  $P(X \leq 0.5) = P(X \in \{0.2, 0.4, 0.5\})$

$$= P(0.2) + P(0.4) + P(0.5)$$

$$= P(X=0.2) + P(X=0.4) + P(X=0.5)$$

$$= 0.1 + 0.2 + 0.2 = \underline{\underline{0.5}}$$

c)  $P(0.25 < X < 0.75)$

$$= P(X \in \{\underline{0.4}, \underline{0.5}\})$$

$$= P(X=0.4) + P(X=0.5)$$

$$= 0.2 + 0.2$$

$$= \underline{\underline{0.4}}$$

d)  $P(X=0.2 \mid X < 0.6)$

$$= \frac{P(\underline{X=0.2} \text{ and } (X < 0.6))}{P(X < 0.6)}$$

$$= \frac{P(X=0.2)}{P(X < 0.6)} = \frac{0.1}{0.5} = \underline{\underline{0.2}}$$

» Answer to previous problem...



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2. Find  $P(X = 2, Y = 6)$

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2. Find  $P(X = 2, Y = 6)$
3. Find  $P(X > 3 \mid Y = 2)$



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4. Let  $Z = X + Y$ . Find the range and PMF of  $Z$

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2. Find  $P(X = 2, Y = 6)$
3. Find  $P(X > 3 \mid Y = 2)$
4. Let  $Z = X + Y$ . Find the range and PMF of  $Z$
5. Find  $P(X = 4 \mid Z = 8)$

» Answer to previous problem...

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## » Problem 3

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Consider an exam that contains 20 MCQs.

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1. What is PMF of  $X$ ?

## » Problem 3

$$X = \underbrace{(10)}_{\text{correct}} + \underbrace{(10)}_{\text{random}} + \underbrace{Y}_{\text{random}} \quad 10 \text{ } \underline{\geq 5}$$

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1. What is PMF of  $X$ ?
2. What is  $P(X > 15)$ ?

Sol: Let  $Y =$  # correct answer to the remaining 10 MCQs.  
that you answer randomly.

$$\Rightarrow X = Y + 10$$

Goal: Find PMF of  $Y$ .

# » Answer to previous problem...

For each question, succ. prob. is  $\frac{1}{4}$

We perform 10 independent Bernoulli( $\frac{1}{4}$ ) trials and  $Y$  is the no. of Success.

$\Rightarrow Y \sim \text{Binomial}(10, \frac{1}{4})$ .

$$\text{So, } P_Y(y) = \begin{cases} \binom{10}{y} \left(\frac{1}{4}\right)^y \left(\frac{3}{4}\right)^{10-y} & \text{for } y = 0, 1, 2, \dots, 10 \\ 0, & \text{otherwise} \end{cases}$$

Need to find PMF of  $X$ .

$$X = Y + 10$$

$$R_X = \{10, 11, \dots, 20\}$$

$$\begin{aligned} P_X(10) &= P(X=10) = P(Y+10=10) \\ &= P(Y=0) = \binom{10}{0} \left(\frac{1}{4}\right)^0 \left(\frac{3}{4}\right)^{10-0} \\ &= \left(\frac{3}{4}\right)^{10} \end{aligned}$$

$$\begin{aligned} P_X(11) &= P(Y+10=11) = P(Y=1) \\ &= \binom{10}{1} \left(\frac{1}{4}\right)^1 \left(\frac{3}{4}\right)^{10-1} = 10 \cdot \frac{1}{4} \left(\frac{3}{4}\right)^9 \end{aligned}$$

» Answer to previous problem...

$$P_X(k) = P(X=k) = P(Y+10=k) = P(Y=k-10) \quad P(X > 15) \quad 10 + \{6, 7, 8, 9, 10\} > 15$$

$$= \binom{10}{k-10} \left(\frac{1}{4}\right)^{k-10} \left(\frac{3}{4}\right)^{20-k}$$

$$\Rightarrow Y = 6, 7, 8, 9, 10$$

To summarize

$$P_X(k) = \begin{cases} \binom{10}{k-10} \left(\frac{1}{4}\right)^{k-10} \left(\frac{3}{4}\right)^{20-k} & k=10, 11, \dots, 20 \\ 0 & \text{otherwise} \end{cases}$$

$$P_Y(y) = \begin{cases} \binom{10}{y} \left(\frac{1}{4}\right)^y \left(\frac{3}{4}\right)^{10-y} & y=6, 7, \dots, 10 \\ 0 & \text{otherwise} \end{cases}$$

$$P_X(k) = \begin{cases} \binom{10}{k-10} \left(\frac{1}{4}\right)^{k-10} \left(\frac{3}{4}\right)^{20-k} & k=16, \dots, 20 \\ 0 & \text{otherwise} \end{cases}$$

## » Problem 4

$$\begin{aligned} P(X > 15) &= P_X(16) + P_X(17) + \dots + P_X(20) \\ &= \binom{10}{6} \left(\frac{1}{4}\right)^6 \left(\frac{3}{4}\right)^4 + \dots + \binom{10}{10} \left(\frac{1}{4}\right)^{10} \left(\frac{3}{4}\right)^0 \end{aligned}$$

=

## » Problem 4

### Problem

Average number of customers arriving at a grocery store per hour is 10.



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Average number of customers arriving at a grocery store per hour is 10. Let  $X$  denote the number of customers arriving from 10AM to 11 : 30AM.

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Average number of customers arriving at a grocery store per hour is 10. Let  $X$  denote the number of customers arriving from 10AM to 11:30AM. What is  $P(10 < X \leq 15)$ ?

Sol<sup>n</sup> We have interval of length 1.5 hours  
 $\Rightarrow$  no. of customers in this interval is

$$X \sim \text{Poisson} \left( \lambda = 1.5 \times 10 = 15 \right)$$
$$\Rightarrow P(10 < X \leq 15) = \sum_{k=11}^{15} P_X(k) = \sum_{k=11}^{15} \frac{e^{-15} \cdot 15^k}{k!} = \dots$$