

RecapChapter 7

↓  
 $\left\{ \begin{array}{l} \bullet X \text{ is r.v. with pdf } f(x) \\ \text{then if } g(x) \text{ is a function of } X, \text{ then what is} \\ \text{the pdf of } g(x). \end{array} \right.$

Today's classMoment generating function

What are moments? If  $X$  is a r.v. with pdf  $f(x)$ .

$\int$  or  $\sum x f(x) \leftarrow E(X) \rightarrow$  first moment

$\int$  or  $\sum x^2 f(x) \leftarrow E(X^2) \rightarrow$  second moment

$\int$  or  $\sum x^r f(x) \leftarrow E(X^r) \rightarrow r^{\text{th}}$  moment of r.v.  $X \rightarrow$  Notation  $M'_X$



To get rid of individual calculations, we are looking for a function which can give us all the moments.

↳ MOMENT GENERATING FUNCTION.

Definition Given a r.v.  $X$  with pdf  $f(x)$ , a function

$$\text{function of } t \leftarrow E(e^{tX}) = \int_{-\infty}^{\infty} e^{tx} f(x) dx$$

↓  
 This function is called moment generating function for r.v.  $X$  denoted by  $M_X(t)$

Theorem

If  $M_X(t)$  is moment generating function, then

$$\text{first moment} = \left. \frac{d}{dt} M_X(t) \right|_{t=0}$$

$$\text{Second moment} = \left. \frac{d^2}{dt^2} M_X(t) \right|_{t=0}$$

$$\vdots$$

$$r^{\text{th}} \text{ moment} = \left. \frac{d^r}{dt^r} M_X(t) \right|_{t=0}$$

(without proof)

Que If  $M_X(t) = t^3 + 3t^2 + 2$  is moment generating function of a

Que If  $M_X(t) = t^3 + 3t^2 + 2$  is moment generating function of a r.v.  $X$ , then find its mean & variance.

Sol<sup>n</sup> For mean  $\mu \rightarrow$  first moment  $= \left. \frac{d}{dt} M_X(t) \right|_{t=0}$   
 $= (3t^2 + 6t) \Big|_{t=0}$   
 $= 0$

Variance = Second moment - (1st moment)<sup>2</sup>  
 $= (6t + 6) \Big|_{t=0} - (0)^2 = \boxed{6} \text{ Ans.}$

Que Find the moment generating function of a binomial r.v.  $X$  and use it to verify that  $\mu = np$  &  $\sigma^2 = npq$ .

Sol<sup>n</sup>  $M_X(t) = E(e^{tX}) = \sum_{x=0}^n \underbrace{(e^{tx})}_{\substack{\downarrow \text{pdf} \\ nC_x p^x q^{n-x}}} f(x)$

$$M_X(t) = \sum_{x=0}^n nC_x (pe^t)^x q^{n-x}$$

$$M_X(t) = (pe^t + q)^n \quad \swarrow \text{Binomial expansion}$$

mean  $= \left. \frac{d}{dt} M_X(t) \right|_{t=0} = \left. n(pe^t + q)^{n-1} (pe^t) \right|_{t=0}$   
 $= n(p+q)^{n-1} (pe^0)$   
 $= \boxed{np} \text{ Ans.}$

For Variance

$M_2'$ : Second moment  $= \left. \frac{d^2}{dt^2} M_X(t) \right|_{t=0} = np \left[ \underbrace{(e^t)^2}_{pe^{2t}} (pe^t + q)^{n-1} + (n-1)(pe^t + q)^{n-2} (pe^t)^2 \right]_{t=0}$   
 $= np [1 + (n-1)p]$

$$\sigma^2 = M_2' - (M_1')^2 =$$

$$np[1 + (n-1)p] - n^2 p^2 = np(1-p) = npq$$

$$np + n^2 p^2 - n^2 p^2 = npq$$

$$np(1-p)$$

Que Show that the moment generating function of a r.v.  $X$  having a normal probability distribution with mean  $\mu$  & variance  $\sigma^2$  is given by

$$M_X(t) = \exp\left(\mu t + \frac{1}{2}\sigma^2 t^2\right).$$

Sol<sup>n</sup>

$$\begin{aligned} M_X(t) &= \int_{-\infty}^{\infty} e^{tx} \text{pdf}(x) dx = \int_{-\infty}^{\infty} e^{tx} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx \\ &= \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} e^{tx} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx \\ &= \frac{1}{\sqrt{2\pi}\sigma} \int_{-\infty}^{\infty} e^{\left(-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2 + tx\right)} dx \quad \rightarrow \text{Try to complete the square} \\ &\quad \downarrow \text{Integrate it.} \\ &= \exp\left(\mu t + \frac{1}{2}\sigma^2 t^2\right) \end{aligned}$$

Some important properties

(1)  $M_{X+a}(t) = e^{at} M_X(t).$

Proof

$$\begin{aligned} M_{X+a}(t) &= E\left[e^{t(X+a)}\right] \\ &= E\left[e^{tX} \cdot e^{at}\right] \\ &= e^{at} E\left[e^{tX}\right] = e^{at} M_X(t) \end{aligned}$$

(2)  $M_{aX}(t) = M_X(at)$

Proof

$$E\left[e^{t(aX)}\right] = E\left[e^{(at)X}\right] = M_X(at)$$

(3) If  $X_1, X_2, \dots, X_n$  are  $n$  independent r.v. with moment generating functions  $M_{X_1}, M_{X_2}, \dots, M_{X_n}$  then the

r.v.  $Y = X_1 + X_2 + \dots + X_n$  has moment generating function  
 $= \mu_{X_1}(t) \mu_{X_2}(t) \dots \mu_{X_n}(t)$   
 product.

Proof  $E[e^{tY}] = E[e^{t(X_1 + X_2 + \dots + X_n)}]$

$$= E[e^{tX_1} e^{tX_2} \dots e^{tX_n}] \quad \text{independence}$$

$$= E[e^{tX_1}] E[e^{tX_2}] \dots E[e^{tX_n}]$$

$$= \mu_{X_1} \mu_{X_2} \dots \mu_{X_n}$$

Chapter 7 ends here