

# Moment Generating Functions

(1)

Given a random variable  $X$ , its moment generating function (mgf) is given by:

$$m_x(t) = E_x[e^{tx}]$$

Ex. 1a) Let  $X \sim \text{exp}(\lambda)$

$$\begin{aligned} m_x(t) &= \int_0^{\infty} e^{tx} \lambda e^{-\lambda x} dx = \int_0^{\infty} \lambda e^{(t-\lambda)x} dx \\ &= \frac{\lambda}{\lambda-t} \underbrace{\int_0^{\infty} (\lambda-t) e^{-(\lambda-t)x} dx}_{=1} = \frac{\lambda}{\lambda-t} \quad \text{for } t < \lambda \end{aligned}$$

Properties of the mgf:

i)  $m_x(0) = E_x[e^{0x}] = E_x[1] = 1$

ii)  $m_x^{(k)}(0) = E_x[X^k]$  where  $m_x^{(k)}(t) = \frac{d^k}{dt^k} m_x(t)$

PF.

$$\frac{d^k}{dt^k} E_x[e^{tx}] \stackrel{*}{=} E_x \left[ \frac{d^k}{dt^k} e^{tx} \right] = E_x \left[ \frac{d^{k-1}}{dt^{k-1}} (t e^{tx}) \right] = \dots$$

$$= E_x [X^k e^{tx}]$$

so  $m_x^{(k)}(0) = E_x [X^k e^0] = E_x [X^k]$

Ex. 1b) Let  $X \sim \text{exp}(\lambda)$  so  $m_x(t) = \frac{\lambda}{\lambda-t}$  for  $t < \lambda$ , then

$$m_x'(t) = \frac{(\lambda-t) \cdot 0 - \lambda(-1)}{(\lambda-t)^2} = \frac{\lambda}{(\lambda-t)^2} \Rightarrow m_x'(0) = \frac{\lambda}{\lambda^2} = \frac{1}{\lambda}$$

$$m_x''(t) = \frac{(\lambda-t)^2 \cdot 0 - \lambda[2(\lambda-t)]}{(\lambda-t)^4} = \frac{2\lambda(\lambda-t)}{(\lambda-t)^4} \Rightarrow m_x''(0) = \frac{2\lambda^2}{\lambda^4} = \frac{2}{\lambda^2}$$

Note:

(2)

i) Not all random variables have mgfs.

- An example is the Cauchy Distribution, its pdf is given by

$$f_X(x; \nu, \tau) = \frac{1}{\pi \tau \left(1 + \left(\frac{x-\nu}{\tau}\right)^2\right)} = \frac{1}{\pi \tau} \left(\frac{\tau^2}{(x-\nu)^2 + \tau^2}\right)$$

ii) If two distributions have the same mgf, then they are identically distributed.

- Consequently, the mgf is another way to specify the probability distribution.

- The proof of (ii) relies on the theory of Laplace Transforms.

Result: Suppose  $X_1, X_2, \dots, X_n$  are i.i.d (independent & identically distributed). Define the r.v.  $S$  as,

$$S = \sum_{i=1}^n (a_i X_i + b_i) \quad \text{where } a_i, b_i \in \mathbb{R}$$

then it follows  $m_S(t) = \exp\left(t \sum_{i=1}^n b_i\right) \prod_{i=1}^n m_{X_i}(a_i t)$

PF:

$$m_S(t) = E[e^{tS}] = E\left[\exp\left(t \sum_{i=1}^n (a_i X_i + b_i)\right)\right]$$

$$= \exp\left(t \sum_{i=1}^n b_i\right) E\left[\prod_{i=1}^n e^{t a_i X_i}\right] \underset{\substack{\uparrow \\ \text{by independence}}}{=} \exp\left(t \sum_{i=1}^n b_i\right) \prod_{i=1}^n E_{X_i}[e^{t a_i X_i}]$$

$$= \exp\left(t \sum_{i=1}^n b_i\right) \prod_{i=1}^n m_{X_i}(a_i t) \quad \text{Q.E.D.}$$

Corollary I: Suppose  $Y = aX + b$ , if  $X$  has a mgf then it follows

$$m_Y(t) = e^{bt} m_X(at)$$

Corollary II: Suppose  $X_1, X_2, \dots, X_n$  are i.i.d. Define the r.v.  $S$  as,

$$S = \sum_{i=1}^n X_i$$

Then  $m_S(t) = (m_X(t))^n$ .

Ex. 2) Let  $Y \sim \text{Gamma}(K, \lambda)$ , then

$$\begin{aligned} m_Y(t) &= \frac{\lambda^K}{\Gamma(K)} \int_0^{\infty} e^{ty} y^{K-1} e^{-\lambda y} dy = \frac{\lambda^K}{\Gamma(K)} \int_0^{\infty} y^{K-1} e^{(t-\lambda)y} dy \\ &= \frac{\lambda^K}{(\lambda-t)^K} \underbrace{\frac{(\lambda-t)^K}{\Gamma(K)} \int_0^{\infty} y^{K-1} e^{-(\lambda-t)y} dy}_{=1} = \left(\frac{\lambda}{\lambda-t}\right)^K \text{ for } t < \lambda \end{aligned}$$

But,  $\frac{\lambda}{\lambda-t}$  is the mgf of an  $\text{exp}(\lambda)$ , thus if  $X_1, X_2, \dots, X_n$  are i.i.d  $\text{exp}(\lambda)$ , then  $S = \sum_{i=1}^n X_i \sim \text{Gamma}(n, \lambda)$ .