

## MATH323 - Tutorial 7. (Feb. 22<sup>nd</sup> / 17)

Def: The  $k^{\text{th}}$  moment of a random variable  $Y$  taken about the origin is defined to be  $E(Y^k) = \mu_k'$ .

The moment-generating function  $m(t) := E(e^{tY})$  for random variable  $Y$  exists if there exists a positive constant  $b$  such that  $m(t)$  is finite for  $|t| \leq b$ .

Theorem: If  $m(t)$  exists, then for any positive integer  $k$ ,

$$\left. \frac{d^k}{dt^k} m(t) \right|_{t=0} = m^{(k)}(0) = E(Y^k) = \mu_k'.$$

1. Derive the moment generating function for the Binomial distribution and use it to calculate the mean and variance.

Solution:

Let  $Y \sim \text{Binomial}(n, p)$ , then  $P(Y=y) = \binom{n}{y} p^y (1-p)^{n-y}$  where  $y \in \{0, 1, \dots, n\}$  and  $0 \leq p \leq 1$ .

$$\begin{aligned} m(t) &= E(e^{tY}) = \sum_{y=0}^n \binom{n}{y} p^y (1-p)^{n-y} e^{ty} \\ &= \sum_{y=0}^n \binom{n}{y} (e^t p)^y (1-p)^{n-y} \end{aligned}$$

which by the binomial expansion

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

$$m(t) = (1-p + pe^t)^n$$

$$\frac{d}{dt} m(t) = n(1-p + pe^t)^{n-1} pe^t$$

$$\frac{d^2}{dt^2} m(t) = np \left\{ e^t (1-p + pe^t)^{n-1} + e^t (n-1)(1-p + pe^t)^{n-2} pe^t \right\}$$



$$\begin{aligned}\left. \frac{d}{dt} m(t) \right|_{t=0} &= n(1-p+pe^0)^{n-1} pe^0 \\ &= n(1-p+p)^{n-1} p \\ &= np = \mathbb{E}(Y)\end{aligned}$$

$$\begin{aligned}\left. \frac{d^2}{dt^2} m(t) \right|_{t=0} &= np \left\{ e^0 (1-p+pe^0)^{n-1} + e^0 (n-1)(1-p+pe^0)^{n-2} pe^0 \right\} \\ &= np \{ 1 + (n-1)p \} = np(1+np-p) \\ &= np + n^2 p^2 - np^2 = \mathbb{E}(Y^2)\end{aligned}$$

$$\begin{aligned}V(Y) &= \mathbb{E}(Y^2) - (\mathbb{E}(Y))^2 \\ &= np + n^2 p^2 - np^2 - (np)^2 \\ &= np + n^2 p^2 - np^2 - n^2 p^2 = np - np^2 = np(1-p)\end{aligned}$$

2. (Exercise 3.156, 8): Suppose that  $Y$  is a random variable with moment-generating function  $m(t)$ .

a) What is  $m(0)$ ?

b) If  $W = 3Y$ , show the moment-generating function of  $W$  is  $m(3t)$ .

c) If  $X = Y - 2$ , show the moment-generating function of  $X$  is  $e^{-2t} m(t)$ .

d) If  $Z = aY + b$ , show the moment-generating function of  $Z$  is  $e^{tb} m(at)$ .

Solution:

$$\begin{aligned}\text{a) } m(t) &= \mathbb{E}(e^{tY}) \\ \Rightarrow m(0) &= \mathbb{E}(e^{0Y}) = \mathbb{E}(1) = 1\end{aligned}$$

$$\begin{aligned}\text{b) let } m_W(t) &\text{ be the mgf of } W. \\ m_W(t) &= \mathbb{E}(e^{tW}) = \mathbb{E}(e^{t(3Y)}) = \mathbb{E}(e^{(3t)Y}) \\ &= m(3t)\end{aligned}$$

$$\begin{aligned}\text{c) let } m_X(t) &\text{ be the mgf of } X. \\ m_X(t) &= \mathbb{E}(e^{tX}) = \mathbb{E}(e^{t(Y-2)}) = \mathbb{E}(e^{tY-2t}) \\ &= \mathbb{E}(e^{-2t} e^{tY}) \\ &= e^{-2t} \mathbb{E}(e^{tY}) \\ &= e^{-2t} m(t)\end{aligned}$$



d) let  $m_Z(t)$  be the mgf of  $Z$ .

$$\begin{aligned} m_Z(t) &= E(e^{tZ}) = E(e^{t(aY+b)}) \\ &= E(e^{(at)Y} e^{tb}) \\ &= e^{tb} E(e^{(at)Y}) \\ &= e^{tb} m(at) \end{aligned}$$

### Tchebysheff's Theorem

let  $Y$  be a random variable with mean  $\mu$  and finite variance  $\sigma^2$ .  
Then, for any constant  $k > 0$ ,  
 $P(|Y - \mu| < k\sigma) \geq 1 - \frac{1}{k^2}$  or  $P(|Y - \mu| \geq k\sigma) \leq \frac{1}{k^2}$ .

3. (Exercise 3.170) The U.S. mint produces dimes with an average diameter of 0.5 inch and standard deviation 0.01. Using Tchebysheff's theorem, find a lower bound for the number of coins in a lot of 400 coins that are expected to have a diameter between 0.48 and 0.52.

Solution:

let  $Y$  be the diameter of a coin. Then,

$$\begin{aligned} &P(0.48 < Y < 0.52) \\ &= P(0.48 - 0.50 < Y - 0.50 < 0.52 - 0.50) \\ &= P(-0.02 < Y - 0.50 < 0.02) \\ &= P(|Y - 0.50| < 0.02) \\ &= P(|Y - 0.50| < 2(0.01)) \end{aligned}$$

Then by Tchebysheff's theorem,

$$\begin{aligned} P(|Y - 0.50| < 2(0.01)) &\geq 1 - \frac{1}{2^2} \\ &= 1 - \frac{1}{4} \\ &= \frac{3}{4} \end{aligned}$$

Thus, a lower bound on the probability is  $3/4$  which implies a lower bound for the number of coins out of 400 is  $400 \times 3/4 = 300$ .



## A look ahead

### Discrete Random Variable

• Support: Finite or infinitely countable

• Probability: probability mass function  
 $TP(X=x) = p(x)$

• Total Probability:  $\sum_{\text{supp}(X)} p(x) = 1$

• Expected Value:  $\sum_{\text{supp}(X)} x p(x) = E(X)$

• Second moment:  $\sum_{\text{supp}(X)} x^2 p(x) = E(X^2)$

• Expectation of a function of X:  $\sum_{\text{supp}(X)} g(x) p(x) = E(g(X))$

• Moment Generating function:  $m(t) = \sum_{\text{supp}(X)} e^{tx} p(x)$

• Tchebysheff's Theorem: Holds

### Continuous Random Variable

Uncountable

cumulative distribution function  
 $TP(X \leq x) = F(x)$

density function  
 $f(x) = \frac{dF(x)}{dx} = F'(x)$

$\int_{\text{supp}(X)} f(x) dx = F(\infty) = 1$

$\int_{\text{supp}(X)} x f(x) dx = E(X)$

$\int_{\text{supp}(X)} x^2 f(x) dx = E(X^2)$

$\int_{\text{supp}(X)} g(x) f(x) dx = E(g(X))$

$m(t) = \int_{\text{supp}(X)} e^{tx} f(x) dx$

Holds