

Normal moment generating function

Stat 567

2-7-17

(continued)

①

$$\phi(t) = \int_{-\infty}^{\infty} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2\sigma^2}(x^2 - 2(\mu + \sigma^2 t)x + (\mu + \sigma^2 t)^2 - (\mu + \sigma^2 t)^2 + \mu^2)} dx$$

$$= e^{-\frac{1}{2\sigma^2}(-(\mu + \sigma^2 t)^2 + \mu^2)} \underbrace{\int_{-\infty}^{\infty} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2\sigma^2}(x - (\mu + \sigma^2 t))^2} dx}_1$$

$$= e^{-\frac{1}{2\sigma^2}(-\mu^2 - 2\mu\sigma^2 t - \sigma^4 t^2 + \mu^2)}$$

$$= e^{\mu t + \frac{\sigma^2 t^2}{2}} \quad \text{Check } \phi(0) = 1 \quad \checkmark$$

$$\phi'(t) = e^{\mu t + \frac{1}{2}\sigma^2 t^2} (\mu + \sigma^2 t)$$

②

$$\phi''(t) = e^{\mu t + \frac{1}{2}\sigma^2 t^2} (\sigma^2) + e^{\mu t + \frac{1}{2}\sigma^2 t^2} (\mu + \sigma^2 t)^2$$

$$\phi'(0) = \mu = E[X]$$

$$\phi''(0) = \sigma^2 + \mu^2 = E[X^2]$$

$$\begin{aligned} V[X] &= E[X^2] - (E[X])^2 = \sigma^2 + \mu^2 - \mu^2 \\ &= \sigma^2 \end{aligned}$$

For Bernoulli, Binomial, Geometric, Poisson,
Uniform, Exponential, Gamma, Normal,

You need to know

- 1) pmf or pdf
- 2) values that X takes on
- 3) mgf
- 4) μ & σ^2

Joint distributions

Defn: The joint cdf of 2 random variables X, Y

$$\Rightarrow F(a, b) = P(X \leq a \cap Y \leq b)$$

Note: $\lim_{b \rightarrow \infty} F(a, b) = \lim_{b \rightarrow \infty} P(X \leq a \cap Y \leq b)$

$$\begin{aligned} &= P(X \leq a) \\ &= F_X(a) \end{aligned}$$

Likewise, $F(\infty, b) = F_Y(b)$

Discrete case

Defn: The joint probability mass function is

$$p(x, y) = P(X=x \cap Y=y)$$

(5)

$$\begin{aligned}
 p_X(x) &= P(X=x) \\
 &= \sum_y P(X=x \cap Y=y) \quad \left[\begin{array}{l} \text{Law of} \\ \text{Total} \\ \text{Probability} \end{array} \right] \\
 &= \sum_y p(x,y)
 \end{aligned}$$

Similarly, $p_Y(y) = \sum_x p(x,y)$

Continuous case

Defn: The joint probability density function is a function $f(x,y)$ such that

$$P(X \in A \cap Y \in B) = \int_B \int_A f(x,y) dx dy \quad (6)$$

Note: $F(a,b) = P(X \leq a \cap Y \leq b)$

$$= \int_{-\infty}^b \int_{-\infty}^a f(x,y) dx dy$$

$$F_X(a) = F(a, \infty) = \int_{-\infty}^{\infty} \int_{-\infty}^a f(x,y) dx dy$$

$$F_Y(b) = F(\infty, b) = \int_{-\infty}^b \int_{-\infty}^{\infty} f(x,y) dx dy$$

(7)

By the Fund. Thm. of Calculus in 2 dim,

$$\frac{\partial^2 F(x, y)}{\partial y \partial x} = f(x, y)$$

Example: Roll 2 dice. Let $X = \min$, $Y = \max$

$P(x, y)$		y						
x		1	2	3	4	5	6	
	1	$1/36$	$2/36$	$2/36$	$2/36$	$2/36$	$2/36$	$1/36$
	2	0	$1/36$	$2/36$	$2/36$	$2/36$	$2/36$	$9/36$
	3	0	0	$1/36$	$2/36$	$4/36$	$2/36$	$7/36$
	4	0	0	0	$1/36$	$2/36$	$2/36$	$5/36$
	5	0	0	0	0	$1/36$	$2/36$	$3/36$
	6	0	0	0	0	0	$1/36$	$1/36$

$P_X(x)$

(8)

Example: Let $f(x, y) = c(x^2 + y)$, $0 < x < 1$, $0 < y < 1$

Find c . $\int_0^1 \int_0^1 c(x^2 + y) dx dy \stackrel{\text{Set}}{=} 1$

$$= c \int_0^1 \left(\frac{x^3}{3} + yx \right) \Big|_{x=0}^1 dy$$

$$= c \int_0^1 \left(\frac{1}{3} + y - 0 \right) dy = c \left[\frac{1}{3}y + \frac{y^2}{2} \right]_{y=0}^1$$

$$= c \left(\frac{1}{3} + \frac{1}{2} - 0 \right) = c \cdot \frac{5}{6} \quad \therefore c = \frac{6}{5}$$

(9)

$$\begin{aligned}
 \text{Find } f_x(x) &= \int_0^1 f(x,y) dy \\
 &= \int_0^1 \frac{6}{5}(x^2+y) dy \\
 &= \frac{6}{5} \left(x^2 y + \frac{y^2}{2} \right) \Big|_{y=0}^1 \\
 &= \frac{6}{5} \left(x^2 + \frac{1}{2} - 0 \right) = \frac{6}{5} \left(x^2 + \frac{1}{2} \right) \quad 0 < x < 1
 \end{aligned}$$

(10)

$$\begin{aligned}
 \text{Find } f_y(y) &= \int_0^1 \frac{6}{5}(x^2+y) dx \\
 &= \frac{6}{5} \left(\frac{x^3}{3} + yx \right) \Big|_{x=0}^1 \\
 &= \frac{6}{5} \left(\frac{1}{3} + y - 0 \right) = \frac{6}{5} \left(\frac{1}{3} + y \right) \quad 0 < y < 1
 \end{aligned}$$

Defn. $E[g(X,Y)] = \begin{cases} \sum_x \sum_y g(x,y) p(x,y) & \text{discrete} \\ \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x,y) f(x,y) dx dy & \text{continuous} \end{cases}$

Prove that $E[aX + bY] = aE[X] + bE[Y]$

(11)

Proof: (continuous case)

$$E[aX + bY] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (ax + by) f(x, y) dx dy$$

$$= a \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f(x, y) dx dy + b \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y f(x, y) dx dy$$

$$= aE[X] + bE[Y]$$

(12)

Let $X \sim \text{Bin}(n, p)$

Then $X = X_1 + X_2 + \dots + X_n$,

where each $X_i = \begin{cases} 0 & \text{if failure} \\ 1 & \text{if success} \end{cases}$ on the i^{th} trial

$\therefore X_i \sim \text{Bern}(p)$

and we know $E[X_i] = p$

$$\therefore E[X] = \sum_{i=1}^n E[X_i] = np$$

Independence

(13)

Defn. The random variables X and Y are independent if $F(a,b) = F_X(a) F_Y(b)$
 $\forall a, b$

Note: (Continuous case)

Assume X & Y are independent.

$$\begin{aligned}\text{Then } f(x,y) &= \frac{\partial^2}{\partial y \partial x} F(x,y) \\ &= \frac{\partial^2}{\partial y \partial x} F_X(x) F_Y(y) \\ &= \frac{\partial}{\partial y} \frac{\partial}{\partial x} \underbrace{F_X(x) F_Y(y)} \\ &= \frac{\partial}{\partial y} F_Y(y) f_X(x) \\ &= f_X(x) f_Y(y)\end{aligned}$$

(14)

Assume $f(x,y) = f_X(x) f_Y(y) \quad \forall x, y$

$$\begin{aligned}\text{Then } F(a,b) &= \int_{-\infty}^b \int_{-\infty}^a f(x,y) dx dy \\ &= \int_{-\infty}^b \int_{-\infty}^a f_X(x) f_Y(y) dx dy\end{aligned}$$

$$= \int_{-a}^b \left[f_y(y) \underbrace{\int_{-a}^a f_x(x) dx}_{F_x(a)} \right] dy$$

(15)

$$= F_x(a) \underbrace{\int_{-a}^b f_y(y) dy}_{F_y(b)}$$

The mid-term exam is Tuesday Feb 14

1 page of notes
+ calculator