

CSE 107 11-10-25

1

Supplemental Lecture

Independence

let X, Y be jointly continuous r.v.s

Defn

X, Y are independent iff

$$f_{X,Y}(x,y) = f_X(x) \cdot f_Y(y)$$

for all $x, y \in \mathbb{R}$.

Equivalently

$$\bullet \quad f_{X|Y}(x|y) = f_X(x) \quad \left\{ \begin{array}{l} \text{for all } y \text{ s.t. } f_Y(y) > 0 \\ \text{and all } x \end{array} \right.$$

$$\bullet \quad f_{Y|X}(y|x) = f_Y(y) \quad \left\{ \begin{array}{l} \text{for all } x \text{ s.t. } f_X(x) > 0 \\ \text{and all } y \end{array} \right.$$

Generalization to 3 or more r.v.'s

$$f_{X,Y,Z}(x,y,z) = f_X(x) \cdot f_Y(y) \cdot f_Z(z)$$

If X, Y are independent, then any
 any events definable in terms
 of X and Y are also independent:

$$\{X \in I\} \text{ and } \{Y \in J\}$$

In Particular, if $I = (-\infty, x]$, $J = (-\infty, y]$:

$$F_{X,Y}(x,y) = P(X \leq x, Y \leq y)$$

$$= \int_{-\infty}^x \int_{-\infty}^y f_{X,Y}(t,s) ds dt$$

$$= \int_{-\infty}^x \int_{-\infty}^y f_X(t) f_Y(s) ds dt$$

$$= \int_{-\infty}^x f_X(t) \left(\int_{-\infty}^y f_Y(s) ds \right) dt$$

$$= \int_{-\infty}^x f_X(t) dt \cdot \int_{-\infty}^y f_Y(s) ds$$

$$= P(X \leq x) \cdot P(Y \leq y)$$

$$= F_X(x) \cdot F_Y(y)$$

Remark: converse is true, i.e.

$$F_{X,Y} = F_X \cdot F_Y \Rightarrow f_{X,Y} = f_X \cdot f_Y$$

$\nwarrow$
 showed

As in the discrete case, we have

$$(1) \quad E[XY] = E[X] \cdot E[Y]$$

$$(2) \quad E[g(X)h(Y)] = E[g(X)] \cdot E[h(Y)]$$

$$(3) \quad \text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$$

Exercise

- Prove (1) (mimic notes notes-10/21/25-Page 6.)
- Prove (2) (Hard: Prob. 44 on P. 134, adapt to continuous case)
- Prove (3) (Easy!)

3.6 Bayes' Rule

16

Recall

$$P(A|B) = \frac{P(B|A) \cdot P(A)}{P(B)}$$

If $A_1, \dots, A_n$ form a Partition of Ω ,
by total Prob. then

$$\begin{aligned} P(B) &= \sum_{i=1}^n P(B \cap A_i) \\ &= \sum_{i=1}^n P(B|A_i) P(A_i) \end{aligned}$$

so for any $k \in \{1, 2, \dots, n\}$

17

$$P(A_k | B) = \frac{P(B | A_k) \cdot P(A_k)}{\sum_{i=1}^n P(B | A_i) \cdot P(A_i)}$$

If X, Y are discrete r.v.s, this becomes

$$P_{X|Y}(x|y) = \frac{P_{Y|X}(y|x) \cdot P_X(x)}{\sum_{t \in \text{range}(X)} P_{Y|X}(y|t) \cdot P_X(t)}$$

In the continuous this becomes

$$f_{X|Y}(x|y) = \frac{f_{Y|X}(y|x) \cdot f_X(x)}{\int_{-\infty}^{\infty} f_{Y|X}(y|t) \cdot f_X(t) dt}$$

Ex.

Let Y be a normal r.v. with $\sigma=1$ and mean X , another r.v. that is continuous uniform on $[1, 3]$.

a) find $f_Y(y)$

Solution

We are given

$$f_X(x) = \begin{cases} \frac{1}{2} & \text{if } 1 \leq x \leq 3 \\ 0 & \text{otherwise} \end{cases}$$

and

$$f_{Y|X}(y|x) = \begin{cases} \frac{1}{\sqrt{2\pi}} e^{-\frac{(y-x)^2}{2}} & \text{if } 1 \leq x \leq 3, y \in \mathbb{R} \\ 0 & \text{otherwise} \end{cases}$$

$\Rightarrow$

$$f_Y(y) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) dx$$

$$= \int_{-\infty}^{\infty} f_X(x) \cdot f_{Y|X}(y|x) dx$$

$$= \frac{1}{2} \int_1^3 \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-y)^2}{2}} dx \quad \begin{cases} u = x - y \\ du = dx \end{cases}$$

$$= \frac{1}{2} \int_{1-\gamma}^{3-\gamma} \frac{1}{\sqrt{2\pi}} \cdot e^{-u^2/2} du$$

$$= \frac{1}{2} (\Phi(z - \gamma) - \Phi(1 - \gamma)) \quad \text{///}$$

b.) find $f_{X|Y}(x|y)$

Solution

$$f_{X|Y}(x|y) = \frac{f_{Y|X}(y|x) \cdot f_X(x)}{f_Y(y)}$$

$$= \begin{cases} \frac{\frac{1}{\sqrt{2\pi}} e^{-\frac{(y-x)^2}{2}}}{\Phi(3-y) - \Phi(1-y)} & \begin{array}{l} \text{if} \\ 1 \leq x \leq 3 \\ y \in \mathbb{R} \end{array} \\ 0 & \text{otherwise} \end{cases}$$

c.) Suppose we sample Y and get $y=3$. What is the probability that $X \leq 2$, i.e. find

$$P(X \leq 2 | Y=3)$$

Exercise: write integral, do

ch. of var. $\begin{cases} u = x - 3 \\ du = dx \end{cases}$

d.) Find $E[Y]$

Exercise: do this