

# LECTURE 28

04-11-18

$$E[Y|X=x] = \begin{cases} \sum_{y \in S_Y} y P_{Y|X=x}(y) & (X,Y) \text{ DISCRETE} \\ \int_{-\infty}^{\infty} y f_{Y|X=x}(y) dy & (X,Y) \text{ CONTINUOUS} \end{cases}$$

$$\text{VAR}(Y|X=x) = E[Y^2|X=x] - (E[Y|X=x])^2$$

$$E[X|Y=y], \quad \text{VAR}[X|Y=y]$$

	0	1	2	3
0	0.84	0.03	0.02	0.01
1	0.06	0.01	0.008	0.002
2	0.01	0.005	0.004	0.001

$$V[Y|X=0] \quad E[Y|X=0] = 0 \left( \frac{0.84}{0.9} \right) + 1 \left( \frac{0.03}{0.9} \right) + 2 \left( \frac{0.02}{0.9} \right) + 3 \left( \frac{0.01}{0.9} \right)$$

$$= \frac{3}{90} + \frac{4}{90} + \frac{3}{90} = \frac{1}{9}$$

$$E[Y^2|X=0] = 1 \left( \frac{0.003}{0.9} \right) + 4 \left( \frac{0.02}{0.9} \right) + 9 \left( \frac{0.01}{0.9} \right)$$

$$= \frac{3+8+9}{90}$$

$$\text{VAR}[Y|X=0] = \frac{2}{9} - \frac{1}{81} = \frac{18}{81} - \frac{1}{81} = \frac{17}{81}$$

$$f(x,y) = \begin{cases} \frac{e^{-x/4} e^{-y}}{4} & (x > 0, y > 0) \\ 0 & \text{OTHERWISE} \end{cases}$$

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$$E[Y|X=3]$$

$$V[Y|X=3]$$

$$f_{X|Y}(y) = \frac{f(X, Y)}{f_Y(y)}$$

$$f_Y(y) = \int_0^{\infty} \frac{e^{-x/4} e^{-y}}{y} dx = \frac{e^{-y}}{y} \left[ -y e^{-x/4} \right]_0^{\infty}$$

$$= \frac{e^{-y}}{y} [y] = e^{-y}$$

$$f_{X|Y}(y) = \frac{e^{-x/4} e^{-y}}{y e^{-y}} = \frac{e^{-x/4}}{y}$$

$$f_{X|Y=3}(y) = \frac{e^{-x/3}}{3}$$

$$f_{X|Y=3}(y) = \frac{e^{-x/3}}{3} \quad \hookrightarrow \text{EXPONENTIAL } \lambda = \frac{1}{3}$$

$$E_c(X|Y=3) = 3$$

$$V(X|Y=3) = 9$$

CALCULATE &amp; UNDERSTAND THIS.

THE REGRESSION FUNCTION

(SYNONYM FOR CONDITIONAL EXPECTATION)

$E[Y|X=x]$  REGRESSION FUNCTION OF  $Y$  ON  
 $X$  DENOTED BY  $Y|X$

SO YOU CAN USE THE SAME FORMULA TO CALC. 04-11-18  
THEM

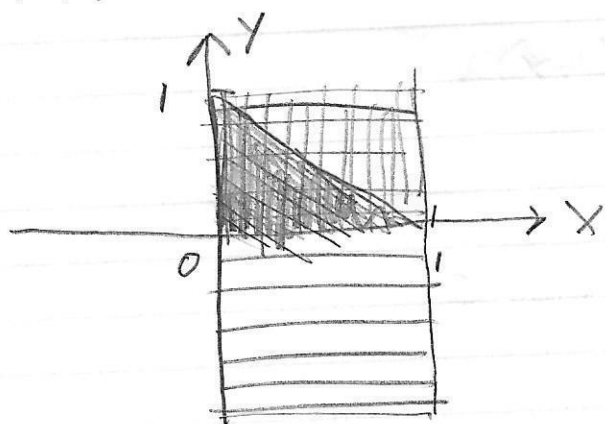
BECAUSE THE JOINT DISTRIBUTION OF  $X, Y$  IS UNKNOWN  
CLASSICAL LINEAR REGRESSION MODEL EXPLORES  
ONLY THE RELATIONSHIP OF THE FORM

$$\mu_{Y|X}(x) = \alpha_1 + \beta_1 x$$

$\alpha_1, \beta_1$  ARE PARAMETERS TO BE ESTIMATES FROM  
THE DATA.

$(X, Y)$  HAS JOINT PDF

$$f(x, y) = \begin{cases} 24xy & 0 \leq x \leq 1-y \\ & 0 \leq y \leq 1 \\ & x+y \leq 1 \\ 0 & \leftarrow \text{OTHERWISE} \end{cases}$$



$$f_{Y|X}(y) = \frac{f(x, y)}{f_Y(y)}$$

$$\begin{aligned} f_Y(y) &= \int_0^{1-y} 24xy \, dx \\ &= 24y \frac{x^2}{2} \bigg|_0^{1-y} \\ &= 12y (1-y)^2 \\ &= 12y(1-y)^2 \quad 0 \leq y \leq 1 \end{aligned}$$

$$E[Y|X] =$$

$$f_{X|Y}(y) = \frac{f(x, y)}{f_Y(y)}$$

$$= \frac{24xy}{12y(1-y)^2} = \frac{2x}{(1-y)^2} \quad 0 \leq x \leq 1-y$$

THE RANGE OF  $X$  DEPENDS ON  $Y$ .

IN FACT  $0 \leq x \leq 1-y$

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$$E[Y|X] = \int_0^{1-x}$$

BY SYMMETRY  $f_Y(x) = \int_0^{1-x} 24xy \, dy$   
 $= 12x(1-x)^2 \quad 0 \leq y \leq 1$

$$f_{Y|X}(y) = \frac{24xy}{12x(1-x)^2} = \frac{2y}{(1-x)^2} \quad 0 \leq y \leq 1-x$$

$$E[Y|X] = \int_0^{1-x} \frac{2y}{(1-x)^2} dy = \left. \frac{2y^2}{2(1-x)^2} \right|_0^{1-x}$$

$$= \frac{2(1-x)^2}{2(1-x)^2} = \frac{2}{3}(1-x)$$

$$\mu_{Y|X} = \frac{2}{3} - \frac{2}{3}x$$

↑ INTERCEPT      SLOPE

FOR A LARGER  $x$ , EXPECT A SMALLER  $y$

INDEPENDENCE OF RANDOM VARIABLES

(CH 2 INDEPENDENCE OF EVENTS)

A, B EVENTS  $P(A \cap B) = P(A)P(B|A)$

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TWO RANDOM VARIABLES ARE INDEPENDENT IF FOR ANY  $x \in \mathbb{R}^1$  &  $y \in \mathbb{R}^1$

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

IN TERMS OF DISTRIBUTION FUNCTIONS,

$$F_{X,Y}(x,y) = F_X(x)F_Y(y) \quad \forall x,y \in \mathbb{R}^1$$

CHARACTERIZED INDEPENDENCE WITH PMF/PDF.

$(X,Y)$  DISCRETE

$$X \perp\!\!\!\perp Y \iff p(x,y) = p_X(x)p_Y(y) \quad \forall (x,y) \in S_{X,Y}$$