

• midterm 1: 1-31-25 Friday.

• hw3: Problem 8 added.

Let X be geometric with param p .

Let $A_1 = \{X=1\} = \{1^{\text{st}} \text{ flip is head}\}$

$A_2 = \{X>1\} = \{1^{\text{st}} \text{ flip is tail}\}$

so $P(A_1) = p$ and $P(A_2) = 1-p$.

note $A_1 \cap A_2 = \emptyset$ and $A_1 \cup A_2 = \Omega$

Also $E[X|A_1] = 1$, and

$$E[X|A_2] = E[1+X] = 1 + E[X]$$

Hence, by total expectation theorem

$$E[X] = P(A_1) \cdot E[X|A_1] + P(A_2) \cdot E[X|A_2]$$

$$= p \cdot 1 + (1-p) \cdot (1 + E[X])$$

$$E[X] = p + (1-p) + E[X] - pE[X]$$

$$\therefore pE[X] = 1$$

$$\therefore E[X] = \boxed{\frac{1}{p}} \quad \checkmark$$

Similarly

$$E[X^2 | A_1] = 1$$

and

$$E[X^2 | A_2] = E[(1+X)^2]$$

$$= E[1 + 2X + X^2]$$

$$= 1 + 2E[X] + E[X^2]$$

hence

$$E[X^2] = P(A_1) \cdot E[X^2 | A_1] + P(A_2) \cdot E[X^2 | A_2]$$

$$= P \cdot 1 + (1-P) \left(1 + 2 \cdot \frac{1}{P} + E[X^2] \right)$$

$$= \cancel{P} + (1-\cancel{P}) + (1-P) \frac{2}{P} + \cancel{E[X^2]} - P E[X^2]$$

$$P E[X^2] = 1 + \frac{2-2P}{P} = \frac{P+2-2P}{P} = \frac{2-P}{P}$$

$$\therefore E[X^2] = \frac{2-p}{p^2} = \frac{2}{p^2} - \frac{1}{p} \quad \checkmark$$

finally

$$\text{Var}(X) = \left(\frac{2}{p^2} - \frac{1}{p} \right) - \frac{1}{p^2}$$

$$= \frac{1}{p^2} - \frac{1}{p} = \boxed{\frac{1-p}{p^2}} \quad \checkmark$$

2.7 Independence

Defn

We say two r.v.s X, Y are

independent iff for all

$x \in \text{range}(X), y \in \text{range}(Y)$ we have

$$\{X=x\}, \{Y=y\}$$

are independent events, i.e.

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

equivalently

$$P_{X,Y}(x,y) = P_X(x) \cdot P_Y(y)$$

This is equivalent to

$$\frac{P_{X,Y}(x,y)}{P_Y(y)} = P_X(x)$$

i.e.

$$P_{X|Y}(x|y) = P_X(x)$$

□

Defn

We say 2 r.v.s X, Y are conditionally indep. given event A

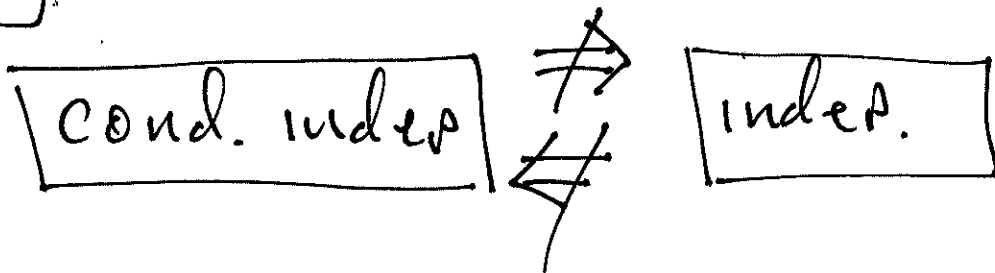
iff

$$P(X=x, Y=y|A) = P(X=x|A) \cdot P(Y=y|A)$$

i.e.

$$P_{X,Y|A}(x,y) = P_{X|A}(x) \cdot P_{Y|A}(y)$$

Warning:



see fig. 2.15 on p.111

Theorem

If X, Y are independent r.v.s,
then

$$E[X \cdot Y] = E[X] \cdot E[Y]$$

Proof

$$E[XY] = \sum_x \sum_y xy P_{X,Y}(x,y)$$

$$= \sum_x \sum_y xy P_X(x) \cdot P_Y(y)$$

$$= \left(\sum_x x P_X(x) \right) \cdot \left(\sum_y y P_Y(y) \right)$$

$$= E[X] \cdot E[Y]$$



Thus

If X, Y are indep., then so are $g(X)$ and $h(Y)$.

see Problem #44, p. 134 for proof.

Thus

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

Theorem

if X, Y are indep., then

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

Proof

let X, Y be indep. define

$$U = X - E[X]$$

$$V = Y - E[Y]$$

so

$$E[U^2] = \text{Var}(X)$$

and $E[V^2] = \text{Var}(Y)$

Also

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$$E[U] = E[X] - E[X] = 0$$

$$E[V] = 0$$

Also

$$U+V = (X+Y) - E[X+Y]$$

Thus

$$\text{Var}(X+Y) = E[(U+V)^2]$$

$$= E[U^2 + 2UV + V^2]$$

$$= E[U^2] + 2E[UV] + E[V^2]$$

$$E[U] \cdot E[V]$$

$$0 \cdot 0$$

1.2

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

