Conditional PMFs

$$P_{X|A}(x|A) = P(X=x|A)$$

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

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

$$= P(Y=y)$$

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

$$= \frac{\int_{X|Y,Z} (x,y)}{\int_{X|Y,Z} (x,y,z)}$$

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2. 
$$P_{X,Y|Z}(x,y|Z) = \frac{P_{X,Y,Z}(x,y,Z)}{P_{Z}(Z)}$$

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Conditional expectation

$$E[X|A] = \sum_{x} x P_{X|A}(x)$$

$$A = \{Y = y\}$$
  $\boxed{T[X|Y = y]} = \sum_{x} x P_{X|Y}(x|y)$ 

Total probability of E

$$A_1$$
,  $A_2$ — $A_n$ : partition of  $\Omega$ 

$$P_X(x) = P(A_1) P_{X|A_1}(x) + \cdots$$

Independent

of r.u. 
$$P(X=x \text{ and } A) = P(X=x)P(A)$$
 for all  $x \in P_{X|A}(x) = P_{X}(x)$   
 $P(A|X=x) = P(A)$ 

of two r.v. 
$$P(x=x \ (Y=y) = P(X=x) P(Y=y)$$

$$P_{x,y}(x,y) = P_{x}(x) P_{y}(y)$$

$$P_{x|y}(x|y) = P_{x}(x)$$

$$P_{y|x}(y|y) = P_{y}(y)$$

Independence E

It X, Y are independent

Proof 
$$[ [g(x,y)] g(x,y) = xy$$

$$= \sum_{x,y} [x,y] P_{x,y}(x,y) = \sum_{x,y} [x,y] P_{x}(x) P_{x}(y)$$

$$= \sum_{x} > (P_{x}(x)) \sum_{y} y P_{y}(y)$$



## Independence l'uniance

Proof 
$$V(X+Y) = E[(X+Y)^2] = E[X^2+2XY+Y^2]$$

Variance of binomial
$$X = X_1 + \cdots \times N \longrightarrow \frac{1}{2} \frac{1}{$$