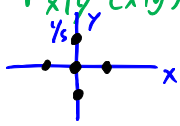


JOINT PMFs

$P_{X,Y}(x,y) = P_r((X=x) \cap (Y=y)) \rightarrow$ joint

$P_{X|Y}(x|y) = P_{X,Y}(x,y) / P_Y(y) \rightarrow$ marginal

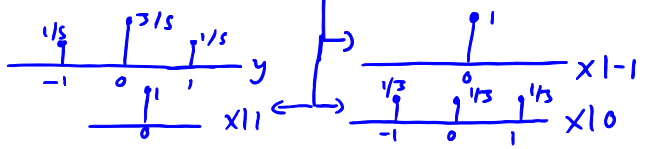
EX



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

independent?
for $(x,y)=(1,1)$, 0

$P_{X,Y}(x,y) \stackrel{?}{=} P_X(x) P_Y(y)$
 $\neq \frac{1}{5} \cdot \frac{1}{5}$



$P_{X|Y}(x|y) = \frac{P_{Y|X}(y|x) P_X(x)}{P_Y(y)}$ (Bayes!)

VARIANCE FOR SUM OF RVs

$X = RV$

$\sigma_x^2 = E(X^2) - E^2(X)$

$\sigma_z^2 = E(Z^2) - E^2(Z)$

$Z = X+Y$

$E(Z) = E(X) + E(Y)$

$= E[(X+Y - E(X) - E(Y))^2]$

$\sigma_x^2 + \sigma_y^2 + 2\sigma_{xy} \leftarrow = E[(X - E(X)) + (Y - E(Y))]^2$

COVARIANCE

$\sigma_{xy} = cov(X,Y) = E[(X - E(X))(Y - E(Y))] = E(XY) - E(X)E(Y)$

CLAIM

If X, Y indep, then X, Y uncorrelated ($\sigma_{xy} = 0$).

PROOF

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

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

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

$\sigma_{xy} = E(XY) - E(X)E(Y) = 0 \leftarrow = \sum_x x P_X(x) \sum_y y P_Y(y) = E(X)E(Y)$

The reverse isn't necessarily true.

MARKOV INEQUALITY

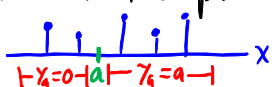
X nonnegative RV

$P(X \geq a) \leq \frac{E(X)}{a} \quad \forall a > 0$

If $E(X)$ is small, the probability that X takes large values is small.

PROOF

$\gamma_a = \begin{cases} 0 & x < a \\ a & x \geq a \end{cases}$



$\gamma_a \leq X \rightarrow$

$E(\gamma_a) = a P(X \geq a) \leq E(X)$



CHEBYSHEV'S INEQUALITY

X RV $\mu = E(X) \quad \sigma^2 = \sigma_x^2 \quad P(|X - \mu| \geq c) \leq \frac{\sigma^2}{c^2} \quad \forall c > 0$

Probability that X deviates from mean by c is bounded by $\frac{\sigma^2}{c^2}$. Prove w/Markov!

PROOF

$Y = (X - \mu)^2, \quad a = c^2 \quad Y \geq 0, \quad P(Y \geq a) = P(Y \geq c^2) \leq \frac{E(Y)}{a} = \frac{E((X - \mu)^2)}{c^2}$

$P((X - \mu)^2 \geq c^2) \leq \frac{\sigma^2}{c^2}$ the left side here is the same as $P(|X - \mu| \geq c)$

