

Q1 Derive a generalized version of Bayes' rule stated as

$$P(X|Y, Z) = \frac{P(Y|X, Z) \cdot P(X|Z)}{P(Y|Z)}$$

A generalized version would be  $P(X|Y, Z, \lambda)$

$$P(X|Y, Z, \lambda) = \frac{P(Y, X, Z, \lambda)}{P(Y, Z, \lambda)}$$

Applying chain rule to numerator and denominator.

$$= \frac{P(Y|X, Z, \lambda) \cdot P(X|Z, \lambda) \cdot P(Z|\lambda) \cdot P(\lambda)}{P(Y|Z, \lambda) \cdot P(Z|\lambda) \cdot P(\lambda)}$$

$$= \frac{P(Y|X, Z, \lambda) \cdot P(X|Z, \lambda) \cdot P(Z|\lambda) \cdot P(\lambda)}{P(Y|Z, \lambda) \cdot P(Z|\lambda) \cdot P(\lambda)}$$

$$= \frac{P(Y|X, Z, \lambda) \cdot P(X|Z, \lambda)}{P(Y|Z, \lambda)}$$

Q2.

Symmetry

$$(X \perp Y | Z) \Rightarrow (Y \perp X | Z)$$

Assume  $(X \perp Y | Z)$  holds

then

$$P(X, Y | Z) = P(X | Z) P(Y | Z)$$

$\therefore$  Multiplication is commutative

$$P(X, Y | Z) = P(Y | Z) P(X | Z) = P(Y, X | Z)$$

Thus  $(Y \perp X | Z)$  also holds.

Decomposition

Assume  $(X \perp Y, W | Z)$  holds.

~~Then~~  $P(X, Y, W | Z) = P(X | Z) P(Y, W | Z)$

$\because P(A \cap B | C) = P(A | C) P(B | C)$

No we show,

$$\begin{aligned} P(X, Y, W | Z) &= P(X | Z) P(Y, W | Z) \\ &= \sum_w P(X, Y, w | Z) \\ &= \sum_w P(X | Z) P(Y, w | Z) \\ &= P(X | Z) \sum_w P(Y, w | Z) \\ &= P(X | Z) P(Y | Z) \end{aligned}$$

$\therefore$  ~~By definition~~

Q3.

Variance is written as

$$\text{Var}_p[X] = E_p[(X - E_p[X])^2]$$

Expanding square

$$= E_p[X^2 + E_p^2[X] - 2XE_p[X]]$$

∵ linearity

$$= E_p[X^2] + E_p^2[X] - 2E_p[XE_p[X]]$$

$$\because E_p[E_p[X]] = E_p[X] \text{ and}$$

$$= E_p[X^2] + E_p^2[X] - 2E_p^2[X]$$

$$= E_p[X^2] - E_p^2[X]$$

$$= \boxed{\therefore E[X^2] - (E[X])^2}$$

Q4.

$$E[X] = \mu$$

$$E[Y] = E[aX + b]$$

$$= aE[X] + E[b] \quad \text{Applying linearity}$$

$$= aE[X] + b$$

$$= a\mu + b$$

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

$$\text{Var}[Y] = E[(aX + b)^2] - (E[aX + b])^2$$

$$= E[a^2X^2 + b^2 + 2abX] - (\cancel{E[aX]} + b)^2$$

$$= a^2 E[X^2] + \cancel{b^2} + \cancel{2abE[X]}$$

$$- (a^2(E[X])^2 + \cancel{b^2} + \cancel{2abE[X]})$$

$$= a^2 E[X^2] - a^2 (E[X])^2$$

$$= a^2 (E[X^2] - (E[X])^2)$$

$$= a^2 \sigma^2$$