

- ① a) If W is a random variable representing the negation of X and Z is a random variable representing the negation of Y , then X is also independent of Z and Y is independent of W . So we have

$(X=0) \perp (Y=0)$, $(X=0) \perp (Y=1)$, $(X=1) \perp (Y=0)$ and if $(X=0) \perp (Y=0)$ then $(X \neq 0) \perp (Y \neq 0)$ so all 4 possibilities are independent, therefore $X \perp Y$

- b) For non binary variables there are multiple possibilities in the negation, so $(X=0) \perp (Y=0)$ does not imply $(X \neq 0) \perp (Y \neq 0)$ because there is more than one possibility for $X \neq 0$ and $Y \neq 0$

- ② X_1 is conditionally independent of X_2 given X_3 in a distribution P if P satisfies $(X=X_1 \perp Y=X_2 \mid Z=X_3)$ for all values $x_1 \in \text{Val}(X_1)$, $x_2 \in \text{Val}(X_2)$, and $x_3 \in \text{Val}(X_3)$. The independence statement is universal for all values of x_1, x_2 , and x_3 .

Since $(X_1 \perp X_2 \mid X_3)$ we have $P(X_1 \mid X_2, X_3) = P(X_1)$, thus $P(X_1 \cap X_2 \mid X_3) = P(X_1 \mid X_3) P(X_2 \mid X_3)$

If we let X_3 take on all possible values we have

$$P(X_1 \cap X_2) = \phi_1(X_1) \phi_2(X_2)$$

Letting X_1 and X_2 take all values we can write

$$P(X) = \phi_1(X_1) \phi_2(X_2)$$

- ③ Variance is the expectation of the squared difference between X and its expected value. $\Rightarrow \text{Var}[X] = E[(X - E[X])^2]$

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

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

Since $E[X]$ is just a number it can be factored out so

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

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

④ $I(X;Y) = E_P \left[\log \frac{P(X|Y)}{P(X)} \right]$ If X and Y are independent then $P(X|Y) = P(X)$ so $E_P [\log(1)] = 0$
 If X and Y are not independent then knowing what Y is will give you more information about the value of X so $\frac{P(X|Y)}{P(X)} > 1$ and $E_P [\log(x>1)] > 0$. So $I(X;Y) \geq 0$

⑤ If X and Y are independent then $P(X|Y) = P(X)$ so
 $I(X;Y) = E_P \left[\log \frac{P(X|Y)}{P(X)} \right] = E_P \left[\log \frac{P(X)}{P(X)} \right] = E_P [\log(1)] = 0$

⑥ a) B, C, D, F b) A, E, F c) Yes d) No e) Yes f) Yes
 g) Yes h) No i) No j) No, because B, C are not connected by an edge
 k) Yes, because every node in $\{B, E, F\}$ is connected by an edge and adding more nodes outside the subgraph cannot produce a clique
 l) B, C, D, F, E, A