

Q A 1.

A  $5+14+5=24$

5/5

i. T since given  $Y=1$ ,  $X$  either equals to 0 or 1.

ii. F consider  $X \begin{matrix} 0.7 & 0.2 \\ 0.3 & 0.8 \end{matrix} Y$  then  $p(X=1|Y=1) + p(X=1|Y=0)$   
 $= 0.8 + 0.3 \neq 1$

iii. F since  $p(X=1, Y=1, Z=0) = 1 - p(X=1, Y=1, Z=1)$   
 $= 1 - p(X=1)p(X \neq 1)p(Z=1)$   
 $= 1 - p(X=1)p(Y=1)p(Z=0)$   
 $= 1 - p(X=1)p(Y=1) + p(X=1)p(Y=1)p(Z=0)$   
 $= p(X=1)p(Y=1)p(Z=0) \neq$   
 $1 - p(X=1)p(Y=1) = 0$  which is not always true

iv. T  $\frac{p(X=0, Y=0)}{p(X=0, Y=1)} = \frac{p(Y=0, X=0)}{p(Y=1, X=0)}$   
 $= \frac{p(Y=0|X=0)p(X=0)}{p(Y=1|X=0)p(X=0)}$   
 $= \frac{p(Y=0|X=0)}{p(Y=1|X=0)}$

v. T  $p(X=0, Y=0) + p(X=1, Y=0) = \frac{p(X=0, Y=0) + p(X=1, Y=0)}{p(X=0, Y=0)}$   
 $= p(Y=0)$   
 $= \frac{p(X=0, Y=0)}{p(X=0, Y=0)} p(Y=0)$   
 $= \frac{p(X=0, Y=0)}{p(X=0|Y=0)p(Y=0)}$   
 $= \frac{p(X=0, Y=0)}{p(X=0|Y=0)}$

14/14

$$p(w=1) = 0.9$$

$$p(w=1 | h=1) = 0.95$$

$$p(w=1 | h=0) = 0.85$$

$$i) \quad \cancel{p(h=1)} =$$

$$\text{we know } p(w=1) = p(w=1 | h=1) p(h=1) + p(w=1 | h=0) p(h=0)$$

$$\text{so } p(w=1) = p(w=1 | h=1) p(h=1) + p(w=1 | h=0) p(h=0) - p(w=1 | h=0) p(h=1)$$

$$\text{sub. numbers in } 0.9 = 0.95 p(h=1) + 0.85 - 0.85 p(h=1)$$

$$0.05 = 0.1 p(h=1)$$

$$\text{so } p(h=1) = 0.5$$

$$ii) \quad p(h=1 | w=1) = \frac{p(w=1 | h=1) p(h=1)}{p(w=1)} \quad \text{using Bayes theorem}$$

$$= \frac{0.95 \times 0.5}{0.9}$$

$$= \frac{19}{36}$$

$$iii) \quad p(I=1 | h=1) = 0.9$$

$$p(w=1 | h=1, I=1) = 2 p(w=1 | h=1, I=0)$$

we want

$$p(w=1 | h=1, I=0)$$

we know that  ~~$p(w=1) = p$~~

$$p(w=1 | h=1) = p(w=1, I=1 | h=1) + p(w=1, I=0 | h=1)$$

$$= p(w=1 | I=1, h=1) p(I=1 | h=1)$$

$$+ p(w=1 | I=0, h=1) p(I=0 | h=1)$$

$$= 2 p(w=1 | h=1, I=0) p(I=1 | h=1)$$

$$+ p(w=1 | I=0, h=1) p(I=1 | h=1)$$

substitute numbers in:

$$0.95 = 2 \times 0.9 \times p(w=1 | h=1, I=0) + 0.1 p(w=1 | I=0, h=1)$$

$$\text{we get } p(w=1 | h=1, I=0) = \frac{0.95}{1.9} = \frac{1}{2}$$

(5/6)

i. the maximum likelihood estimate of  $\theta_h$  is 0.6

✓ 2

ii. As the number of trials  $N$  increase, the likelihood <sup>estimates</sup> approaches true probabilities of the underlying distribution. ~~LLN?~~

LLN?

iii. 
$$p(\theta | D') = \frac{p(D' | \theta) p(\theta)}{p(D')}$$
 Bayes theorem

which expresses the posterior probability in terms of the prior and the evidence, doesn't depend on  $D'$  ~~alone~~ alone.

max  $p(\theta | D')$

$$B \quad 12+8+4=24$$

12/12

$$\begin{aligned} i) \quad H(X) &= \frac{1}{2} \log_2 2 + \frac{1}{3} \log_2 3 + 2 \times \frac{1}{12} \log_2 12 \\ &= \frac{1}{2} + \frac{1}{3} \log_2 3 + \frac{1}{6} \log_2 4 + \frac{1}{6} \log_2 3 \\ &\approx \frac{1}{2} + \frac{1}{3} + \frac{1}{2} \times 1.58 \\ &= 1.623 \end{aligned}$$

$$ii) \quad p(Y) =$$

$$\begin{aligned} p(Y = \text{Naks}) &= \frac{1}{6} \times \frac{1}{2} + \frac{1}{4} \times \frac{1}{3} + \frac{1}{2} \times \frac{1}{12} \times 2 \\ &= \frac{3}{12} = \frac{1}{4} \end{aligned}$$

$$p(Y = \text{Ray}) = 1 - \frac{1}{4} = \frac{3}{4}$$

$$\begin{aligned} \text{Hence } H(Y) &= \frac{1}{4} \log_2 4 + \frac{3}{4} \log_2 \frac{4}{3} \\ &= \frac{1}{2} + \frac{3}{2} - \frac{3}{4} \log_2 3 \\ &\approx 2 - \frac{3}{4} \times 1.58 \\ &\approx 0.8 \end{aligned}$$

$$iii) \quad H(Y|X=a) \approx 0.65 < 0.8 < 1 = H(Y|X=d)$$

$$\text{Note } H(Y|X=d) = \log_2 2 = 1$$

$$\begin{aligned} \text{and } H(Y|X=a) &= \frac{1}{6} \log_2 6 + \frac{5}{6} \log_2 \frac{6}{5} \\ &= \frac{1}{6} + \frac{1}{6} \log_2 3 + \frac{5}{6} \log_2 6 - \frac{5}{6} \log_2 5 \\ &= \frac{1}{6} + \frac{5}{6} + \log_2 3 - \frac{5}{6} \log_2 5 \\ &\approx 1 + 1.58 - \frac{5}{6} \times 2.32 \\ &\approx 0.65 \end{aligned}$$

iv. because  $[H(Y|X)$  contains less information] than  $H(Y)$  since given  $X$  we are less uncertain about  $Y$  (unless  $X \perp Y$ , for which  $H(Y|X) = H(Y)$ )

$$v. \quad H(Z|X) = 0 \text{ since } p(Z|X=a, b, c) = (1, 0) \text{ and } p(Z|X=d) = (0, 1)$$

this is because  $Z$  is deterministic in terms of  $X$ . given  $x$  there are no uncertainty left for the value of  $Z$ .

(8/9)

i) if  $(X, Y, Z)$  form a Markov chain, then

$$p(X, Y, Z) = p(X) p(Y|X) p(Z|Y) \quad \text{and} \quad X \perp\!\!\!\perp Z | Y$$

but then

$$\begin{aligned} p(Z, Y, X) &= p(Z) p(Y|Z) p(X|Y, Z) \\ &= p(Z) p(Y|Z) p(X|Y) \quad \text{since } X \perp\!\!\!\perp Z | Y \end{aligned}$$

hence  $(Z, Y, X)$  also form a Markov chain

ii) the data processing inequality states that (for the Markov chain  $(X, Y, Z)$ )

$$I(X; Y) \geq I(X; Z)$$

since  $(Z, Y, X)$  also forms a Markov chain, ~~and~~ we have

$$I(Z; Y) \geq I(X; Z)$$

$$\text{and so we get } I(X; Z) \leq \min(I(X; Y), I(Y; Z))$$

since mutual information is symmetric

Intuition?

iii) let  $X, Y, Z \in \{0, 1\}$  and  ~~$y = \begin{cases} 0 & \text{if } x=0, 1 \end{cases}$~~

and if  $x=0$ , then  $y=0$  or  $1$  with equal chance,

same when  $x=1$ .  $p(X) = (0.5, 0.5)$

and for  ~~$x$~~   $y=0$ ,  $z=0$  and  $y=1$  implies  $z=1$ .

then

$$\begin{aligned} I(X; Y) &= H(Y) - H(Y|X) \\ &= 1 - 1 = 0 \end{aligned}$$

$$\begin{aligned} \text{and } I(Y; Z) &= H(Z) - H(Z|Y) \\ &= 1 - 0 = 1 \end{aligned}$$

$$\text{i.e. } I(X; Y) < I(Y; Z)$$

$$i) E[Z] = E[X] + E[Y] = 8000$$

4/4

$$ii) P(Z \geq a) < \frac{E[Z]}{a}$$

$$P(Z \geq 20,000) < \frac{E[Z]}{20,000}$$

$$= 0.4$$

iii) since  $X$  and  $Y$  are dependent, exploring every possible combination and count is tedious.