

## Qair-2 sol<sup>n</sup>

Q.1

(a) Assume  $A, B$  &  $C$  be 3 collections of RV

containing  $X, Y$  &  $Z$  respectively. So each collection is a non-empty collection of RV with 1 RV in each.

$$\therefore I(A; B|C) \geq 0$$

$$\therefore H(A|C) - H(A|B, C) \geq 0$$

(2)

$$\therefore H(A|C) \geq H(A|B, C)$$

Replacing collections by the RV in the respective collection:

$$\therefore H(X|Z) \geq H(X|Y, Z)$$

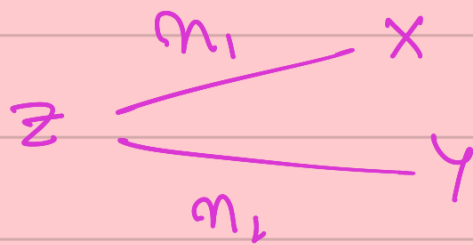
(2)

Also, we are given that equality holds if  $A$  &  $B$  are independent given  $C$  which here means  $X$  &  $Y$  are independent given  $Z$ . So, here equality holds if  $X$  &  $Y$  are independent given  $Z$ .

(don't cut marks)

Comm<sup>n</sup> Setup interpretation: Not Necessary

Signal observed by two people



independent

$n_1$  &  $n_2 \rightarrow$  noise

here  $H(X|Z) = H(X|Y, Z)$

$$(b) \quad H(X_1, \dots, X_n | Y) = \sum_{y \in \text{supp}(P_Y)} P_Y(y) H(X_1, \dots, X_n | Y=y)$$

$$(2) \quad \left\{ \begin{array}{l} \text{where } H(X_1, \dots, X_n | Y=y) = - \sum_{(x_1, \dots, x_n) \in \text{supp}(P_{X_1, \dots, X_n | Y=y})} P_{X_1, \dots, X_n | Y=y}(x_1, \dots, x_n) \log P_{X_1, \dots, X_n | Y=y}(x_1, \dots, x_n) \end{array} \right.$$

Chain Rule states that

$$H(X_1, \dots, X_n | Y) = H(X_1 | Y) + H(X_2 | X_1, Y) + H(X_3 | X_1, X_2, Y) \\ + \dots + H(X_n | X_1, X_2, \dots, X_{n-1}, Y)$$

To prove this,

$$H(X, Y | Z) = \sum_z P(Z=z) H(X, Y | Z=z)$$

$$= - \sum_{x, y, z} P(X=x, Y=y | Z=z) P(Z=z) \log P(X=x, Y=y | Z=z)$$

$$= - \sum_{x, y} P(X=x | Z=z) P(Z=z) \log P(X=x | Z=z)$$

$$- \sum_{x, y, z} P(Y=y | X=x, Z=z) P(X=x, Z=z) \log P(Y=y | X=x, Z=z)$$

$$= H(X | Z) + H(Y | X, Z)$$

$X \& Y$

we showed this for two variables. it will  
be similarly extended to  $n$  variables as

$$P(X_1=x_1, X_2=x_2, \dots, X_n=x_n | Y=y) = P(X_1=x_1 | Y=y) P(X_2=x_2 | X_1=x_1, Y=y) \\ \dots P(X_n=x_n | X_1=x_1, \dots, X_{n-1}=x_{n-1}, Y=y)$$

which comes from breaking the joint probability  
sequentially.

①  $\Rightarrow$  For  $H(X_1, \dots, X_n | Y) = \sum_{i=1}^n H(X_i | Y)$ , it

holds when all  $X_i$  are independent of  
all other  $X_j$  given  $Y$ . This comes from the  
understanding that

$$H(X_i | X_1, \dots, X_{i-1}, Y) = H(X_i | Y)$$

if  $X_i$  is independent of  $X_1, \dots, X_{i-1}$  given  $Y$ .

Q.2

$$X \sim \text{Ber}(0.5)$$

$$\Rightarrow P(X=0) = P(X=1) = 0.5$$

$$\textcircled{2} \quad \Rightarrow P_{Y/X}(0/0) = P_{Y/X}(1/1) = 1-p$$

$$P_{Y/X}(E/0) = P_{Y/X}(E/1) = p$$

$$\textcircled{1} \quad P_Y(0) = \sum_x P_{Y/X}(0/x) P_X(x)$$

$$= \frac{1-p}{2}$$

$$P_Y(1) = \frac{1-p}{2}$$

$$P_Y(E) = p$$

$$\therefore H(Y) = -p \log p - \frac{(1-p)}{2} \log \left( \frac{1-p}{2} \right) \times 2$$

$$= -p \log p - (1-p) \log(1-p) + (1-p)$$

$$H(Y/X) = \sum_x P_X(x) H(Y/X=x)$$

$$\textcircled{4} \quad H(Y|X=0) = H_2(P) = -p \log p - (1-p) \log (1-p)$$

$$H(Y|X=1) = H_2(P)$$

$$\therefore H(Y|X) = H_2(P)$$

$$\therefore I(X;Y) = \cancel{H_2(P)} + (1-p) - \cancel{H_2(P)}$$

$$I(X;Y) = 1-p$$

Q.3

Since  $X$  is binomial RV,

$$P(X=i) = {}^nC_i p^i (1-p)^{n-i}$$

here  $n=4$  &  $p=0.1$

$\frac{1}{2}$

$\therefore$

$x$	$P_x(x)$
0	$9^4 \times 10^{-4}$
1	$4 \times 9^3 \times 10^{-4}$
2	$6 \times 9^2 \times 10^{-4}$
3	$4 \times 9 \times 10^{-4}$
4	$10^{-4}$

$\frac{1}{21}$

$\therefore$

To get avg length short, we assign shorter codewords to values of  $X$  which have higher probability

value of $X$	codeword
'0'	0
'1'	10
'2'	110
'3'	1110
'4'	1111