

End Sem Exam 2023 Info & Communication Solutions.

1.

(a) Video : 30 fps. Each frame is 720×1080 p

Each pixel needs 8 bits (as it can take 255 values)

\Rightarrow Per second, we need bit rate of $30 \times 720 \times 1080 \times 8$

$$\approx 0.1867 \times 10^8 \text{ bps}$$

(b) Shannon entropy of each pixel is $\frac{3}{4} \log 4 + \frac{253}{1024} \log \frac{1024}{253}$
 ≈ 3.95 bits

Optimal source code can have at most avg length
 $= 4.95 \approx 5$ bits

Thus now bit rate requirement will be $\hookrightarrow H(x) + 1$

$$\approx 30 \times 720 \times 1080 \times 5$$

$$= 0.116 \times 10^9 \text{ bps.}$$

(c) Now in this, the entropy of the frame itself is only 1.371 bits.

Thus, we need bit rate $1.371 \times 30 \approx 42$ bps.

The channel cannot support this as the channel has sink rate only 1.2 bps.

$$2. (a) \underline{m} = (0 \ 0 \ 1 \ 1), \quad \underline{m} G = (0 \ 0 \ 1 \ 1) \begin{pmatrix} I_4 & \left| \begin{matrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \\ 1 & 1 & 1 \end{matrix} \right. \end{pmatrix}$$

Codeword is $= (0 \ 0 \ 1 \ 1 \ 1 \ 0 \ 0)$

(b) Suppose 6th bit is flipped, that is, we receive
 $(0 \ 0 \ 1 \ 1 \ 1 \ 1 \ 0)$.

The H matrix is $[P^T \ I_3] = \begin{pmatrix} 1 & 1 & 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 0 & 0 & 1 \end{pmatrix}$

Computing the syndrome $H y^T = \text{sum of columns } 3, 4, 5, 6,$
 which gives us $\underline{s} \xrightarrow{\downarrow} \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$

(Some students may not know this term)
 'Syndrome'

Thus, comparing with H matrix, we see that it is the sixth column of the matrix. Hence $e = (0 \ 0 \ 0 \ 0 \ 0 \ 1 \ 0)$

\Rightarrow Estimated codeword is $(0 \ 0 \ 1 \ 1 \ 1 \ 0 \ 0)$, which was the transmitted codeword. Hence decoding succeeds.

(c) Assume that first two coordinates are flipped. Then we get the syndrome $\underline{s} = \underline{h}\underline{y}^T$ as sum of first 2 columns = $\begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$ which is the 3rd column. Hence, decoder 'thinks' the error was in the third position. & flips the third position to get a codeword estimate, which will be wrong.

This happens for any fix codeword & any two bit flips as the sum of any two columns of a H matrix results in some other third column.

Some students may not write this, which may incur only 0.5 mark loss

$$C_{BSC} = \max_{P_X} I(X;Y).$$

3. BSC channel capacity. Assume $X \sim \text{Bern}(\alpha)$

(a)

$$I(X;Y) = H(Y) - H(Y|X)$$

$$H(Y|X=0) = H(Y|X=1) = H_2(P) \Rightarrow H(Y|X) = H_2(P)$$

$$P_Y(1) = (P_X(1)(1-p) + P_X(0)p) = \alpha(1-p) + (1-\alpha)p$$

$$P_Y(0) = (1-\alpha)(1-p) + \alpha p.$$

$$\Rightarrow H(Y) = -(\alpha(1-p) + (1-\alpha)p) \log(\alpha(1-p) + (1-\alpha)p) - ((1-\alpha)(1-p) + \alpha p) \log((1-\alpha)(1-p) + \alpha p)$$

$$= -(\alpha + p - 2p\alpha) \log(\alpha + p - 2p\alpha) \\ - (1 - \alpha - p + 2p\alpha) \log(1 - \alpha - p + 2p\alpha)$$

$$\frac{dH(Y)}{d\alpha} = - (1 - 2p) \log(\alpha + p - 2p\alpha) \\ - (-1 + 2p) \log(1 - \alpha - p + 2p\alpha)$$

Cancels

$$\left[\begin{array}{l} - \frac{(\alpha + p - 2p\alpha)}{(\alpha + p - 2p\alpha)} (1 - 2p) \log_e^2 \\ - \frac{(1 - \alpha - p + 2p\alpha)}{(1 - \alpha - p + 2p\alpha)} (-1 + 2p) \log_e^2 \end{array} \right]$$

The above is 0 when $\log(\alpha + p - 2p\alpha) = \log(1 - \alpha - p + 2p\alpha)$

$$(i.e.) \alpha + p - 2p\alpha = 1 - \alpha - p + 2p\alpha$$

$$\Rightarrow 2\alpha + 2p = 1 + 4p\alpha$$

Since this holds for every p , we must get
 $\alpha = 0.5$.

Plugging $\alpha = 0.5$ in $H(Y)$ gives us that $H(Y) = 1$.

We know that $H(Y) \leq 1$ (as Y is binary valued
 $\& \log_2(|Y|) = 1$)

Thus $\alpha = 0.5$ is the highest value.

$$\text{Hence } \max_{P_X} I(X; Y) = \max_{P_X} H(Y) - H(Y|X) \\ = 1 - H_2(\alpha).$$

(Procedure could be bit simpler when student uses $H(X) - H(X|Y)$)

(b) Prob of error when using repetition code: (For convenience we assume n is odd. Roughly idea works for n even)

$$L_{\text{rep}} : \left\{ \begin{array}{l} (0 \dots 0), \\ \text{all-zero} \end{array}, \begin{array}{l} (1, \dots, 1) \\ \text{all-one} \end{array} \right\} \subseteq \{0, 1\}^n.$$

Decoding rule: $\hat{x}_{\text{MAP}} = \underset{\underline{x} \in L_{\text{rep}}}{\operatorname{argmax}} p(\underline{x}|y)$ (Prob that \underline{x} is for x , given y is received)

$$= \underset{\underline{x} \in L_{\text{rep}}}{\operatorname{argmax}} \frac{p(y|\underline{x}) p(\underline{x})}{p(y)}$$

On $p(\underline{x}) = \frac{1}{2^n}$ for $\underline{x} \in L_{\text{rep}}$. Further $p(y)$ doesn't change.

$$\text{Hence } \hat{x}_{\text{MAP}} = \underset{\underline{x} \in L_{\text{rep}}}{\operatorname{argmax}} p(y|\underline{x})$$

Now suppose $\underline{x} = (0, \dots, 0)$. Then we make an error when $p(y|0 \dots 0) < p(y|1 \dots 1)$

Let $\omega(y)$ denote no. of 1s in y .

$$\text{Thus } P(y|0\cdots 0) = P^{\omega(y)} (1-p)^{n-\omega(y)}$$

$$\& P(y|1\cdots 1) = (1-p)^{\omega(y)} p^{n-\omega(y)}$$

Thus when $\underline{x} = (0\cdots 0)$ is true, we make an error when

$$P^{\omega(y)} (1-p)^{n-\omega(y)} < (1-p)^{\omega(y)} p^{n-\omega(y)}$$

$$\Rightarrow \text{when } \left(\frac{p}{1-p}\right)^{\omega(y)} \left(\frac{1-p}{p}\right)^{n-\omega(y)} < 1$$

$$\Rightarrow \left(\frac{1-p}{p}\right)^{n-2\omega(y)} < 1$$

$$\text{Now, as } p < 0.5, \frac{1-p}{p} > 1.$$

$$\text{Thus } \left(\frac{1-p}{p}\right)^{n-2\omega(y)} < 1 \text{ iff } 2\omega(y) > n.$$

(a) When more than $n/2$ flips happen.

$$\text{Thus } P(\text{error } 00\cdots 0 \text{ is true}) = P(\text{more than } \frac{n}{2} \text{ bits})$$

$$= \sum_{i=\frac{n}{2}+1}^n \binom{n}{i} p^i (1-p)^{n-i}$$

||| we can calculate $P(\text{error } 1\cdots 1 \text{ is true})$

$$= \sum_{i=\frac{n}{2}+1}^n \binom{n}{i} p^i (1-p)^{n-i}$$

The average of the two is $P(\text{error})$

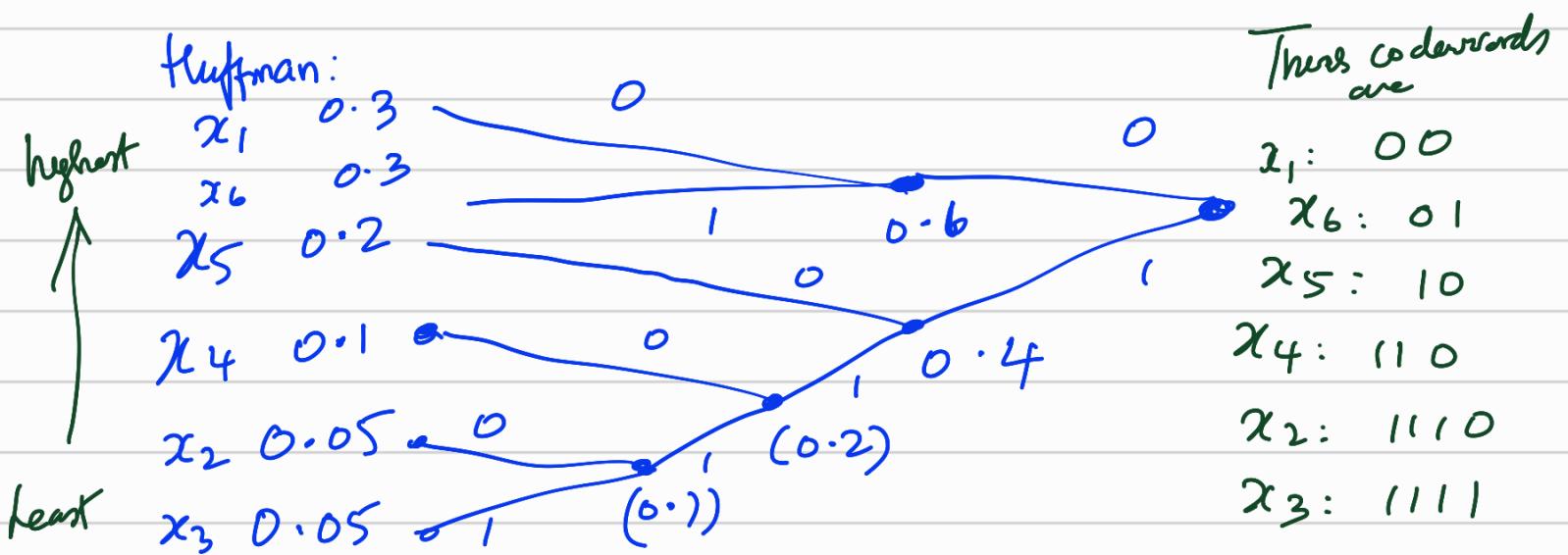
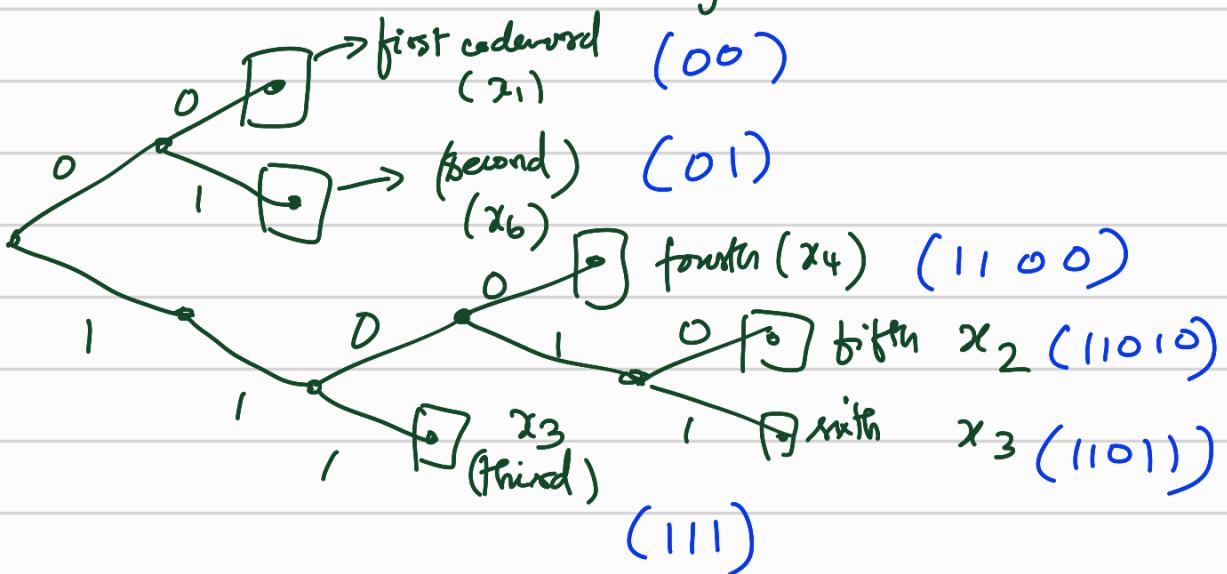
$$P(\text{error}) = P_x(0) P(\text{error} | 0) + P_x(1) P(\text{error} | 1)$$

$$= \sum_{i=n/2+1}^n \binom{n}{i} p^i (1-p)^{n-i}.$$

4. (a) Shannon - Fano technique: Choose lengths as follows

$-\log 0.3$	$-\log 0.05$	$-\log 0.05$	$-\log 0.17$	$-\log 0.2$
(2)	(5)	(5)	&	(2)
x_1	x_2	x_3		x_6

Thus a SF code is obtained as follows



(b)

Expected lengths are

$$\begin{aligned} L_{SF} &= 0.3 \times 4 + 0.05 \times 10 + 0.1 \times 4 + 0.2 \times 3 \\ &= 1.2 + 0.5 + 0.4 + 0.6 = 2.7 \text{ bits} \end{aligned}$$

$$\begin{aligned} L_{Huff} &= 0.8 \times 2 + 0.1 \times 3 + 0.1 \times 4 \\ &= 1.6 + 0.7 = 2.3 \text{ bits} \end{aligned}$$

Thus $L_{Huff} < L_{SF}$.

(c) Entropy is $2.271 \Rightarrow L_{Huff}$ is close to $H(x)$
 but L_{SF} is larger.

5. [Hard question as students have not seen].

Recall for Gaussian channel: $P(y|x) = \frac{1}{\sqrt{2\pi N_0/2}} e^{-\frac{(y-x)^2}{N_0}}$

MAP rule is as before $\hat{x}_{MAP} = \operatorname{argmax}_{x \in \{A, -A\}} P(x|y)$

$$\hat{x}_{MAP} = \operatorname{argmax}_{x \in \{A, -A\}} \frac{\dot{P}(y|x) p(x)}{p(y)}$$

$$= \operatorname{argmax}_{x \in \{A, -A\}} P(y|x)$$

$$x \in \{A, -A\}$$

When x is uniform.

$$\text{But } p(y|x) = \frac{1}{\sqrt{2\pi \frac{N_0}{2}}} e^{-\frac{(y-x)^2}{2\frac{N_0}{2}}}$$

$$= \frac{1}{\sqrt{\pi N_0}} e^{-\frac{(y-x)^2}{N_0}}$$

\Rightarrow MAP rule essentially because

$$\hat{x} = \underset{x \in \{A, -A\}}{\operatorname{argmax}} e^{-\frac{(y-x)^2}{N_0}}$$

$$= \underset{x \in \{A, -A\}}{\operatorname{argmin}} |y-x| \quad (\text{ } |y-x| \text{ is absolute diff between } x \text{ & } y)$$

Thus when trx symbol is A, we have an error when y is closer to -A than A.
 $\Rightarrow y \in [-\infty \text{ to } 0]$

When trx symbol is -A, we have an error when y is closer to A than -A. $\Rightarrow y \in [0 \text{ to } \infty]$

$$\Rightarrow P(\text{error}) = P(\text{error} | A \text{ is trx}) P(A \text{ is trx}) + P(\text{error} | -A \text{ is trx}) P(-A \text{ is trx})$$

$$P(\text{error} | A \text{ is trx}) = \int_0^\infty p(y|x=A) dy$$

$$y = -\infty$$

$$= \frac{1}{\sqrt{\pi N_0}} \int_{-\infty}^0 e^{-(y-A)^2/N_0} dy$$

$$P(\text{error} | -A \text{ is } \text{bx}) = \frac{1}{\sqrt{\pi N_0}} \int_0^\infty e^{-(y+A)^2/N_0} dy.$$

$$= \frac{1}{\sqrt{\pi N_0}} \int_{-\infty}^0 e^{-(y-A)^2/N_0} dy$$

\Rightarrow Total error prob is

$$P(\text{error}) = \frac{1}{\sqrt{\pi N_0}} \int_{-\infty}^0 e^{-(y-A)^2/N_0} dy.$$

6. Clearly we see that $q(x,y) \geq 0 \quad \forall x,y$.

$$\begin{aligned} \text{Now } \sum_{x,y} q(x,y) &= \sum_{x,y} P_x(x) P_y(y) \\ &= \left(\sum_x P_x(x) \right) \left(\sum_y P_y(y) \right) \\ &= 1 \times 1 = 1. \end{aligned}$$

Hence $q(x,y)$ is a valid distribution.

(Same holds in continuous case)

$$\text{Now } I(x; y) = \sum_{x, y \in \text{supp}(P_{x,y})} p(x, y) \log \frac{p(x, y)}{p_x(x) p_y(y)}$$

$$= -I(x; y)$$

$$= \sum_{(x, y) \in \text{supp}(P_{x,y})} p(x, y) \log \frac{p_x(x) p_y(y)}{P_{x,y}(x, y)}$$

As log is concave, we get that

$$\text{avg of log} \leq \log(\text{avg})$$

$$\Rightarrow -I(x; y) \leq \log \left[\sum_{(x, y) \in \text{supp}(P_{x,y})} p(x, y) \left(\frac{p_x(x) p_y(y)}{P_{x,y}(x, y)} \right) \right] \\ \leq \log \sum_{(x, y) \in \text{supp}(P_{x,y})} p_x(x) p_y(y)$$

$$\text{Now } \sum_{(x, y) \in \text{supp}(P_{x,y})} p_x(x) p_y(y) \leq 1$$

$$\Rightarrow -I(x; y) \leq 0 \Rightarrow I(x; y) \geq 0.$$

(If student has mixed $\text{supp}(P_{x,y})$ it is OK).

Further as $H(Y|X)$ & $H(X|Y)$ are nonnegative,
we have that (Simple proof is good if written,
so even otherwise is ok)

$$I(X;Y) = H(Y) - H(Y|X)$$

$$\leq H(Y)$$

$$\& I(X;Y) = H(X) - H(X|Y) \leq H(X)$$

$$\Rightarrow I(X;Y) \leq \min(H(X), H(Y))$$

7.

$$(i) X_1(f) = \int_{t=-\infty}^{\infty} x(3t-6) e^{-j2\pi f t} dt$$

$$\text{Let } t' = 3t - 6 \Rightarrow t = \left(\frac{t'+6}{3}\right)$$

$$\Rightarrow X_1(f) = \int_{t=-\infty}^{\infty} x(t') e^{-j2\pi f \left(\frac{t'+6}{3}\right)} dt'$$

$$= e^{-j4\pi f} \int_{t=-\infty}^{\infty} x(t') e^{-j2\pi f t'} f/3 dt'$$

$$X_1(f) = e^{-j4\pi f} X(f/3).$$

$$(ii) X_2(f) = \int_{t=-\infty}^{\infty} x(t) e^{j2\pi f t} e^{-j2\pi f t} dt$$

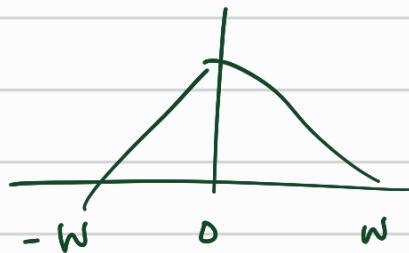
$$= \int_{t=-\infty}^{\infty} x(t) e^{-j2\pi(f-f_0)t} dt$$

$$= X(f - f_0)$$

Answers to other parts:

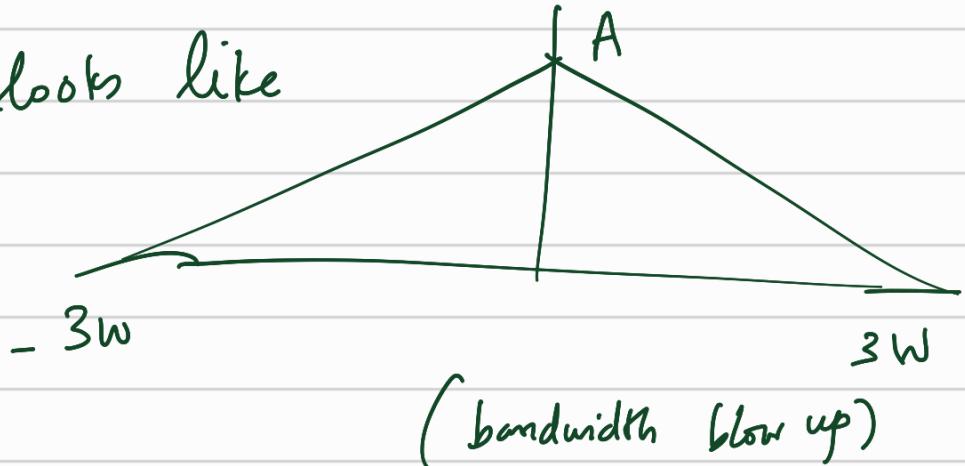
(a)

Assume $X(f)$



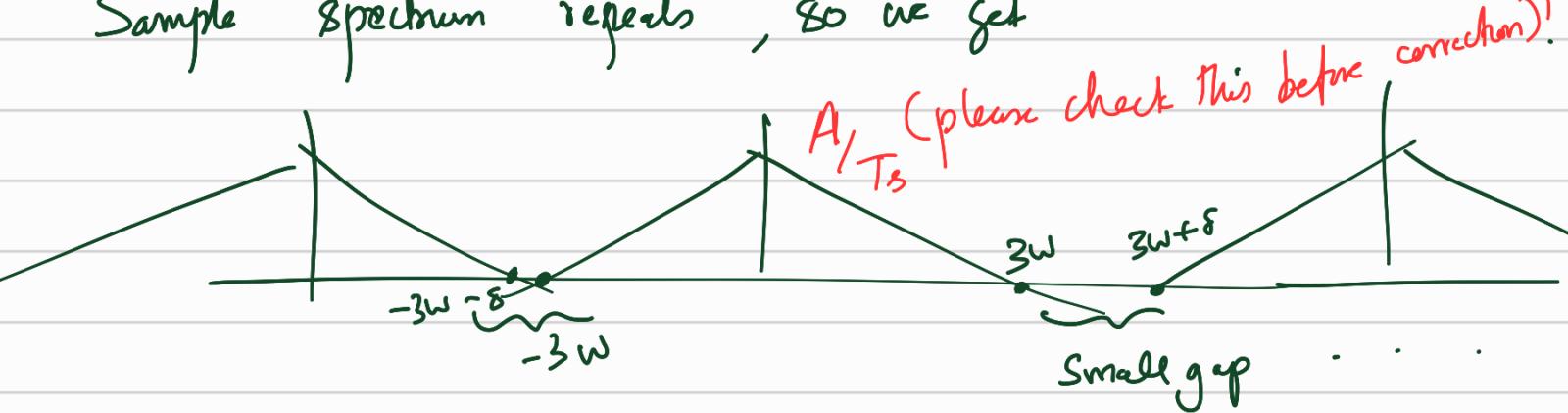
(if such general $X(f)$ is not assumed then cut 1 mark)

Then $|X_1(f)|$ looks like



(bandwidth blow up)

Sample spectrum repeats, so we get



- (b) $x_2(t)$ is closer to the idea of modulation as it looks like the spectrum has undergone freq shift, just like $x_2(t) \cos 2\pi f_0 t$.
- (c) If $x(t)$ is like a baseband signal, we take only five part of the spectrum for bandwidth. This is given as W
- (i) $x_1(t)$ has BW of $3W$ as there is no overlap
 - (ii) $x_2(t)$ has BW $2W$ as fo is given as large enough & $x_2(t)$ is more like passband signal

8. (a) In given question, $g(x, y) = x + y$.

Thus $\mathbb{E}[x+y] = \sum_{x,y} p_{x,y}(x,y)(x+y)$

$$= \sum_{x,y} p_{x,y}(x,y)x + \sum_{x,y} p_{x,y}(x,y)y$$

$$= \sum_x \left(\sum_y p_{x,y}(x,y) \right) x + \sum_y \left(\sum_x p_{x,y}(x,y) \right) y$$

$$= \sum_x p_x(x) x + \sum_y p_y(y) y$$

$$\mathbb{E}[x+y] = \mathbb{E}[x] + \mathbb{E}[y].$$

(b) Given n RVs x_1, \dots, x_n with joint prob mass fn p_{x_1, \dots, x_n} . Let $g(x_1, \dots, x_n)$ be any fn of the n RVs. Then the

expectation

$$\mathbb{E}[g(x_1, \dots, x_n)] \triangleq \sum_{(x_1, \dots, x_n)} p_{x_1, \dots, x_n}(x_1, \dots, x_n) g(x_1, \dots, x_n)$$

(defined as).

Then show that

$$\mathbb{E}[x_1 + \dots + x_n] = \mathbb{E}[x_1] + \dots + \mathbb{E}[x_n]$$