

Q1

a) The entropy is

$$\begin{aligned}
 H(x) &= - \sum_i p_i \log_2 p_i \\
 &= - (0.12 \log_2 (0.12) \\
 &\quad + 0.08 \log_2 (0.08) \\
 &\quad + 0.2 \log_2 (0.2) \\
 &\quad + 0.35 \log_2 (0.35) \\
 &\quad + 0.2 \log_2 (0.2) \\
 &\quad + 0.05 \log_2 (0.05))
 \end{aligned}$$

$$\underline{\underline{= 2.33 \text{ bits}}}$$

b) The average codeword length is lower bounded by the entropy of the source for error free reconstruction. So, the minimum possible average codeword length is $\underline{\underline{H(x) = 2.33 \text{ bits}}}$.

c) For a uniformly distributed source with the same alphabet, each symbol would have equal probability. i.e. $\frac{1}{6}$

So, the entropy of the uniformly distributed source is:

$$H(x) = - \sum_{i=1}^6 \frac{1}{6} \log_2 \left(\frac{1}{6} \right)$$

$$= - 6 \times \frac{1}{6} \log_2 \left(\frac{1}{6} \right)$$

$$= \log_2 (6)$$

$$= \underline{\underline{2.585 \text{ bits}}}$$

Comparing this with the entropy of the original source, we can see that the original source has lower entropy, indicating less uncertainty.

d) To design the Huffman code, we start by sorting source outputs in decreasing order of their probabilities.

x_i	code	$P(x_i)$
x_1	00	0.35
x_2	10	0.20
x_3	11	0.20
x_4	010	0.12
x_5	0110	0.08
x_6	0111	0.05

The average codeword length is

$$L_{avg} = \sum_{i=1}^6 P(x_i) \times \text{length}(x_i)$$

$$\begin{aligned}
 &= 0.35 \times 2 \\
 &+ 0.20 \times 2 \\
 &+ 0.20 \times 2 \\
 &+ 0.12 \times 3 \\
 &+ 0.08 \times 4 \\
 &+ 0.05 \times 4
 \end{aligned}$$

$$= \underline{\underline{2.38 \text{ bits/source output}}}$$

It is clear that the average codeword length is very close to the source entropy.

Q2.

a) The entropy is:

$$H(x) = - \sum_i p_i \log_2 p_i$$

$$= - (0.2 \log_2 (0.2))$$

$$+ 0.4 \log_2 (0.4)$$

$$+ 0.4 \log_2 (0.4))$$

$$= \underline{1.52 \text{ bits}}$$

b) To design the Huffman code,

we start by sorting source outputs in decreasing order of their probabilities.

x_i	code	$P(x_i)$
a_3	0	0.4
a_2	10	0.4
a_1	11	0.2

The average codeword length is

$$L_{avg} = \sum_{i=1}^3 P(x_i) \times \text{length}(x_i)$$

$$= 0.4 \times 1 + 0.4 \times 2 + 0.2 \times 2$$

$$= \underline{1.6 \text{ bits / source output}}$$

c)

For the second extension of the source (taking 2 letters at a time), we first calculate the probabilities of each pair.

$$P(a_1 a_1) = 0.2 \times 0.2 = 0.04$$

$$P(a_1 a_2) = 0.2 \times 0.4 = 0.08$$

$$P(a_1 a_3) = 0.2 \times 0.4 = 0.08$$

$$P(a_2 a_1) = 0.4 \times 0.2 = 0.08$$

$$P(a_2 a_2) = 0.4 \times 0.4 = 0.16$$

$$P(a_2 a_3) = 0.4 \times 0.4 = 0.16$$

$$P(a_3 a_1) = 0.4 \times 0.2 = 0.08$$

$$P(a_3 a_2) = 0.4 \times 0.4 = 0.16$$

$$P(a_3 a_3) = 0.4 \times 0.4 = 0.16$$

∴ New alphabet is

$$\{a_1 a_1, a_1 a_2, a_1 a_3, a_2 a_1, a_2 a_2, a_2 a_3, a_3 a_1, a_3 a_2, a_3 a_3\}$$

∴ Corresponding probabilities are

$$\{0.04, 0.08, 0.08, 0.08, 0.16, 0.16, 0.16, 0.16\}$$

To design the Huffman code,
we start by sorting source outputs in decreasing
order of their probabilities.

x_i	Code	$P(x_i)$
$a_3 a_3$	000	0.16 — 0
$a_3 a_2$	001	0.16 — 0.32 0
$a_2 a_3$	010	0.16 — 0.32 0.64 0
$a_2 a_2$	011	0.16 — 0.32 0.64 0 1
$a_3 a_1$	0100	0.08 — 0.08 0.20 0
$a_2 a_1$	0110	0.08 — 0.16 0.36 1
$a_1 a_3$	111	0.08 —
$a_1 a_2$	1010	0.08 — 0.12
$a_1 a_1$	1011	0.04 —

The average codeword length is

$$\begin{aligned}
 L_{avg} &= \sum_{i=1}^9 P(x_i) \times \text{length}(x_i) \\
 &= 0.16 \times 3 + 0.16 \times 3 + 0.16 \times 3 \\
 &\quad + 0.16 \times 3 + 0.08 \times 3 + 0.08 \times 3 \\
 &\quad + 0.08 \times 3 + 0.08 \times 4 + 0.04 \times 4 \\
 &= \underline{\underline{3.12 \text{ bits / source output pair}}}
 \end{aligned}$$

The average number of required binary letters per each source output letter is $\frac{L_{avg}}{2}$

$$= \frac{3.12}{2} = \underline{\underline{1.56 \text{ bits / source output}}}$$

d)

Huffman coding of the original source requires 1.6 bits / source output letter.

$$\therefore \text{Efficiency (in %)} = \frac{\text{Entropy (H(x))}}{L_{avg}} = \frac{1.52}{1.56} \times 100 = 95\%$$

Huffman coding of second extension source requires 1.56 bits / source output letter.

$$\therefore \text{Efficiency} = \frac{1.52}{1.56} \times 100 = 97.4\%$$

\therefore Huffman coding of the second extension of the source is more efficient coding scheme.

Q3.

To design the Huffman code,
we start by sorting source outputs in decreasing
order of their probabilities.

x_i	code	$P(x_i)$	
x_1	A	0.23	A
x_2	BA	0.20	A
x_3	BB	0.18	B 0.54 B
x_4	BC	0.16	C
x_5	CA	0.13	A
x_6	CB	0.05	B 0.23 C
x_7	CC	0.05	C

The average codeword length is

$$\begin{aligned}
 L_{avg} &= \sum_{i=1}^7 P(x_i) \times \text{length}(x_i) \\
 &= 0.23 \times 1 + 0.20 \times 2 + 0.18 \times 2 \\
 &\quad + 0.16 \times 2 + 0.13 \times 2 + 0.05 \times 2 \\
 &\quad + 0.05 \times 2 = 1.77 \\
 &= \underline{\underline{1.77 \text{ ternary symbol / source output}}}
 \end{aligned}$$

For a fair comparison of the average codeword length with the entropy of the source, we should compute the entropy of ternary code with logarithms in base 3. So,

$$H(x) = - \sum_i P_i \log_3 P_i$$

$$= - (0.05 \log_3 (0.05))$$

$$+ 0.2 \log_3 (0.2)$$

$$+ 0.05 \log_3 (0.05)$$

$$+ 0.16 \log_3 (0.16)$$

$$+ 0.18 \log_3 (0.18)$$

$$+ 0.23 \log_3 (0.23)$$

$$+ 0.13 \log_3 (0.13))$$

$$= \underline{1.66 \text{ ternary symbols / source symbol}}$$

$$\text{Average code word (L}_{\text{avg}}\text{)} = 1.77$$

$$\text{Entropy (H}(x)\text{)} = 1.66$$

As expected, $L_{\text{avg}} \geq H(x)$

Q4.

Encoding (Lempel-Ziv compression)

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Locations	Content	Codeword
1	B	0 B
2	A	0 A
3	AA	2 A
4	AB	2 B
5	AAA	3 A
6	ABA	4 A
7	BA	1 A
8	BB	1 B
9	B BB	8 B
10	B BA	8 A
11	ABAB	6 B
12	A AB	3 B
13	A BABB	11 B
14	A BABBA	13 A
15	A BB	4 B
16	AA BB	12 B
17	ABB	15

Decoding (Recovering the original sequence)

Location	Codeword	Content
1	0 8	B
2	0 A	A
3	2 A	AA
4	2 B	AB
5	3 A	AAA
6	4 A	A B A
7	1 A	B A
8	1 B	BB
9	BB	BBB
10	8 A	B B A
11	6 B	A B A B
12	3 B	A A B
13	11 B	A B A B B
14	13 A	A B A B B A
15	4 B	A B B
16	12 B	A A B B
17	15	A B B

Q5.

Given:

$$\begin{aligned}
 & P(x=0, y=0) \\
 & = P(x=0, y=1) \\
 & = P(x=1, y=1) \\
 & = \frac{1}{5}
 \end{aligned}$$

and $\sum_i \sum_j P(x_i, y_j) = 1$,
we get that

$$P(x=1, y=0) = 1 - \frac{3}{5} = \frac{2}{5}$$

The marginal probabilities are

$$P(x=0) = P(x=0, y=0) + P(x=0, y=1) = \frac{2}{5}$$

$$P(x=1) = P(x=1, y=0) + P(x=1, y=1) = \frac{3}{5}$$

$$P(y=0) = P(x=0, y=0) + P(x=1, y=0) = \frac{3}{5}$$

$$P(y=1) = P(x=0, y=1) + P(x=1, y=1) = \frac{2}{5}$$

$$g) H(x) = - \sum_{i=1}^2 p_i \log_2 p_i$$

$$= - \left(\frac{2}{5} \log_2 \left(\frac{2}{5} \right) + \frac{3}{5} \log_2 \left(\frac{3}{5} \right) \right)$$

$$\therefore = 0.971 \text{ bits / source output}$$

b) $H(Y) = - \sum_{i=1}^n p_i \log_2 p_i$

$$= - \left(\frac{3}{5} \log_2 \left(\frac{3}{5} \right) + \frac{2}{5} \log_2 \left(\frac{2}{5} \right) \right)$$

$$= \underline{0.971 \text{ bits / source output.}}$$

c) $H(X, Y)$

The joint entropy

$$H(X, Y) = - \sum_{i=1}^n \sum_{j=1}^m p(x_i, y_j) \log_2 p(x_i, y_j)$$

$$= - \left(\frac{1}{5} \log_2 \left(\frac{1}{5} \right) + \frac{1}{5} \log_2 \left(\frac{1}{5} \right) \right.$$

$$\left. + \frac{1}{5} \log_2 \left(\frac{1}{5} \right) + \frac{2}{5} \log_2 \left(\frac{2}{5} \right) \right)$$

$$= \underline{1.922 \text{ bits / source output.}}$$

d) The conditional entropy is

$$H(X|Y) = H(X, Y) - H(Y)$$

$$= 1.922 - 0.971$$

$$= \underline{0.951 \text{ bits / source output}}$$

$$H(Y|X) = H(X, Y) - H(X)$$

$$= 1.922 - 0.971$$

$$= \underline{0.951 \text{ bits / source output}}$$

c) The mutual information is

$$\begin{aligned} I(x; y) &= H(x) - H(x|y) \\ &= 0.971 - 0.951 \\ &= \underline{\underline{0.02 \text{ bits/source output}}}. \end{aligned}$$