

Q1 Requiring one bit above entropy

Monday, August 29, 2011 10:20 AM

We want to come up with some simple example. Therefore we shall start by considering the Bernoulli RV X which has only two possible values:

Consider $X \sim \text{Bernoulli}(p_1)$: $p(x) = \begin{cases} p_1, & x=1, \\ 1-p_1, & x=0, \\ 0, & \text{otherwise.} \end{cases}$

Because there are only two possible values, the pairing in the Huffman coding process must be between these two values:

x	$p(x)$		$c(x)$	$l(x)$
0	$1-p_1$	0	0	1
1	p_1	1	1	1

Therefore, Huffman coding (without extension) always needs 1 bit per symbol.

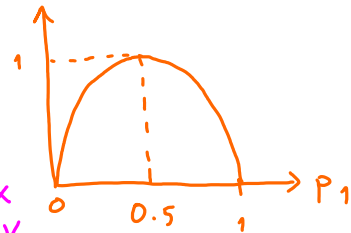
Now, we want the expected length to be $\approx H(X)+1$
 $= 1$

So, we must have $H(X) \approx 0$.

By definition, $H(X) = -\sum_x p(x) \log_2 p(x) = -(1-p_1) \log_2(1-p_1) - p_1 \log_2 p_1$

You may recall that in Section 2.4, this expression is called the binary entropy function. The plot of this function is shown in the lecture notes; it is sketched below:

Note that to get this function to be ≈ 0 , we need to consider $p_1 \approx 0$ or $p_1 \approx 1$.



Also note that we don't want to have $p_1 = 0$ or $p_1 = 1$ because they would make our RV X degenerated (deterministic). For degenerated RV, we don't have to waste any bit to convey its value. (The value is already pre-determined.) For example, if $p_1 = 1$, we know that $P[X=1]=1$ and hence $X \equiv 1$ all the time. There is no uncertainty. Anyone can guess the value of X with 100% accuracy by simply guessing the value 1 every time. Alternatively, we may think of the situation here as sending the empty string (ϵ) to the receiver. So, $\mathbb{E}[l(X)] = \mathbb{E}[0] = 0$. Because $H(X)$ is also 0, we have $\mathbb{E}[l(X)] = H(X)$ and not $H(X)+1$.

More formally we can take $\lim_{p_1 \rightarrow 0}$ or $\lim_{p_1 \rightarrow 1}$ on the function and show that $H(X) \rightarrow 0$

$$\left(\lim_{\alpha \rightarrow 0^+} \alpha \ln \alpha = \lim_{\alpha \rightarrow 0^+} \frac{1}{\frac{1}{\alpha}} = \lim_{\alpha \rightarrow 0^+} (-\alpha) = 0 \right)$$

$$\left(\lim_{\alpha \rightarrow 0^+} \alpha \ln \alpha = \lim_{\alpha \rightarrow 0^+} \frac{\frac{1}{\alpha}}{-\frac{1}{\alpha^2}} = \lim_{\alpha \rightarrow 0^+} (-\alpha) = 0 \right)$$

↑
L'Hôpital's rule

So, if p_i is close to 1 or 0, the entropy will be almost 0.

But we still need $\mathbb{E}[\ell(x)] = 1$ bit to send x .

So, the expected length will be very close to $H(x) + 1$.

Q2 ML and MAP detectors for repetition code

Tuesday, September 20, 2011 4:36 PM

Recall that the code rate of a repetition code is $\frac{1}{n}$ where n is the number of times that an info-bit is repeated.

Here, the code rate is $\frac{1}{5}$. So, $n=5$.

Codebook :

\underline{x}	\underline{z}
0	00000
1	11111

Usually, for block coding, this part is also a row vector.

This is because we work with k info-bit at a time.

In which case, \underline{z} would be a row vector of length k .

However, for repetition code, we have $k=1$.

(a) In class, we have shown that for repetition code over BSC with $p < 0.5$, the ML decoder is the same as a majority-vote decoder:

$$\hat{s}_{ML}(\underline{y}) = \begin{cases} 0 & \text{if } n_0 > n_1 \\ 1 & \text{if } n_0 < n_1 \end{cases}$$

$n_1 =$ the number of 1's in the observed vector \underline{y}
 $n_0 =$ the number of 0's in the observed vector \underline{y}

Therefore,

n_0	n_1	$\hat{s}_{ML}(\underline{y})$
0	5	1
1	4	1
2	3	1
3	2	0
4	1	0
5	0	0

more than half

$P(\epsilon) =$ the probability that the ≥ 3 bits flipped by the BSC

$$= \sum_{k=3}^5 (\text{the probability that the } \geq 3 \text{ bits flipped by the BSC} = k)$$

$$= \binom{5}{3} p^3 (1-p)^2 + \binom{5}{4} p^4 (1-p) + \binom{5}{5} p^5 (1-p)^0$$

$$= 10 p^3 (1-p)^2 + 5 p^4 (1-p) + p^5$$

$$= 0.317$$

$p = 0.4$

(b) In class, we have shown that the MAP detector for block coding over the BSC (with crossover probability p) is given by

$$\hat{x}_{\text{MAP}}(\underline{y}) = \arg \max_{\underline{x}} \left(\frac{p}{1-p} \right)^{d(\underline{x}, \underline{y})} P(\underline{x}).$$

We have also shown that when the block coding is done simply by using repetition code,

$$\hat{x}_{\text{MAP}}(\underline{y}) = \begin{cases} \underline{0} & \text{when } \left(\frac{p}{1-p} \right)^{n_1 - n_0} \frac{p_0}{p_1} > 1 \\ \underline{1} & \text{when } \left(\frac{p}{1-p} \right)^{n_1 - n_0} \frac{p_0}{p_1} < 1 \end{cases}$$

$p = 0.4$

$\frac{p}{1-p} = \frac{0.4}{0.6} = \frac{2}{3}$ $\frac{p_0}{p_1} = \frac{0.4}{0.6} = \frac{2}{3}$

n_1	n_0	$n_1 - n_0$	$\left(\frac{p}{1-p} \right)^{n_1 - n_0}$	$\times \frac{p_0}{p_1}$	$\hat{s}_{\text{MAP}}(\underline{y})$
0	5	-5	$(2/3)^{-5}$	> 1	0
1	4	-3	$(2/3)^{-3}$	> 1	0
2	3	-1	$(2/3)^{-1}$	$= 1$	Any
3	2	1	$(2/3)$	< 1	1
4	1	3	$(2/3)^3$	< 1	1
5	0	5	$(2/3)^5$	< 1	1

so, can set $\hat{s}_{\text{MAP}}(\underline{y}) = \hat{s}_{\text{ML}}(\underline{y})$ above.

In which case, same probability of error.

If this is set to be 0, then it is the same as \hat{s}_{ML} .

Remark ① Let's try to set this to be 1 so that the detector is different from \hat{s}_{ML} .

$$\text{Then, } P[\hat{s} \neq s | s=0] = P[\hat{s} \neq 0 | s=0] = \sum_{k=2}^5 p^k (1-p)^{5-k} = 0.663$$

$$P[\hat{s} \neq s | s=1] = P[\hat{s} \neq 1 | s=1] = \sum_{k=4}^5 p^k (1-p)^{5-k} = 0.087$$

$$\text{so, } P[\varepsilon] = P[\hat{s} \neq s | s=0] P[s=0] + P[\hat{s} \neq s | s=1] P[s=1]$$

$$\approx 0.663 \times 0.4 + 0.087 \times 0.6 = 0.317 \rightarrow \text{same performance!}$$

Remark ② The "any" or "don't care" or "does not matter" in the table means that the decoder can choose to guess 1 or 0 without changing the overall error-rate performance. However, some decoder may issue a "decoding error" warning at the receiver and avoid making a guess. (We will not discuss this type of decoder.)

(c)

$$(i) P[S=0 | Y=01001] = \frac{P[Y=01001 | S=0] P[S=0]}{P[Y=01001]} = \frac{p^2(1-p)^3 \times p_0}{P[Y=01001]} \quad \leftarrow P[S=0] = 0.45$$

conditioned on $s=0$, the repetition code would be 00000 and therefore, $Y=01001$ means the BSC flips exactly two bits.

By the total probability theorem,

$$P[Y=01001] = P[Y=01001 | S=0] P[S=0] + P[Y=01001 | S=1] P[S=1]$$

$$= p^2(1-p)^3 p_0 + p^3(1-p)^2 p_1 \approx 0.0282$$

$$P[S=1] = 0.55$$

conditioned on $s=1$, the repetition code would be 11111 and therefore $Y=01001$ means the BSC flips exactly 3 bits.

$$\text{Therefore, } P[S=0 | Y=01001] \approx 0.5510$$

$$(ii) P[S=1 | Y=01001] = 1 - P[S=0 | Y=01001] = 0.4490$$

$$(iii) P[S=0 | Y=01001] > P[S=1 | Y=01001].$$

Therefore, it is more likely that $s=0$ was transmitted.

$$(d) (i) P[S=0 | Y=01001] = \frac{p^2(1-p)^3 p_0}{p^2(1-p)^3 p_0 + p^3(1-p)^2 p_1} = \frac{1}{1 + \frac{p}{1-p} \frac{p_1}{p_0}}$$

$$= \frac{1}{1 + \frac{2}{3} \times \frac{1-p_0}{p_0}} = \frac{3p_0}{p_0 + 2}$$

$$(ii) P[S=1 | \underline{Y} = 01001] = 1 - \frac{1 + \frac{x}{3} \frac{1}{p_0}}{1 + \frac{x}{3} \frac{1}{p_0}} = \frac{2 - 2p_0}{p_0 + 2}$$

(iii) $P[S=0 | \underline{Y} = 01001]$ is more likely iff

$$\frac{3p_0}{p_0 + 2} > \frac{2 - 2p_0}{p_0 + 2}$$

$$p_0 > \frac{2}{5} = 0.4$$

$P[S=1 | \underline{Y} = 01001]$ is more likely iff $p_0 < 0.4$

When $p_0 = 0.4$, the conditional probabilities are the same and the two cases are equally likely.

Majority voting would always guess 0 because the ~~x~~0s in 01001 is greater than the ~~x~~1s. This would agree with our answer here when $p_0 > 0.4$.

Q3 ML and MAP detectors for block coding

Tuesday, October 09, 2012 9:50 AM

(a) We can find the most likely transmitted codeword by using the MAP decoder

$$\hat{\underline{x}}_{\text{MAP}}(\underline{y}) = \arg \max_{\underline{x}} P[\underline{x} = \underline{x} | \underline{y} = \underline{y}]$$

↖ 01001

$$= \arg \max_{\underline{x}} Q(\underline{y} | \underline{x}) p(\underline{x})$$

Because all the codewords are equally likely. The term $p(\underline{x})$ in the above expression does not depend on \underline{x} and hence it can be eliminated. In other words, when the codewords are equally likely,

$$\hat{\underline{x}}_{\text{MAP}}(\underline{y}) = \hat{\underline{x}}_{\text{ML}}(\underline{y}) = \arg \max_{\underline{x}} Q(\underline{y} | \underline{x}).$$

In class, we have shown that when the crossover probability p of the BSC is < 0.5 , the ML decoder is the same as the minimum distance decoder.

So, we will use minimum-distance decoder here:

$$\hat{\underline{x}}(\underline{y}) = \arg \min_{\underline{x}} d(\underline{x}, \underline{y})$$

\underline{x}	$d(\underline{x}, \underline{y})$
00000	2
01000	1 ← minimum
10001	2
11111	3

↖ 01001

From the table, we see that the most likely transmitted codeword is **01000**.

(b)

	\underline{x}			
	00000	01000	10001	11111
00000		1	2	5
01000			3	4
10001				3
11111				

↖ $d_{\text{min}} = 1$

(c) For transmission of block codes over BSC, we have

$$\hat{x}_{\text{MAP}}(y) = \arg \max_{\underline{x}} \left(\frac{p}{1-p} \right)^{d(\underline{x}, y)} p(\underline{x})$$

Here, we have $p = 0.1$. So, $\frac{p}{1-p} = \frac{0.1}{0.9} = \frac{1}{9}$

\underline{x}	$d(\underline{x}, y)$	$p(\underline{x})$	$\left(\frac{p}{1-p} \right)^{d(\underline{x}, y)} p(\underline{x})$
00000	2	0.1	0.0012
01000	1	0.1	0.0111 ← maximum
10001	2	0.1	0.0012
11111	3	0.7	0.0010

From the table, we see that the most likely transmitted codeword is still **01000**.

Q4: H and I

Wednesday, October 02, 2013 10:19 AM

First, we need to find the unknown constant c .

The given description for the joint pmf can be expressed using the joint pmf matrix as

$$P = \begin{matrix} & \begin{matrix} y \\ x \end{matrix} & \begin{matrix} 2 & 4 \end{matrix} \\ \begin{matrix} 1 \\ 3 \end{matrix} & \begin{bmatrix} 3c & 5c \\ 5c & 7c \end{bmatrix} \end{matrix}$$

Recall that, for any joint pmf, we must have $\sum_x \sum_y P(x,y) = 1$.

Therefore, the sum of all elements in the P matrix should be 1:

Here, we have $3c + 5c + 5c + 7c = 1 \Rightarrow 20c = 1 \Rightarrow c = \frac{1}{20}$.

a) $H(X,Y) = H\left(\begin{bmatrix} \frac{3}{20} & \frac{1}{4} & \frac{1}{4} & \frac{7}{20} \end{bmatrix}\right) = -\frac{3}{20} \log_2 \frac{3}{20} - \frac{2}{4} \log_2 \frac{1}{4} - \frac{7}{20} \log_2 \frac{7}{20}$
 ≈ 1.9406 bits.

To find $H(X)$ and $H(Y)$, we need the marginal pmfs $p(x)$ and $q(y)$, respectively.

These can be found from the sums along the rows and columns of the P matrix.

$$\begin{matrix} & \begin{matrix} y \\ x \end{matrix} & \begin{matrix} 2 & 4 \end{matrix} \\ \begin{matrix} 1 \\ 3 \end{matrix} & \begin{bmatrix} 3c & 5c \\ 5c & 7c \end{bmatrix} \end{matrix} \begin{matrix} \rightarrow 8c = \frac{8}{20} = \frac{2}{5} \\ \rightarrow 12c = \frac{12}{20} = \frac{3}{5} \end{matrix}$$

$\downarrow \qquad \qquad \downarrow$

$$\begin{matrix} & & \begin{matrix} 8c & 12c \\ \parallel & \parallel \\ q(y) & 2/5 & 3/5 \end{matrix} \end{matrix}$$

b) $H(X) = H\left(\begin{bmatrix} \frac{2}{5} & \frac{3}{5} \end{bmatrix}\right) \approx 0.9710$

c) $H(Y) = H\left(\begin{bmatrix} \frac{2}{5} & \frac{3}{5} \end{bmatrix}\right) \approx 0.9710$

d) $H(X|Y) = H(X,Y) - H(Y) \approx 0.9697$

e) $H(Y|X) = H(X,Y) - H(X) \approx 0.9697$

f) $I(X;Y) = H(X) + H(Y) - H(X,Y) \approx 0.0013$

Q5: H and I (when X and Y are independent)

Wednesday, October 02, 2013 10:19 AM

First, we need to find the unknown constant β

The given description for the joint pmf can be expressed using the joint pmf matrix P as

$$P = \begin{array}{c|cc} & y & \\ \hline x & 1 & 3 \\ \hline 3 & \frac{1}{15} & \frac{4}{15} \\ 4 & \frac{2}{15} & \beta \end{array}$$

Recall that, for any joint pmf, we must have $\sum_x \sum_y \overbrace{p(x,y)}^{P[X=x, Y=y]} = 1$.

Therefore, the sum of all elements in the P matrix should be 1:

$$\text{Here, we have } \frac{1}{15} + \frac{4}{15} + \frac{2}{15} + \beta = 1 \Rightarrow \beta = 1 - \frac{7}{15} = \frac{8}{15}$$

(a) To check independence, we need the marginal pmfs $p(x)$ and $q(y)$, respectively.

These can be found from the sums along the rows and columns of the P matrix.

$$P = \begin{array}{c|cc} & y & \\ \hline x & 1 & 3 \\ \hline 3 & \frac{1}{15} & \frac{4}{15} \\ 4 & \frac{2}{15} & \frac{8}{15} \end{array} \begin{array}{l} \rightarrow \frac{5}{15} = \frac{1}{3} \\ \rightarrow \frac{10}{15} = \frac{2}{3} \end{array} \left. \vphantom{P} \right\} \Rightarrow p = \left[\frac{1}{3} \quad \frac{2}{3} \right]$$

$$\begin{array}{c} \downarrow \\ \frac{3}{15} \\ \downarrow \\ \frac{12}{15} \end{array} \left. \vphantom{P} \right\} \Rightarrow q = \left[\frac{1}{5} \quad \frac{4}{5} \right]$$

Recall that two random variables X and Y are independent if and only if

$$p(x,y) = p(x)q(y) \text{ for all pair } (x,y)$$

This is equivalent to $P = p^T q$.

Notice that in this problem, $P = p^T q$. Therefore, $X \perp\!\!\!\perp Y$.

(i) $H(X,Y) \stackrel{X \perp\!\!\!\perp Y}{=} H(X) + H(Y) \approx 1.6402$

(ii) $H(X) \stackrel{X \perp\!\!\!\perp Y}{=} H\left(\left[\frac{1}{3} \quad \frac{2}{3}\right]\right) \approx 0.9183$

(iii) $H(Y) \stackrel{X \perp\!\!\!\perp Y}{=} H\left(\left[\frac{1}{5} \quad \frac{4}{5}\right]\right) \approx 0.7219$

(iv) $H(X|Y) \stackrel{X \perp\!\!\!\perp Y}{=} H(X) \approx 0.9183$

(v) $H(Y|X) \stackrel{X \perp\!\!\!\perp Y}{=} H(Y) \approx 0.7219$

(vi) $I(X;Y) = 0$ because $X \perp\!\!\!\perp Y$