

4.2

$$\begin{aligned}
 I(x,y) &= \sum_x \sum_y p(x,y) \log \frac{p(x,y)}{p(x)p(y)} \quad \text{break things apart} \\
 &= \sum_x \sum_y \underbrace{p(x,y) \cdot \log(p(x,y))}_{-\sum_y p(x,y) \cdot \log p(y)} - \sum_x p(x,y) \cdot \log p(x) \\
 &\quad \text{drop the summation with no terms present}
 \end{aligned}$$

$\frac{4.9}{H(x,y)} =$
 $-\sum_{x,y} p(x,y) \cdot \log p(x,y)$ so signs flip

by the def. of conditional entropy

$$H(y|x) = H(x,y) - H(x) \Rightarrow H(x,y) = H(y|x) + H(x)$$

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can be plugged into above

4.3 a. Using majority voting: $\epsilon = \text{prob. of single channel error}$

$$P_{\text{error}} = P(\text{at least 2 wrong bits})$$

$$= 3 \cdot P(2 \text{ wrong}) + P(3 \text{ wrong})$$

$\underbrace{\text{Binomial dist}}$

$$= 3(1-\epsilon)\epsilon^2 + \epsilon^3$$

3 is from choose 2

$$= 3\epsilon^2 - 3\epsilon^3 + \epsilon^3$$

$(1-\epsilon)$: prob of getting 1 right

$$= 3\epsilon^2 - 2\epsilon^3$$

$$\epsilon^2 = P(2 \text{ wrong})$$

b. $P_{\text{error}} = 3\epsilon_1^2 - 2\epsilon_1^3$ Negt the probs. for error

$$\text{where } \epsilon_1 = 3\epsilon^2 - 2\epsilon^3$$

$$= 3(3\epsilon^2 - 2\epsilon)^2 - 2(3\epsilon^2 - 2\epsilon)^3$$

c. each round of voting triples # sent
so n rounds requiring 3^n bits

so we have

$$P(1 \text{ right}) \cdot P(2 \text{ wrong})$$

$$\cdot 3$$

↑
can happen 3 ways.

Not sure how to write this recursion naturally
but when plotted

4.4

$$\text{Diff. Entropy} = \int_{-\infty}^{\infty} p(x) \log p(x) dx$$

where for a gaussian process $p(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$

$$\text{diff entropy} = - \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \cdot \log \left[\frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \right] dx$$

where $\mu = \text{mean}$
 $\sigma^2 = \text{variance}$

substitute $x = x - \mu^2$

$$\therefore - \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \cdot \left[-\frac{1}{2} \log(2\pi\sigma^2) - \frac{x^2}{2\sigma^2} \right] dx$$

choose nat. log to make it easier.

THIS is $p(x)$! integrated from $-\infty$ to ∞ is $= 1$

$$= \frac{1}{2} \log(2\pi\sigma^2) \cdot \int_{-\infty}^{\infty} \left(\frac{1}{\sqrt{2\pi\sigma^2}} \right) \left(e^{-\frac{(x-\mu)^2}{2\sigma^2}} \right) dx + \int_{-\infty}^{\infty} \left(\frac{1}{\sqrt{2\pi\sigma^2}} \right) \left(e^{-\frac{(x-\mu)^2}{2\sigma^2}} \right) \left(\frac{x^2}{2\sigma^2} \right) dx$$

can't do the same trick

4.5

a. Telephone line $\Delta f = 3300 \text{ Hz}$

$$\text{SNR} = 20 \text{ dB}$$

what is capacity (in bits / second)

Shannon-

Hartley

Theorem

$$= \Delta f \log_2 \left(1 + \frac{s}{n} \right)$$

$$= 3300 \cdot \log_2 (1 + 10^2)$$

$$= 3300 \log_2 (10)$$

$$= 3300 \cdot 6.6 = 21945 \text{ bits/s}$$

$$\text{b. } 1 \cdot 10^9 \text{ bits/s} = 3300 \cdot \log_2 \left(1 + \frac{s}{n} \right)$$

$$\log_2 \left(1 + \frac{s}{n} \right) = 3.03 \cdot 10^5$$

$$\log_{10} \left(1 + \frac{s}{n} \right) = 3.03 \cdot 10^5 \cdot \log_{10} 2$$

$$\approx 91212$$

$$\approx 10^5$$

$$10 \log (s/n) = 10^5 \text{ dB}$$

$$\text{SNR} = 10^5 \text{ dB}$$

4.6

$$f(x_1, x_2, \dots, x_n) = \frac{1}{n} \sum_{i=1}^n x_i$$

MCQ A

$$\begin{aligned} \langle f(x_1, x_2, \dots, x_n) \rangle &= \left\langle \frac{1}{n} \sum_{i=1}^n x_i \right\rangle = \frac{1}{n} \sum_{i=1}^n \langle x_i \rangle \\ &= \frac{1}{n} \sum_{i=1}^n \overbrace{x_0}^{\text{✓}} \quad \sum_{i=1}^n x_0 = x_0 + x_0 + x_0 \dots \stackrel{n \text{ times}}{=} n \cdot x_0 \\ &= \frac{1}{n} \cdot n \cdot x_0 = x_0 \end{aligned}$$

Cramér - Rao Bound states that the variance of the estimator must be \geq the inverse of the Fisher Information

4.1

continuity