

Homework 1

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Problem 1

Define.

Cov = Getting COVID

DF = Disease Free

$+$ = Test positive

$-$ = Test negative

From the question, we learned

$$Sensitivity = P(+|Cov)$$

$$Specificity = P(-|DF)$$

a) $P(Cov) = 0.5\%$

b) $P(Cov) = 5\%$

We want $P(Cov|+)$ for base cases.

$$\begin{aligned} P(Cov|+) &= \frac{P(+|Cov) \cdot P(Cov)}{P(+)} \\ &= \frac{P(+|Cov) \cdot P(Cov)}{P(+|Cov) \cdot P(Cov) + P(+|DF) \cdot P(DF)} \\ &= \frac{Sensitivity \cdot P(Cov)}{Sensitivity \cdot P(Cov) + (1 - Specificity) \cdot (1 - P(Cov))} \end{aligned}$$

For a)

$$P(Cov|+) = \frac{0.65 \times 0.005}{0.65 \times 0.005 + 0.01 \times 0.995} = \frac{65}{264} \approx 0.246 = 24.6\%$$

For b)

$$P(Cov|+) = \frac{0.65 \times 0.05}{0.65 \times 0.05 + 0.01 \times 0.95} = \frac{65}{84} \approx 0.774 = 77.4\%$$

Problem 2

- a).** number of bits needed for keyboard symbols: 8bits
2615 characters total size = 2615 bytes
ls -alF shows 2659 bytes
not all characters on the keyboard, unicode resulting in this

b).

$$1\text{byte} = 2^3\text{bits}$$

$$1\text{KiB} = 2^{13}\text{bits}$$

$$1\text{MiB} = 2^{23}\text{bits}$$

$$1\text{GiB} = 2^{33}\text{bits}$$

$$1\text{byte} = 8\text{bits}$$

$$1\text{KB} = 8 \times 10^3\text{bits}$$

$$1\text{MB} = 8 \times 10^6\text{bits}$$

$$1\text{GB} = 8 \times 10^9\text{bits}$$

c).

$$80\text{GB} = 8 \times 10^9\text{bytes}$$

$$\text{number of pages} = 8 \times 10^9 / 2659 \approx 3008649.868 \approx 3008650$$

d). it means our information reduces $1\text{GB} = 8 \times 10^9\text{bits}$

$$\Delta S = (k_B \ln 2) \times 8 \times 10^9$$

$$\therefore Q = T \cdot \Delta S = 300 \times k_B \times \ln 2 \times 8 \times 10^9 = 300 \times 1.38 \times 10^{-23} \times \ln 2 \times 8 \times 10^9 J \approx 2.87 \times 10^{-12}$$

e). We need $80 \times 2.87 \times 10^{-12} \approx 2.30 \times 10^{-10} J$

Problem 3

a).

$$\begin{aligned}
 H(X, Y) &= - \sum_{i,j} P(X_i, Y_j) \log P(X_i, Y_j) \\
 &= - \sum_{i,j} P(Y_j, X_i) \log P(Y_j, X_i) \\
 &= H(Y, X) \\
 H(X : Y) &= H(X) + H(Y) - H(X, Y) \\
 &= H(Y) + H(X) - H(Y, X) \\
 &= H(Y : X)
 \end{aligned}$$

b). It's obvious the information of any event ≥ 0 , $H(Y|X)$ is always true

$$H(X : Y) = H(Y) + H(X) - H(X, Y)$$

$$\begin{aligned}
 \because H(X) - H(X, Y) &= - \sum_i P(X_i) \log P(X_i) + \sum_{i,j} P(Y_j|X_i) P(X_i) \log(P(Y_j|X_i) P(X_i)) \\
 &= \sum_i P(X_i) \sum_j (P(Y_j|X_i) \log(P(Y_j|X_i) P(X_i)) - \log(P(X_i)))
 \end{aligned}$$

$$\begin{aligned}
 &\because P(Y_j|X_i) \leq 1 \\
 &\therefore P(Y_j|X_i) \log(P(Y_j|X_i) P(X_i)) \leq \log(P(Y_j|X_i) P(X_i)) \leq \log(P(X_i)) \\
 &\therefore \sum_i P(X_i) \sum_j (P(Y_j|X_i) \log(P(Y_j|X_i) P(X_i)) - \log(P(X_i))) \leq 0 \\
 &\therefore H(X) \leq H(X, Y)
 \end{aligned}$$

Meanwhile,

$$\begin{aligned}
 P(Y_j|X_i) \log(P(Y_j|X_i) P(X_i)) &= \log(P(Y_j|X_i) P(X_i)) = \log(P(X_i)) \\
 \iff P(Y_j|X_i) &= 1 \\
 \iff Y &= f(X)
 \end{aligned}$$

c).

$$\begin{aligned}
 H(X, Y) &= H(X) + H(Y) - H(X : Y) \\
 \because H(X : Y) &\geq 0 \\
 \therefore H(X, Y) &\leq H(X) + H(Y)
 \end{aligned}$$

When equality,

$$H(X : Y) = 0$$

$$\iff \log\left(\frac{P(X, Y)}{P(X)P(Y)}\right) = 1$$

$$\iff P(X, Y) = P(X)P(Y)$$

$$\iff X \text{ and } Y \text{ are independent}$$