

Introduction to Quantum Information Processing

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1 Mutual Information - Solution

1. We expand the expression to get:

$$I(X; Y) = H(X) - H(X|Y) = - \sum_x p(x) \log_2 p(x) + \sum_{x,y} p(x,y) \log_2 p(x|y)$$

We know that a probability distribution of variable X can be computed from the mutual probability of X and Y by summing all the values in range for Y :

$$p(x) = \sum_y p(x,y)$$

Plugging this into the previous equation, we get:

$$I(X; Y) = \sum_{x,y} p(x,y) \log_2 p(x|y) - \sum_{x,y} p(x,y) \log_2 p(x) = \sum_{x,y} p(x,y) \log_2 \left(\frac{p(x|y)}{p(x)} \right)$$

From logarithm rules and conditional probability definition:

$$\log_2 \left(\frac{p(x|y)}{p(x)} \right) = - \log_2 \left(\frac{p(x)p(y)}{p(x,y)} \right)$$

Plugging it, we get:

$$I(X; Y) = - \sum_{x,y} p(x,y) \log_2 \left(\frac{p(x)p(y)}{p(x,y)} \right)$$

2. We can write $I(X; Y)$ using the logarithm-identity $\log_a x \cdot \log_b a = \log_b x$ as:

$$K \cdot I(X; Y) = - \sum_{x,y} p(x,y) \ln \left(\frac{p(x)p(y)}{p(x,y)} \right)$$

Where $K = \ln 2 > 0$. By using the identity $\ln t \leq t - 1$ for $t > 0$ and knowing that $\frac{p(x)p(y)}{p(x,y)}$ is non-negative (because probability cannot be negative, obv.) we show:

$$-K \cdot I(X; Y) \leq \sum_{x,y} p(x,y) \left(\frac{p(x)p(y)}{p(x,y)} - 1 \right) = \sum_{x,y} (p(x)p(y) - p(x,y))$$

$\ln t \leq t - 1$ is equality **iff** $t = 1$.

Corollary 1.0.1. $-K \cdot I(X; Y) = \sum_{x,y} (p(x)p(y) - p(x,y))$ iff X, Y are independent. **Proof:** If X, Y are independent $\iff p(x)p(y) = p(x,y)$ for every pair of $x, y \iff \forall x, y : \frac{p(x)p(y)}{p(x,y)} = 1 \iff$

We expand the expression to get sums over probabilities (which we can reduce to 1):

$$\begin{aligned}
 -K \cdot I(X; Y) &\leq \sum_x \left(p(x) \cdot \sum_y p(y) \right) - \sum_{x,y} p(x, y) \\
 \Rightarrow I(X; Y) &\geq K \cdot \sum_x p(x) - 1 = 0 \Rightarrow \boxed{I(X; Y) \geq 0}
 \end{aligned}$$

From 1.0.1 and the above expansion we can conclude that $I(X; Y) = 0 \iff X, Y$ are independent.

2 Entropy and Mutual Information

1.

$$\begin{aligned}
 Y &= \begin{cases} 1, & \text{the keys are in the pocket} \\ 0, & \text{otherwise} \end{cases} \\
 X &= \begin{cases} i \in [1, 100], & \text{the keys are in the } i\text{-th location} \\ 0, & \text{otherwise} \end{cases}
 \end{aligned}$$

2.

	Y = 0	Y = 1	
X = 0	$p(X = 0, Y = 0) = 0$ $p(X = 0 Y = 0) = 0$ $p(Y = 0 X = 0) = 0$		$p(X = 0) = 0.99$ $p(X = 0, Y = 1) = 0.99$ $p(X = 0 Y = 1) = 1$ $p(Y = 1 X = 0) = 1$
X = i > 0	$i > 0$ $p(X = i, Y = 1) = 0$ $p(X = i Y = 1) = 0$ $p(Y = 1 X = i) = 0$		$p(X = i > 0) = 0.0001$ $i > 0$ $p(X = i, Y = 0) = 0.0001$ $p(X = i Y = 0) = 0.01$ $p(Y = 0 X = i) = 1$
	$p(Y = 0) = 0.01$	$p(Y = 1) = 0.99$	1

3.

$$\begin{aligned}
 H(X) &= - \sum_x p(X = x) \log_2 p(X = x) \\
 &= -p(X = 0) \log_2 p(X = 0) - 100 \cdot (X = i) \log_2 p(X = i) \\
 &= -0.99 \cdot \log_2 0.99 - 100 \cdot 0.0001 \log_2 0.0001 \\
 &= 0.044
 \end{aligned}$$

$$\begin{aligned}
 H(Y) &= - \sum_y p(Y = y) \log_2 p(Y = y) \\
 &= -p(Y = 0) \log_2 p(Y = 0) - (Y = 1) \log_2 p(Y = 1) \\
 &= -0.01 \cdot \log_2 0.01 - 0.99 \log_2 0.99 \\
 &= 0.024
 \end{aligned}$$

$$\begin{aligned}
 H(X|Y=0) &= - \sum_x p(X|Y=0) \log_2 p(X|Y=0) \\
 &= -p(X=0|Y=0) \log_2 p(X=0|Y=0) - 100 \cdot (X=i|Y=0) \log_2 p(X=i|Y=0) \\
 &= -100 \cdot 0.01 \log_2 0.01 \\
 &= 2
 \end{aligned}$$

$$\begin{aligned}
 H(X|Y=1) &= - \sum_x p(X|Y=1) \log_2 p(X|Y=1) \\
 &= -p(X=0|Y=1) \log_2 p(X=0|Y=1) - 100 \cdot (X=i|Y=1) \log_2 p(X=i|Y=1) \\
 &= -1 \cdot 0 \\
 &= 0
 \end{aligned}$$

$$\begin{aligned}
 H(X|Y) &= - \sum_{x,y} p(X=x, Y=y) \log_2 p(X=x|Y=y) \\
 &= -p(X=0, Y=0) \log_2 p(X=0|Y=0) - \sum_{i=1}^{100} p(X=i, Y=0) \log_2 p(X=i|Y=0) \\
 &\quad - p(X=0, Y=1) \log_2 p(X=0|Y=1) - \sum_{i=1}^{100} p(X=i, Y=1) \log_2 p(X=i|Y=1) \\
 &= 0 - 100 \cdot 0.0001 \log_2 0.01 - 0.99 \log_2 1 - 100 \cdot 0 \\
 &= 0.02
 \end{aligned}$$

$$\begin{aligned}
 H(Y|X) &= - \sum_{x,y} p(X=x, Y=y) \log_2 p(Y=y|X=x) \\
 &= -p(X=0, Y=0) \log_2 p(Y=0|X=0) - \sum_{i=1}^{100} p(X=i, Y=0) \log_2 p(Y=0|X=i) \\
 &\quad - p(X=0, Y=1) \log_2 p(Y=1|X=0) - \sum_{i=1}^{100} p(X=i, Y=1) \log_2 p(Y=1|X=i) \\
 &= 0 - 0 - 0 - 0 \\
 &= 0
 \end{aligned}$$

4.