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Tutorial 1

- Proposed Solution -

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Solution of Problem 1

(Conditional Entropy)

Let X, Y, Z be discrete random variables. Proof that:

$$\text{a) } 0 \stackrel{(i)}{\leq} H(X|Y) \stackrel{(ii)}{\leq} H(X).$$

- Equality holds in (i) $\Leftrightarrow X$ is totally dependent on Y .
- Equality holds in (ii) $\Leftrightarrow X$ and Y are independent.

Equality (i) holds iff $H(X|Y) = 0$, that is if X is a deterministic function of Y .

Similarly, (ii) holds iff $H(X) = H(X|Y)$. Since

$$H(X|Y) = H(X) - I(X; Y)$$

we have that this holds only if $I(X; Y) = 0$, that is X and Y are statistically independent.

$$\text{b) } H(X|Y, Z) \leq \min\{H(X|Y), H(X|Z)\}.$$

$$H(X|Y, Z) = H(X|Z) - \underbrace{I(X; Y|Z)}_{\geq 0} \leq H(X|Z).$$

The same can be said for $H(X|Y, Z) \leq H(X|Y)$.

Solution of Problem 2

(Sequence of Random Variables)

Let $X_0 \in \mathcal{X}$ be a discrete random variable with distribution μ_0 . Given the stochastic matrix Π , let X_1, X_2, \dots be the sequence of random variables (with the same support \mathcal{X}) such that μ_n is the distribution¹ of X_n and $\mu_n = \mu_{n-1}\Pi$ for all $n = 1, 2, \dots$

$$\text{a) Show that } \mu_n = \mu_0 \Pi^n.$$

$$\mu_n = \mu_{n-1}\Pi = (\mu_{n-2}\Pi)\Pi = \dots = \mu_0 \underbrace{\Pi\Pi \cdots \Pi}_{n \text{ times}} = \mu_0 \Pi^n.$$

¹Note that μ_n are row vectors for all $n = 0, 1, 2, \dots$

Now assume that μ_n converges to some distribution μ^* , that is $\lim_{n \rightarrow \infty} \mu_n = \mu^*$ and $\mu^* \Pi = \mu^*$.

b) Proof that $D(\mu_n \|\mu^*) \geq D(\mu_{n+1} \|\mu^*)$ for all $n = 1, 2, \dots$

$$D(\mu_n \|\mu^*) = D(\mu_{n-1} \Pi \|\mu^* \Pi) \stackrel{(2.1.13)}{\leq} D(\mu_{n-1} \|\mu^*).$$

c) Show that if μ^* is the uniform distribution then $H(X_n) \leq H(X_{n+1})$ for all $n = 1, 2, \dots$

Since μ^* is the uniform distribution we have

$$D(\mu_n \|\mu^*) = \log |\mathcal{X}| - H(X_n). \quad (\text{see proof of theorem 2.1.14})$$

Then we get

$$\begin{aligned} D(\mu_n \|\mu^*) &\geq D(\mu_{n+1} \|\mu^*) \\ \Rightarrow \log |\mathcal{X}| - H(X_n) &\geq \log |\mathcal{X}| - H(X_{n+1}) \\ \Rightarrow H(X_n) &\leq H(X_{n+1}). \end{aligned}$$

Solution of Problem 3

(A Metric)

A function $\rho(x, y)$ is a metric if for all x, y ,

- $\rho(x, y) \geq 0$,
- $\rho(x, y) = \rho(y, x)$,
- $\rho(x, y) = 0$ if and only if $x = y$,
- $\rho(x, y) + \rho(y, z) \geq \rho(x, z)$.

a) Show that $\rho(X, Y) = H(X|Y) + H(Y|X)$ satisfies the first, second and fourth properties above. If we say that $X = Y$ if there is a one-to-one function mapping from X to Y , then the third property is also satisfied, and $\rho(X, Y)$ is a metric.

- Since conditional entropies are non-negative we have that $\rho(X, Y) \geq 0$.
- By definition $\rho(X, Y) = \rho(Y, X)$.
- $H(X|Y) = 0$ iff there is a deterministic mapping from Y to X . In the same manner $H(Y|X) = 0$ iff there is a deterministic mapping from X to Y . Therefore, $H(X|Y) = 0$ and $H(Y|X) = 0$ only if there is a one-to-one mapping between X and Y .
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$$\begin{aligned} H(X|Y) + H(Y|Z) &\geq H(X|Y, Z) + H(Y|Z) \\ &= H(X|Y, Z) \\ &= H(X|Z) + H(Y|X, Z) \\ &\geq H(X|Z). \end{aligned}$$

Then it follows $\rho(X, Y) + \rho(Y, Z) \geq \rho(X, Z)$.

b) Verify that $\rho(X, Y)$ can also be expressed as

$$\begin{aligned}\rho(X, Y) &= H(X) + H(Y) - 2I(X; Y) \\ &= H(X, Y) - I(X; Y) \\ &= 2H(X, Y) - H(X) - H(Y)\end{aligned}$$

$$\begin{aligned}\rho(X, Y) &= \underbrace{H(X|Y)}_{=H(X)-I(X;Y)} + \underbrace{H(Y|X)}_{=H(Y)-I(X;Y)} = H(X) + H(Y) - 2I(X; Y) \\ &= \underbrace{H(X) + H(Y) - I(X; Y)}_{H(X, Y)} - I(X; Y) = H(X, Y) - I(X; Y) \\ &= H(X, Y) - \underbrace{I(X; Y) + H(X) + H(Y)}_{H(X, Y)} - H(X) - H(Y) = 2H(X, Y) - H(X) - H(Y).\end{aligned}$$

Solution of Problem 4

(A Measure of Correlation)

Let X_1 and X_2 be identically distributed random variables, but not necessarily independent. Let

$$\rho = 1 - \frac{H(X_2|X_1)}{H(X_1)}.$$

a) Show that $\rho = \frac{I(X_1; X_2)}{H(X_1)}$.

Since X_1 and X_2 be identically distributed, we have that $H(X_1) = H(X_2)$. Then

$$\rho = \frac{H(X_1) - H(X_2|X_1)}{H(X_1)} = \frac{H(X_2) - H(X_2|X_1)}{H(X_1)} = \frac{I(X_1; X_2)}{H(X_1)}.$$

b) Show that $0 \leq \rho \leq 1$.

Since $I(X_1; X_2)$ and $H(X_1)$ are non-negative $\rho \geq 0$. On the other hand

$$I(X_1; X_2) = H(X_1) - \underbrace{H(X_1|X_2)}_{\geq 0} \leq H(X_1) \Rightarrow \rho \leq 1.$$

c) When is $\rho = 0$?

$$\rho = 0 \Leftrightarrow I(X; Y) = 0 \Leftrightarrow X, Y \text{ are independent.}$$

d) When is $\rho = 1$?

$\rho = 1 \Leftrightarrow H(X_1|X_2) = 0$, by symmetry $\rho = 1 \Leftrightarrow H(X_2|X_1) = 0$. Therefore, $\rho = 1 \Leftrightarrow X, Y$ have a one-to-one relationship.

Solution of Problem 5

(Entropy of a Sum)

Let X and Y be random variables that take on values x_1, x_2, \dots, x_r and y_1, y_2, \dots, y_r respectively. Let $Z = X + Y$.

- a) Show that $H(Z|X) = H(Y|X)$. Argue that if X, Y are independent, then $H(Y) \leq H(Z)$ and $H(X) \leq H(Z)$. Thus the addition of *independent* random variables adds uncertainty.

Since $Z = X + Y$ we have

$$P(Z = z|X = x) = P(Y = z - x|X = x).$$

Then,

$$\begin{aligned} H(Z|X) &= \sum_{x \in \mathcal{X}} P(X = x) H(Z|X = x) \\ &= - \sum_{x \in \mathcal{X}} P(X = x) \sum_{z \in \mathcal{Z}} P(Z = z|X = x) \log P(Z = z|X = x) \\ &= - \sum_{x \in \mathcal{X}} P(X = x) \sum_{z \in \mathcal{Z}} P(Y = \underbrace{z - x}_y | X = x) \log P(Y = z - x | X = x) \\ &= - \sum_{x \in \mathcal{X}} P(X = x) \sum_{y \in \mathcal{Y}} P(Y = y|X = x) \log P(Y = y|X = x) \\ &= \sum_{x \in \mathcal{X}} P(X = x) H(Y|X = x) \\ &= H(Y|X). \end{aligned}$$

- b) Give an example of (necessarily dependent) random variables in which $H(X) > H(Z)$ and $H(Y) > H(Z)$.

Let X be uniformly distributed and $Y = -X$, thus Y is also uniform. Then $Z = X + Y = 0$, thus $H(Z) = 0$ while $H(X) = \log |\mathcal{X}| > 0$ and $H(Y) = \log |\mathcal{Y}| > 0$.

- c) Under what conditions does $H(Z) = H(X) + H(Y)$?

$$H(Z) \stackrel{(i)}{\leq} H(X, Y) \stackrel{(ii)}{\leq} H(X) + H(Y).$$

$H(Z) = H(X) + H(Y)$ iff (i), (ii) hold with equality, that is

- (i): (X, Y) has a one-to-one relation with Z ,
- (ii): X and Y are independent.