

ex

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1a

$H((X + Y) | X) = H(Y | X)$ follows from the fact that

$$p(x + y | x) := \frac{P(X = x, X + Y = x + y)}{P(X = x)} = \frac{P(X = x, Y = y)}{P(X = x)} = p(y | x),$$

because the two conditions $(X = x, X + Y = x + y)$ and $(X = x, Y = y)$ are equivalent:

$$\begin{cases} X = x \\ Y = y \end{cases} \iff \begin{cases} X = x \\ X + Y = x + y \end{cases}.$$

This yields

$$\begin{aligned} H((X + Y) | X) &= -E \log p((X + Y) | X) \\ &= -E \log p(Y | X) \\ &= H(Y | X). \end{aligned}$$

$H(X + Y) \geq H(Y)$ for independent X, Y follows from

$$\begin{aligned} 0 &\leq I(X + Y; X) \\ &= H(X + Y) - H(X + Y | X) \\ &\stackrel{1a}{=} H(X + Y) - H(Y | X) \\ &\stackrel{ind.}{=} H(X + Y) - H(Y) \\ &\iff H(X + Y) \geq H(Y). \end{aligned}$$

1b

In previous courses a statistic $T(X)$ was said to be sufficient for θ if the conditional distribution of X given T was independent of θ , meaning $p(X | T(X), \theta) = p(X | T(X))$ (we will call this "def 1" when we use this to justify an equality below). Multiple applications of Bayes' rule give

$$p(\theta | X, T(X)) = \frac{p(X | \theta, T(X))p(\theta | T(X))}{p(X | T(X))} \stackrel{def 1}{=} \frac{p(X | T(X))p(\theta | T(X))}{p(X | T(X))} = p(\theta | T(X)).$$

Also, because $T(X)$ is a function of X , we have $P(T(X) = t, X = x) = P(X = x)$ for all values of $(x, t(x))$, meaning $p(X, T(X)) = p(X)$, and analogously $p(\theta, X, T(X)) = p(\theta, X)$. This yields

$$p(\theta | X, T(X)) = \frac{p(\theta, X, T(X))}{p(X, T(X))} = p(\theta | X).$$

Since Shannon entropy only depends on the probability function, this means that

$$H(\theta | X, T(X)) = H(\theta | T(X)) = H(\theta | X).$$

Once we have shown this, it is very easy to show that the definition of sufficiency given in previous courses is equivalent to the information-theoretical one. Namely,

$$I(\theta; X, T(X)) = H(\theta) - H(\theta | X, T(X)) = H(\theta) - H(\theta | T(X)) = I(\theta; T(X)),$$

$$I(\theta; X, T(X)) = H(\theta) - H(\theta | X, T(X)) = H(\theta) - H(\theta | X) = I(\theta; X)$$

meaning $I(\theta; T(X)) = I(\theta; X)$, which is precisely the information-theoretical definition of sufficiency ("def 2").

We shall now show that def 1 \implies def 2. This follows from a few identities which are easily seen by drawing out the Venn-diagram visualization of entropies and informations for three variables. We have

$$I(\theta; X | T(X)) = I(\theta; X) - I(\theta; X; T(X))$$

OBS! Mutual information only defined for 2 RVs! Maybe skip this step

$$= I(\theta; X) - (I(\theta; X) - H(\theta | X) + H(\theta | X, T(X)))$$

$$= H(\theta | X) - H(\theta) + I(\theta; X) + I(\theta, T(X)) - I(\theta; X, T(X))$$

Well... assuming that $I(\theta; X, T(X))$ motsvarar det jag trodde hette $I(\theta; X; T(X))$

alltså intersektionen av alla tre regioner i venn-diagrammet

$$= I(\theta; T(X)) - I(\theta; X, T(X))$$

$$\stackrel{def 2}{=} I(\theta; X) - I(\theta; X, T(X)).$$

Again, being a function of X , $T(X)$ doesn't bring any new information in the system than X already contains. Thus we have $I(\theta; X, T(X)) = I(\theta; X)$, giving

$$I(\theta; X | T(X)) = I(\theta; X) - I(\theta; X) = 0$$

which can only be the case if $X | T(X)$ is independent of θ , i.e. def 1.

We have therefore shown that the two definitions are equivalent.