

Information Theory

Problem Set 07 - Communication Over a Noisy Channel

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1. (a) A discrete memoryless channel is characterized in a formal way as an input alphabet \mathcal{A}_x , an output alphabet \mathcal{A}_y and a set of conditional probability distributions $P(y|x)$, for each $x \in \mathcal{A}_x$.
(b) If a channel is noisy, there is a non null probability that the message sent by a source will be different than the message received by the receiver. The noisy channel will affect the message, making it not easy to decode. To solve this issue, some kind of approach needs to be used to establish a reliable communication over a noisy channel.
(c)
(c)
2. We can solve this using Bayes Theorem.

$$P(X = 1|Y = 0) = \frac{P(Y = 0|X = 1) * P(X = 1)}{P(Y = 0)}$$

$$P(X = 1|Y = 0) = \frac{0.15 * 0.1}{P(Y = 0|X = 0) * P(X = 0) + P(Y = 0|X = 1) * P(X = 1)}$$

$$P(X = 1|Y = 0) = \frac{0.15 * 0.1}{0.85 * 0.9 + 0.15 * 0.1} = 0.0192$$

3. We can again use Bayes Theorem.

$$P(X = 1|Y = 0) = \frac{P(Y = 0|X = 1) * P(X = 1)}{P(Y = 0)}$$

$$P(X = 1|Y = 0) = \frac{0.15 * 0.1}{P(Y = 0|X = 0) * P(X = 0) + P(Y = 0|X = 1) * P(X = 1)}$$

$$P(X = 1|Y = 0) = \frac{0.15 * 0.1}{0.15 * 0.1 + 1.0 * 0.9} = 0.016$$

- 4.

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

$$H(X) = \sum_{x \in X} P(X = x) * \frac{1}{\log_2(P(X = x))}$$

$$H(X|Y) = \sum_{y \in Y} P(X = x^Y = y) * \frac{1}{\log_2(P(X = x|Y = y))}$$

References

- [1] David J. C. MacKay. *Information Theory, Inference and Learning Algorithms*. 7th edition, 2005.