

# Mutual Information

Mutual information is one of many quantities that measures how much one random variable gives about another. It is a dimensionless quantity. Mutual Information can be thought of as the reduction in uncertainty about one random variable given knowledge of another.

- High mutual information indicates a large reduction in uncertainty,
- low mutual information indicates a small reduction,
- zero mutual information between two random variables means the variables are independent.

Efficient communication systems have high mutual information.

# Mutual Information

## Joint Entropies:

Using the input probabilities  $P(x_i)$ , output probabilities  $P(y_i)$ , transition probabilities  $P(y_i|x_i)$ , and joint probabilities  $P(x_i, y_j)$ , we can define the following various entropy functions for a channel with  $m$  inputs and  $n$  outputs:

- $H(X) = -\sum_{i=1}^m P(x_i) \log_2 P(x_i)$
- $H(Y) = -\sum_{j=1}^n P(y_j) \log_2 P(y_j)$
- $H(X, Y) = -\sum_{j=1}^n \sum_{i=1}^m P(x_i, y_j) \log_2 P(x_i, y_j)$

# Mutual Information: Joint Entropy

These entropies can be interpreted as follows:

- $H(X)$  is the average uncertainty of the channel input, and  $H(Y)$  is the average uncertainty of the channel output.
- The joint entropy  $H(X, Y)$  is the average uncertainty of the communication channel as a whole.

# Mutual Information: Conditional Entropy

- The conditional entropy  $H(X|Y)$  is a measure of the average uncertainty remaining about the channel input after the channel output has been observed.
- This is sometimes called the equivocation of  $X$  with respect to  $Y$ .
- The conditional entropy  $H(Y|X)$  is the average uncertainty of the channel output given that  $X$  was transmitted.
- $H(X|Y) = -\sum_{j=1}^n \sum_{i=1}^m P(x_i, y_j) \log_2 P(x_i|y_j)$
- $H(Y|X) = -\sum_{j=1}^n \sum_{i=1}^m P(x_i, y_j) \log_2 P(y_j|x_i)$

# Mutual Information : Useful Identities

Two useful relationships among the types of entropies are

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

(Remark : compare to identities in probability theory)

# Mutual Information : Useful Identities

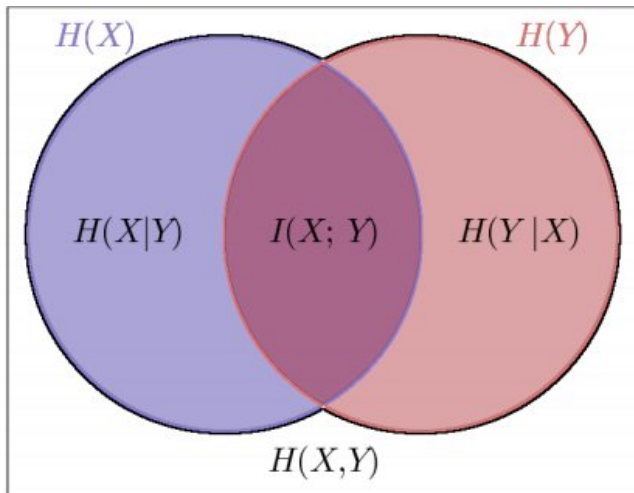


Figure:

# Mutual Information

The mutual information  $I(X; Y)$  of a channel is defined by

$$I(X; Y) = H(X) - H(X|Y) \text{ (b/symbol)}$$

Alternatively we can define it as either

$$I(X; Y) = H(Y) - H(Y|X) \text{ (b/symbol)}$$

or as

$$I(X; Y) = H(Y) + H(X) - H(X, Y) \text{ (b/symbol)}$$

Remark: The mutual information is the reduction of entropy of  $X$  when  $Y$  is known.