#### Mutual information: lecture 4

COMSM0075 Information Processing and Brain

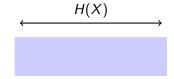
comsm0075.github.io

September 2020

The chain rule for entropy

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

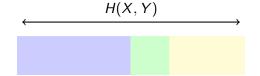
## H(X, Y) = H(X) + H(Y|X)



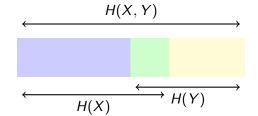
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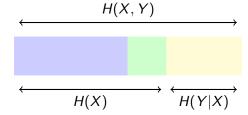
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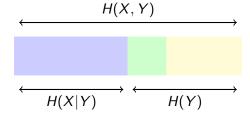
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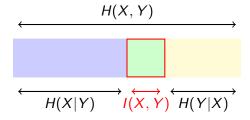


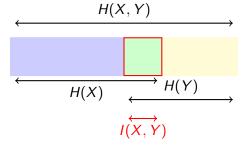
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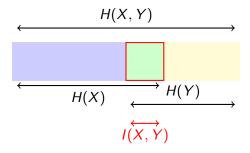


## H(X, Y) = H(Y) + H(X|Y)









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

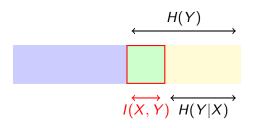
$$\longleftrightarrow H(X)$$

$$\longleftrightarrow H(X|Y) \longrightarrow I(X,Y)$$

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

then substitute H(X,Y) = H(Y) + H(X|Y) to get

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



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### All the mutual informations

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

I(X, Y) = H(X, Y) - H(X|Y) - H(Y|X)

#### Mutual information - direct formula

By substituting the formulas for H(X), H(Y) and H(X, Y) we get

$$I(X,Y) = \sum_{i,j} p_{X,Y}(x_i, y_j) \log_2 \frac{p_{X,Y}(x_i, y_j)}{p_X(x_j)p_Y(y_j)}$$

## Example

$$\begin{array}{c|cccc} & x_0 & x_1 \\ \hline y_0 & 1/4 & 1/4 \\ y_1 & 1/2 & 0 \\ \end{array}$$

has 
$$H(X, Y) = 3/2$$
,  $H(X) \approx 0.81$  and  $H(Y) = 1$  so

$$I(X, Y) \approx 0.31$$

## Mutual information - independent variables

$$I(X, Y) = \sum_{i,j} p_{X,Y}(x_i, y_j) \log_2 \frac{p_{X,Y}(x_i, y_j)}{p_X(x_j)p_Y(y_j)}$$

but if X and Y are independent  $p_{X,Y}(x_i,y_j)=p_X(x_j)p_Y(y_j)$  and hence

$$I(X,Y)=0$$

Mutual information - independent variables

In fact

$$I(X, Y) \geq 0$$

with equality if and if X and Y are independent.

## Mutual information - independent variables

In fact

$$I(X, Y) \geq 0$$

with equality if and if X and Y are independent; note that this is equivalent to

$$H(X) \ge H(X|Y)$$

with equality if and if X and Y are independent; as claimed earlier.

#### Correlation

$$C(X,Y) = \frac{\langle (X - \mu_X)(Y - \mu_Y) \rangle}{\sigma_X \sigma_Y}$$

where  $\mu_X$  is the average of X and  $\sigma_X$  is its standard deviation with similar notation for Y.

#### Correlation

#### Consider

$$\begin{array}{c|ccccc} & -1 & 0 & 1 \\ \hline 1 & 1/4 & 0 & 1/4 \\ 0 & 0 & 1/2 & 0 \\ \end{array}$$

then

$$C(X,Y)=0$$

whereas I(X, Y) = 1.