

# Information and Computation

## Lecture 1

*Joint entropy, conditional entropy and mutual information*

# A probability refresher

- **Definition.** A joint ensemble  $XY$  is a tuple  $(A_X \times A_Y, p_{XY})$  with

$$\begin{aligned} p_X(x) &= \sum_{y \in A_Y} p_{XY}(x, y) \\ p_Y(y) &= \sum_{x \in A_X} p_{XY}(x, y) \\ p_{XY}(x|y) &= \frac{p_{XY}(x, y)}{p_Y(y)} \end{aligned}$$

# Joint entropy

- **Definition.** The entropy (rate) of an ensemble  $XY$  is

$$H(XY) = - \sum_{x \in X, y \in Y} p_{XY}(x, y) \log p_{XY}(x, y)$$

- **Example.**

$$p(\text{Heads}, \text{Heads}) = \frac{1}{4}$$

$$H(XY) = \sum \frac{1}{4} \log \frac{1}{4}$$

$$= \frac{1}{4} (2 + 2 + 2 + 2) = 2$$

# Conditional entropy

- **Definition.** Conditional information,

$$h(x|y) = -\log(p(x|y))$$

- **Definition.** Conditional entropy given event  $y$

$$H(X|y) = - \sum_{x \in A_X} p(x|y) \log(p(x|y))$$

- **Definition.** Conditional entropy

$$H(X|Y) = \sum_y p_Y(y) H(X|y)$$

	Days with no rain	Days with rain
Rotterdam	212	153
Atacama desert	360	5

**Exercise 1.1** - Let us assume that you are living in the Atacama desert where it rarely rains. How much information do the messages 0 and 1 carry?

	Days with no rain	Days with rain
Rotterdam	212	153
Atacama desert	360	5

**Exercise 1.1** - Let us assume that you are living in the Atacama desert where it rarely rains. You receive a 0. How much information does this message carry?

**Exercise 1.2** - Now let us assume that you live in the Netherlands where it does rain quite often, but certainly not every day. How much information do the messages 0 and 1 carry?

	Days with no rain	Days with rain
Summer	182	1
Winter	30	152

**Exercise 1.1** - Let us assume that you are living in the Atacama desert where it rarely rains. You receive a 0. How much information does this message carry?

**Exercise 1.2** - Now let us assume that you live in the Netherlands where it does rain quite often, but certainly not every day. You receive a 0. Does the message contain information?

**Exercise 1.3** - Finally, let us assume that you live in the Netherlands but (boldly) also that you are aware of the current season. You receive a 0. Does the message 0 carry the same information in summer and in winter?

	Days with no rain	Days with rain
Summer	182	1
Winter	30	152

- Compute  $H(\text{Season})$ ,  $H(\text{Weather})$
- Compute  $H(\text{Season}, \text{Weather})$
- Compute  $H(\text{Season} | \text{Weather} = \text{rain})$

$P$	No R	RAIN
	$\frac{182}{365}$	$\frac{1}{365}$
$\sum_m$		
Winter	$\frac{30}{365}$	$\frac{152}{365}$



# Exercise

- Show that  $H(X|Y) \geq 0$

## Exercise. Chain rule

- Show that  $H(XY) = H(X|Y) + H(Y)$

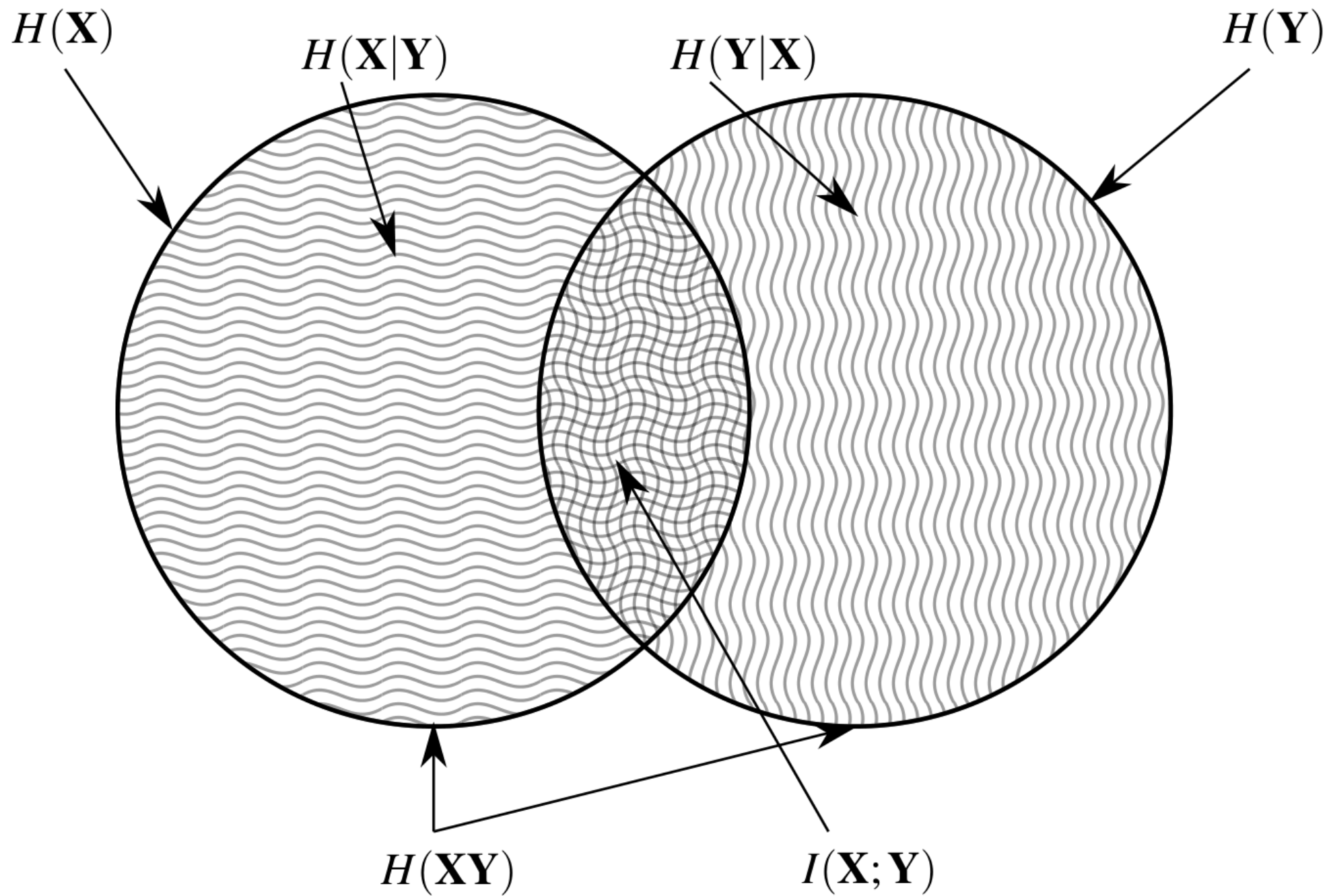
# Exercise

- Show that
  - $H(X, Y) \geq H(X)$
  - $H(X) \geq H(X|Y)$  (conditioning always reduces entropy)

# Mutual information

- **Definition.** Mutual information

$$I(X; Y) = H(X) + H(Y) - H(XY)$$



# Exercise

- Show that
  - $\min(H(X), H(Y)) \geq I(X; Y) \geq 0$

# Data processing inequality

- Show that  $I(X; Y) \geq I(X; f(Y))$

Let  $Z = f(Y)$  and let us expand  $I(X; YZ)$  in two ways:

$$\begin{aligned} 1) \quad I(X; YZ) &= \underbrace{H(X)}_{\text{red}} + H(YZ) - H(XYZ) + \underbrace{H(Y)}_{\text{red}} - H(Y) + H(XY) - \underbrace{H(XY)}_{\text{red}} \\ &= \underbrace{I(X; Y)}_{\text{red}} + \underbrace{H(YZ)}_{\text{green}} - \underbrace{H(XYZ)}_{\text{blue}} - \underbrace{H(Y)}_{\text{green}} + \underbrace{H(XY)}_{\text{blue}} \\ &= I(X; Y) + \underbrace{H(Z|Y)}_{\text{green}} - \underbrace{H(Z|XY)}_{\text{blue}} \\ &= I(X; Y) \end{aligned}$$

$$\begin{aligned}
2) \quad I(X:YZ) &= I(X:Z) + H(Y|Z) - H(XY|Z) - H(Z) + H(XZ) \\
&= I(X:Z) + H(X|Z) - H(XY|Z) \\
&= I(X:Z) + \sum_z p(z) (H(X|z) - H(XY|z)) \\
&= I(X:Z) + \sum_z p(z) I(X:Y|z)
\end{aligned}$$

Hence:

$$\begin{aligned}
I(X:Y) &= I(X:Z) + \sum_z p(z) I(X:Y|z) \\
&\geq I(X:Z)
\end{aligned}$$



# You will do great in the exam if you can

- Evaluate entropy, conditional entropy and mutual information
- Derive basic properties of the above

# Resources

- Lecture notes
- Slides
- “Elements of information theory”. Cover and Thomas. Chapter 2.1-2.8
- “Information Theory, Inference, and Learning Algorithms”. MacKay. Chapter 2