

W01 ~ Information Theory

Claude Shannon (1912-2001)

Mt. Auburn

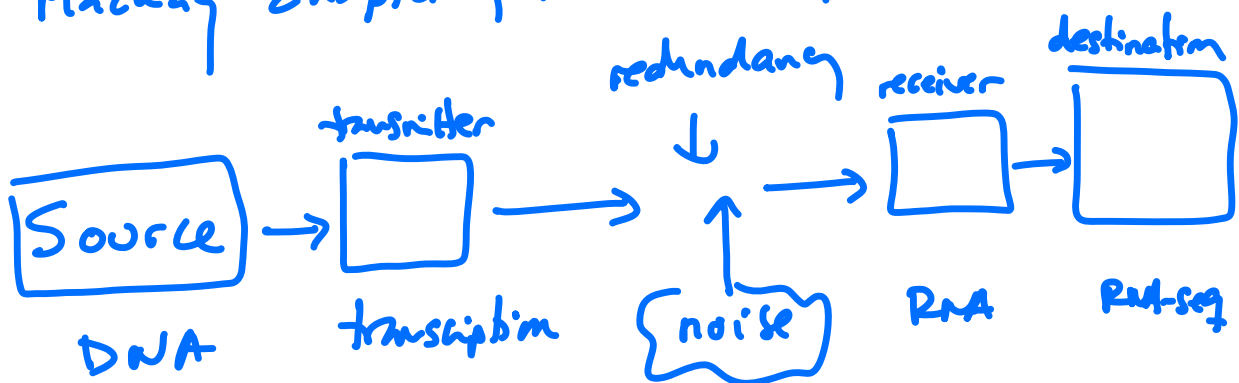
Betty Shannon.

1948

"A mathematical theory of communication"

p1,2
Section 6

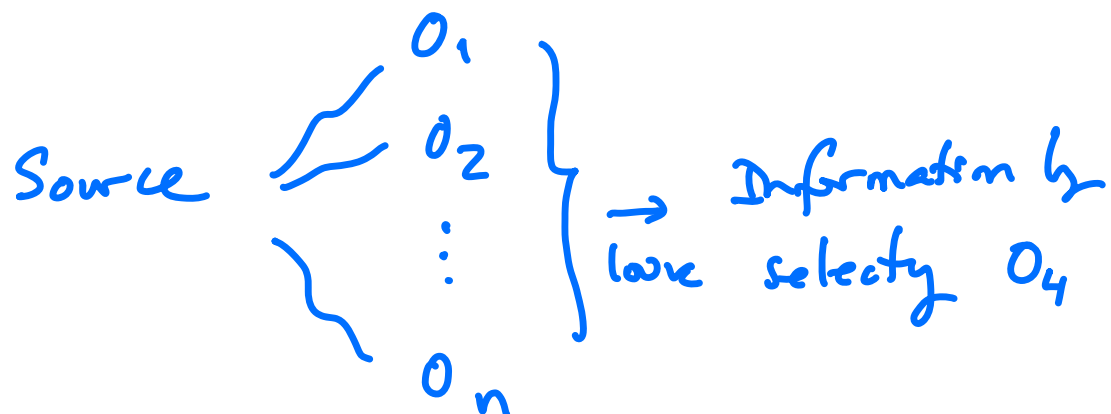
Mackay chapter 2, lectures 1,2



How much information you need to send so that the message is reproduced faithfully?

} Need to quantify Information

- 1 - I had breakfast this morning
- 2 - Today is NOT my birthday
- 3 - Today IS my birthday
- 4 - I ran the 1st Cambridge $\frac{1}{2}$ (2016)
- 5 - I " 1st + 2nd " " (2017)
- 6 - I " 1 + 2 + 3 " " (2019)
- (4th) 17 Nov 2021
- 7 - I've been to Antarctica.



Outcome/Event has prob p

Information of observing $0?$

$$(1) I(o) \propto \frac{1}{p} \quad " \quad p \uparrow \quad I \downarrow$$

* The rarer the event, the lower its probability, and the more info you would obtain by seeing it

* The more possibilities (assuming all equally likely) the more ignorant about outcome, the more info by observing one

$$I \sim n \text{ (\# of outcomes)}$$

(2) Information should be additive

if O_1, O_2 are independent,

$$I(O_1, O_2) = I(O_1) + I(O_2)$$

→ Shannon proposed

$$\left\{ I(o) = \log \frac{1}{p} = -\log p \right.$$

In Shannon's words:

$\left. \begin{array}{l} \text{'useful'} \\ \text{'intuitive'} \\ \text{'suitable'} \end{array} \right\} \text{ subjective!}$

$$O_1 \perp O_2$$

$$P(O_1, O_2) = P(O_1) \cdot P(O_2)$$

$$I(O_1, O_2) = -\log P(O_1, O_2) =$$

$$= -\log P(O_1) \cdot P(O_2)$$

$$= -\log P(O_1) - \log P(O_2)$$

$$= I(O_1) + I(O_2)$$

$$- P_i^2$$

$$\log(a \cdot b) = \log a + \log b$$

$$\log(a/b) = \log a - \log b$$

$$\log(a^n) = n \log a$$

The log base

(natural logs) $\log a = b \Rightarrow a = e^b$

(nats) $\log_{10} a = b_{10} \Rightarrow a = 10^{b_{10}}$

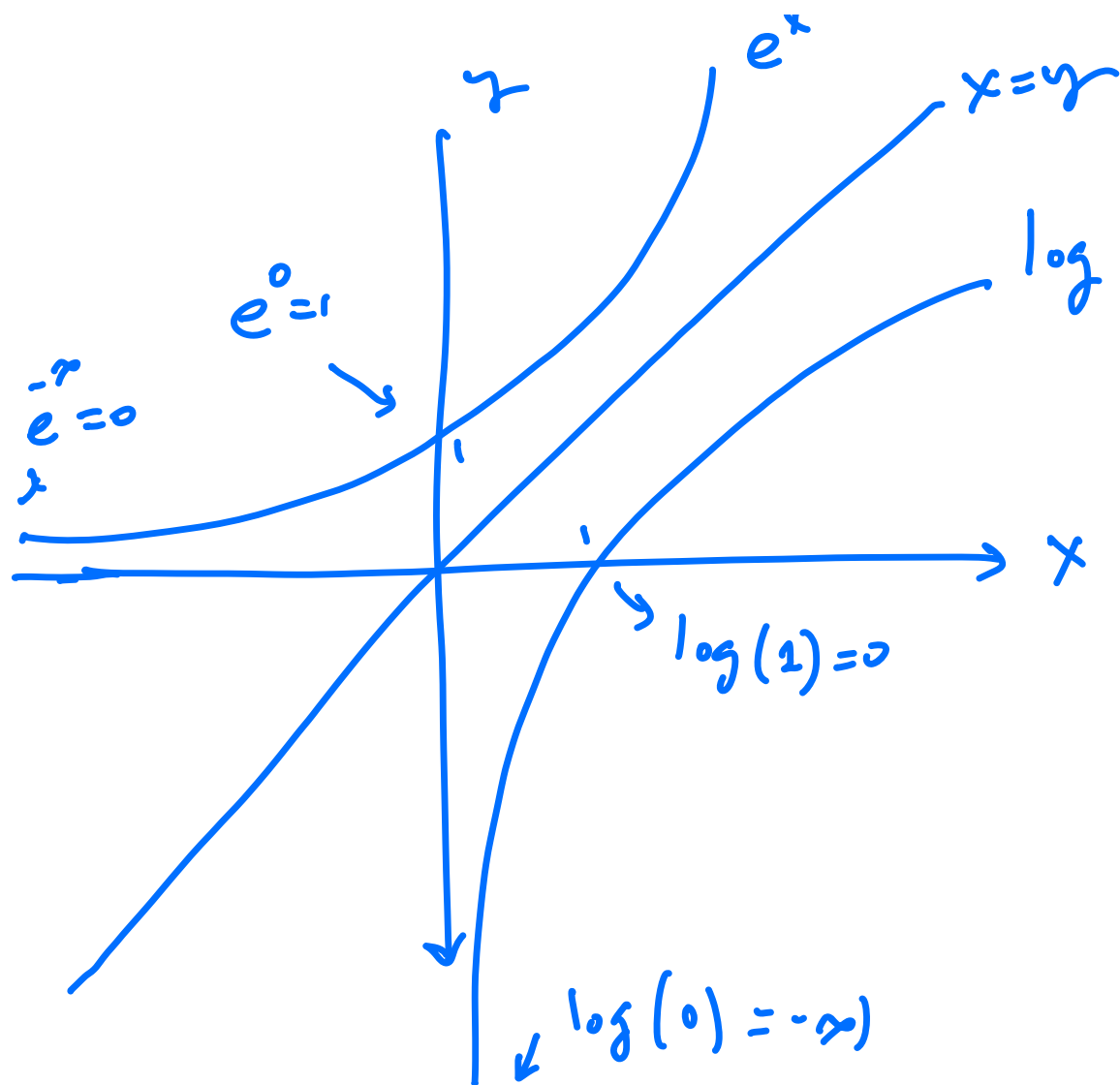
(bits) $\log_2 a = b_2 \Rightarrow a = 2^{b_2}$

$$e^b = 2^{b_2} \Rightarrow b = \log(2^{b_2})$$

$$= b_2 \log 2$$

$$b = b_2 \log 2$$

..



$$\frac{d}{dx} e^{ax} = a e^{ax}$$

$$\frac{d}{dx} \log(ax) = \frac{1}{ax} \cdot a = \frac{1}{x}$$

$$I(o, p) = -\log p$$

	p	I	bits
I had breakfast	1	$\log(1)$	0
My birthday	$\frac{364}{365}$	$\log \frac{364}{365}$	2.6
Not my birthday	$\frac{1}{365}$	$-\log 365$	8.5
1 st Cambridge $\frac{1}{2}$	$\frac{4500}{4.7 \cdot 10^3}$		10
1+2 " "	$\frac{4500}{4.7 \cdot 10^3} \frac{6500}{4.7 \cdot 10^3}$		13.5

Entropy Average Information
of a prob distribution.

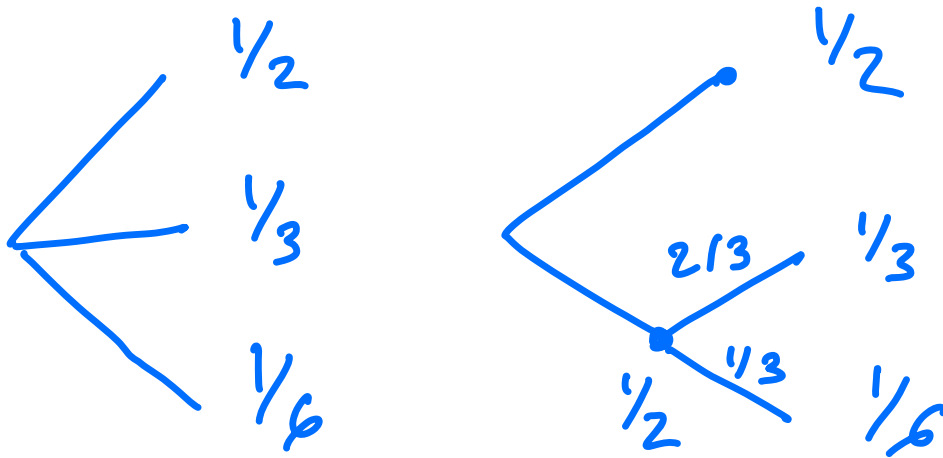
$$a \in X \quad I(a) = -\log P_X(a)$$

$$H(X) = \int_a P_X(a) \cdot \log \frac{1}{P_X(a)} da$$

$$= \sum_i P_i \log \frac{1}{P_i} = - \sum_i P_i \log P_i$$

$$H(X) \geq 0 \quad \text{''} \quad 0 \Leftrightarrow P_i = 0 \text{ except for 1 value}$$

Important property



$$H\left(\frac{1}{2}, \frac{1}{3}, \frac{1}{6}\right) = H\left(\frac{1}{2}, \frac{1}{2}\right) + \frac{1}{2} H\left(\frac{2}{3}, \frac{1}{3}\right)$$

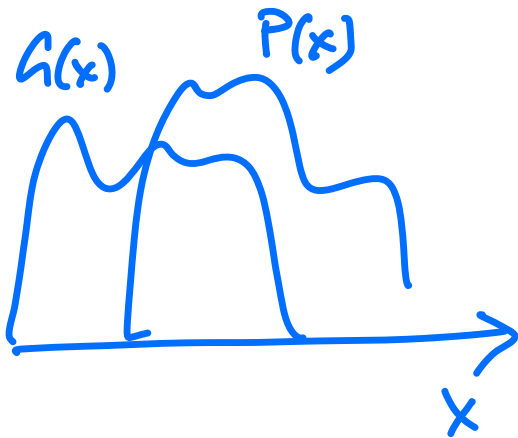
The composition Law

Same Information no matter how
the choices are broken

Relative entropy

the Kullback-Liebler divergence

To compare 2 probability distributions on X



$$D_{KL}(P \parallel Q) = \int_X P(x) \log \frac{P(x)}{Q(x)}$$

$$i) D_{KL}(P \parallel Q) \neq D_{KL}(Q \parallel P)$$

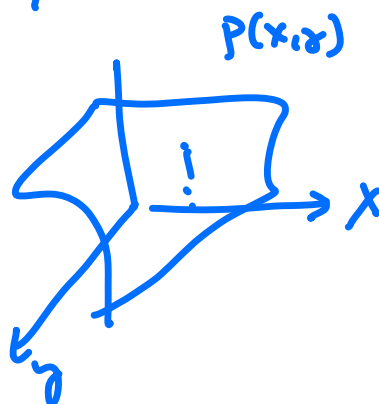
$$ii) D_{KL} \geq 0 \quad \Rightarrow \Leftrightarrow Q = P$$

Mutual Information

two random variables X, Y

$$P_{XY}$$

$$\left. \begin{matrix} P_X \\ P_Y \end{matrix} \right\} \text{ marginals}$$



$$P_X(x) = \int_y P(x, y) dy$$

$$P_Y(y) = \int_x P(x, y) dx$$

$$MI(X, Y) = D_{KL}(P_{XY} \parallel P_X P_Y) = \int_{xy} P(x, y) \log \frac{P(x, y)}{P(x)P(y)} dx dy$$

$$MI \geq 0$$

$$\text{if } X \perp Y \quad P(x, y) = P_X(x)P_Y(y)$$

$$MI = 0$$