

Lec. 2

Shanno's Code (attributed to Fano)

$$l_i = \left\lceil \log \frac{1}{p_i} \right\rceil$$

Check Kraft's inequality:

$$\sum_{i=1}^n \frac{1}{2^{l_i}} = \sum_{i=1}^n \frac{1}{2^{\left\lceil \log \frac{1}{p_i} \right\rceil}} \leq \sum p_i = 1.$$

Expected length of encoding of a symbol:

$$\begin{aligned} &= \sum_{i=1}^n p_i \cdot \left\lceil \log \frac{1}{p_i} \right\rceil \geq \sum p_i \log \frac{1}{p_i} = H(X). \\ &\quad \searrow \\ &\leq \sum p_i \left(\log \frac{1}{p_i} + 1 \right) = H(X) + 1. \end{aligned}$$

The worst-case length can be bad:

e.g., for $X = \begin{cases} 0 & \text{w.p. } 1 - 10^{-4.5} \\ 1 & \text{w.p. } 10^{-5} \end{cases}$

But Huffman is better.

Huffman vs. Shannon:

Huffman was better in the last example

<u>but:</u>	a	$\frac{1}{3}$	2	1
	b	$\frac{1}{3}$	2	2
	c	$\frac{1}{4}$	(2)	(3)
	d	$\frac{1}{12}$	4	3
			Huff Shannon # bits	Shannon Huffman # bits

Lower bound:

Thm. The expected length of encoding by a prefix-free code of a r.v. X is $\geq H(X)$. $\{l_i\}$

Pf.

$$H(X) = - \sum_{i=1}^n P_i \log \frac{1}{P_i}$$

$$= \sum P_i \log \left(\frac{1}{P_i 2^{l_i}} \right) \leq \log \left(\sum P_i \frac{1}{P_i 2^{l_i}} \right)$$

Jensen's

Kraft's

$$\leq \log 1 = 0. \quad \square$$

Improving +1's

Assume the source outputs a sequence of iid draws from X .

$$X_1, \dots, X_m$$

Fact. $H(X_1, \dots, X_m) = m \cdot H(X)$.

Shannon-Fano Code of $X_1, \dots, X_m$
 $\Downarrow$

$$\mathbb{E} \text{ length} \leq m H(X) + 1$$

$$\Rightarrow \text{avg. \# bits used per symbol} \leq H(X) + \frac{1}{m} \checkmark$$

$\Rightarrow$ Thm (Fundamental Thm of Source Coding or Noiseless Coding Thm):

$\forall \epsilon > 0, \exists n$ s.t.

Given iid samples from a r.v. X ,
it's possible to communicate (on avg.)

$$\leq H(X) + \epsilon \text{ bits/sample to reveal } X_1, \dots, X_n \text{ to the other party.}$$

Def. Let X, Y be two (possibly correlated) r.v.'s, Then

$$H(X, Y) := \sum_{x, y \leftarrow (X, Y)} \underbrace{p(x, y)}_{\substack{\downarrow \\ \Pr(X=x, Y=y)}} \log \frac{1}{p(x, y)}.$$

If X, Y are independent $\Rightarrow p(x, y) = p(x)p(y)$

$$\Rightarrow H(X, Y) = H(X) + H(Y).$$

In general, $p(x, y) = p(x)p(y|x)$

$$\Rightarrow H(X, Y) = \sum_{x, y} p(x, y) \log \frac{1}{p(x)p(y|x)}$$

$$= \underbrace{\sum p(x, y) \log \frac{1}{p(x)}}_{H(X)} + \sum_{x, y} p(x, y) \log \frac{1}{p(y|x)}$$

$$= \sum_x p(x) \sum_y p(y|x) \log \frac{1}{p(y|x)}$$

$$= H(X) + \sum_x p(x) \cdot H(Y|X=x)$$

$$\text{Def} = H(X) + H(Y|X).$$

$$\Rightarrow \text{Lemma: } H(X, Y) = H(X) + H(Y|X).$$

$$\text{If } X, Y \text{ are independent: } H(X, Y) = H(X) + H(Y).$$

$$\& \quad H(Y|X) = H(Y).$$

Thm (Chain rule):

(for any ordering)

$$H(X_1, \dots, X_n) = H(X_1) + H(X_2|X_1) + \dots +$$

$$H(X_n | X_1, \dots, X_{n-1}).$$

example: $H(X|X) = 0$

$$X \text{ uniform on } \{0, 1, 2, 3\} \Rightarrow H(X) = 2$$

$$(Y = X \bmod 2) = \{0, 1\} \Rightarrow H(Y) = 1$$

$$H(X|Y) = 1.$$

$$H(Y|X) = 0.$$

$$H(X, Y) = 2.$$

Lemma. $H(y|x) \leq H(y)$.

Proof. $H(y|x) - H(y) = H(x, y) - H(x)$
 $\quad\quad\quad - H(y)$

$$= \sum p(x, y) \log \frac{1}{p(x, y)}$$

$$- \underbrace{\sum_x p(x)}_{=\sum_{x,y} p(x,y)} \log \frac{1}{p(x)} - \underbrace{\sum_y p(y)}_{=\sum_{x,y} p(x,y)} \log \frac{1}{p(y)}$$

$$= \sum_{x,y} p(x, y) \log \frac{p(x)p(y)}{p(x, y)}$$

$$= \mathbb{E}_{p(x,y)} \log Z \leq \log \mathbb{E} Z = 0.$$

$\sum p(x)p(y) = 1$

Cor. ("Union bound"):

$$H(x, y) \leq H(x) + H(y).$$

$$H(x_1, \dots, x_n) \leq \sum_{i=1}^n H(x_i)$$

* Exercise. (follows from "conditioning cannot increase entropy")

$H(\underline{p})$ is a concave function of the distribution.

$$\text{i.e., } \lambda H(\underline{p}) + (1-\lambda) H(\underline{q}) \leq H(\underbrace{\lambda \underline{p} + (1-\lambda) \underline{q}}_{\text{take it as some r.v.}}).$$

Mutual Information.

Def. $I(X; Y) := H(X) - H(X|Y)$

(Reduction in uncertainty of X by virtue of knowing Y)

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

$$\Rightarrow I(X; Y) \geq 0, \quad I(X; Y) = I(Y; X)$$

If X, Y are independent,

$$I(X; Y) = 0.$$

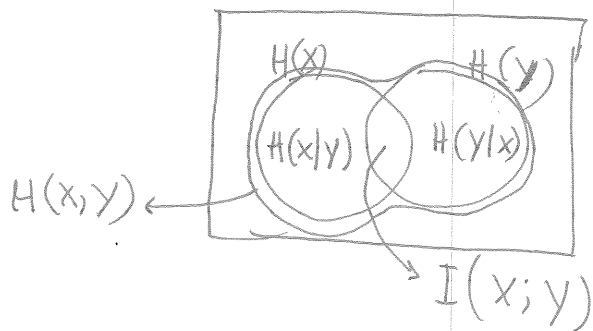
For the $\{0, \dots, 3\}$ example:

$$I(X; Y) = 1.$$

Or, Z_i iid $\{0, 1\}$.

$$X = Z_1, \dots, Z_5, \quad Y = Z_4, Z_5, Z_6, Z_7$$

$$\Rightarrow I(X; Y) = 2.$$



Conditional mutual Information.

$$I(X; Y | Z) \stackrel{\text{def}}{=} H(X|Z) - H(X|Y, Z) \\ = H(Y|Z) - H(Y|X, Z)$$

Exercise: (Chain rule)

$$I(X_1, \dots, X_n; Y) = \sum_{i=1}^n I(X_i; Y | X_1, \dots, X_{i-1}) \\ = I(X_1; Y) + I(X_2; Y | X_1) + \dots + I(X_n; Y | X_1, \dots, X_{n-1})$$

Exercise: Is it always the case that $I(X; Y | Z) \leq I(X; Y)$?