

TTIC 31200 / CMSC 37220

Information & Coding Theory

- TA: Kavya Ravichandran
- Discussion: F 2-3 pm
- Office hours: Th 12:30 - 1:30
- See course page for notes/resources
- 4 Homeworks (60%) + Take-home final (40%)
- Pre-reqs: probability, basic analysis, linear algebra

Contents

Information Theory

≈ 75%

(mathematical perspective)

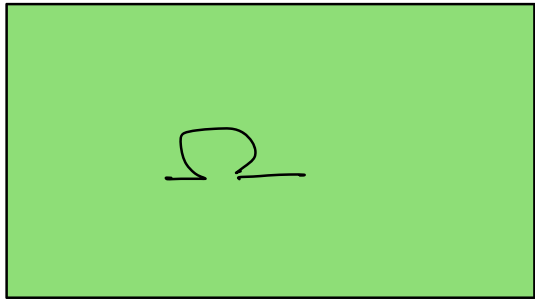
Coding Theory

≈ 25%

(error correction)

- basic concepts + applications
- statistics
- error correcting codes
- TCS applications / miscellaneous topics / quantum information

Recap: Random Variables & Distributions



(discrete) probability space Ω

measure $\mu : \left\{ \begin{array}{c} \text{subsets of} \\ \Omega \end{array} \right\} \rightarrow [0, 1]$

Random variable

$$X : \Omega \rightarrow \mathcal{R}$$

set of
"values" for X

Distribution

$P(X)$ "Event"
Subset of Ω

eg.

$$P[X = a] = 0.3$$
$$P[X = b] = 0.7$$

Notation

$X, Y, Z \dots$ random variables

$x, y, z \dots$ values

$\mathcal{X}, \mathcal{Y}, \mathcal{Z} \dots$ sets of possible values (support)

$P, Q \dots$ distributions

$p, q \dots$ actual probabilities

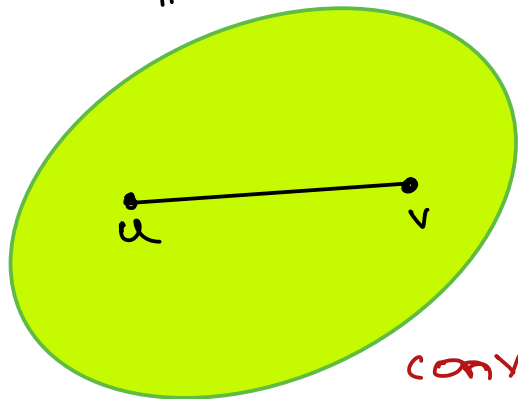
$P[\text{event}] \dots$

$$E[X] = \sum_{a \in \mathcal{X}} a \cdot P[X=a]$$

When well-defined.

Convexity

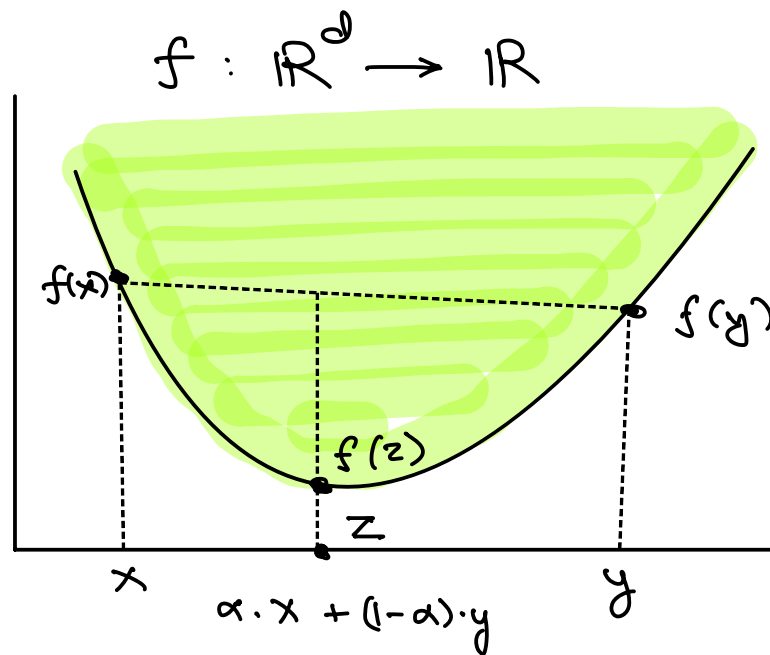
$$S \subset \mathbb{R}^d$$



convex combination

$$u, v \in S \Rightarrow \alpha \cdot u + (1-\alpha) \cdot v \in S$$

$$\forall \alpha \in [0, 1]$$



$$f(z) \leq \alpha \cdot f(x) + (1-\alpha) \cdot f(y)$$
$$\downarrow$$
$$f(\alpha \cdot x + (1-\alpha) \cdot y)$$

g is concave $\Leftrightarrow -g$ is convex

Jensen's inequality

Convex set S , $f: S \rightarrow \mathbb{R}$

random variable X
supported in S

$$f \text{ is convex} \Rightarrow \mathbb{E}[f(X)] \geq f(\mathbb{E}[X])$$

$$f \text{ is concave} \Rightarrow \mathbb{E}[f(X)] \leq f(\mathbb{E}[X])$$

Exercises

- Prove Jensen's inequality when support of X is finite (induction)

- Prove $f(x) = x^2$, $f(x) = x \log x \rightarrow$ convex

$f(x) = \log x \rightarrow$ concave

- Prove Cauchy-Schwarz using Jensen's

$$|\langle u, v \rangle| \leq \|u\|_2 \cdot \|v\|_2$$

Entropy

$$\mathcal{X} = \{ \underbrace{a_1, \dots, a_s}_{p_1 \dots p_s} \}$$

bits needed to provide certainty (specify one value)

= ?

Entropy

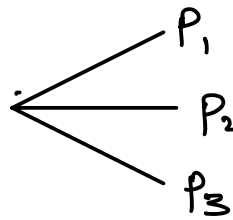
$$\mathcal{X} = \{a_1, \dots, a_n\}$$

$$p_1 \dots \dots \dots p_n$$

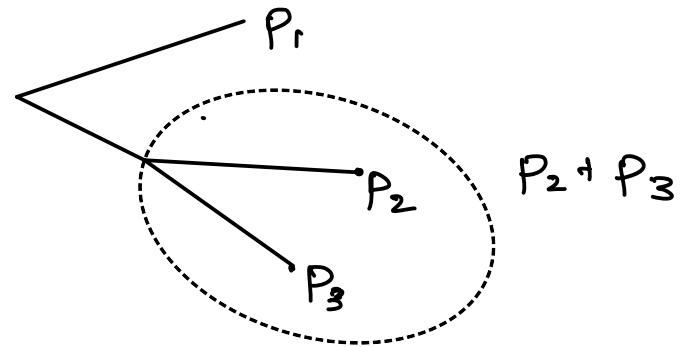
$$P[X = a_i] = p_i$$

- Continuous function of p_1, \dots, p_n
- When $p_1 = \dots = p_n = 1/n$. increasing in n

- Bundling:



\equiv



Entropy

$$P[X = a_i] = p_i$$

$$H(X) = \sum_{i=1}^n p_i \cdot \log \frac{1}{p_i} \quad \log_2$$

$$\frac{\partial \log \frac{1}{p}}{\partial p} = 0$$

$$p_1 = p_2 = \dots = p_n = \frac{1}{n}$$

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

$$\blacktriangleright 0 \leq H(X) \leq \log |X| = \log n$$

Proof: $p_1 \dots p_n$ $H(X) = \sum p_i \cdot \log \frac{1}{p_i}$

$\underbrace{}_{\geq 0} \cdot \underbrace{\log \frac{1}{p_i}}_{\geq 0}$

$$H(X) = \sum_i p_i \cdot \log \frac{1}{p_i}$$

$$= \mathbb{E} [\log Y]$$

for $Y = \frac{1}{p_i}$ wrp. p_i

(Jensen's)

$$\leq \log(\mathbb{E} Y)$$

$$= \log\left(\sum p_i \cdot \frac{1}{p_i}\right) = \log n$$

Source Coding

Prefix free code $C: X \rightarrow \Sigma^*$ s.t. $\forall x \neq y$
finite alphabet
 $C(x) \neq C(y) \circ \sigma$
for any $\sigma \in \Sigma^*$

Fix $\Sigma = \{0,1\}$ (say)

Goal: Prefix-free C with least *expected length*

$$E |C(X)| = \sum_{x \in \mathcal{X}} p_x [|C(x)|]$$

Kraft's inequality

Let $|X| = n$. \exists prefix-free $C: X \rightarrow \{0,1\}^*$ with lengths
 $l_1 \dots \dots l_n$

if and only if $\sum_{i=1}^n \frac{1}{2^{l_i}} \leq 1$

"if" given $l_1 \dots l_n$ s.t. $\sum \frac{1}{2^{l_i}} \leq 1 \exists$ code

"only if" for all prefix-free $C \sum \frac{1}{2^{l_i}} \leq 1$

(Assuming Kraft's inequality)

▶ Let C be any prefix-free code $C: X \rightarrow \{0,1\}^*$.

Then $\mathbb{E}|C(x)| \geq H(x)$

Proof:

$$H(x) - \mathbb{E}|C(x)|$$

$$= \sum p_i \log \frac{1}{p_i} - \sum p_i l_i$$

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

$$= \mathbb{E}[\log Y]$$

$$Y = \frac{1}{p_i \cdot 2^{l_i}} \text{ w.p. } p_i$$

$$\leq \log(\mathbb{E}Y) = \log\left(\sum p_i \frac{1}{p_i \cdot 2^{l_i}}\right) = \log\left(\sum \frac{1}{2^{l_i}}\right) \leq \log 1 = 0$$

Shannon code

▶ There exists prefix-free $C: \mathcal{X} \rightarrow \{0,1\}^*$

with $\mathbb{E} |C(x)| \leq H(x) + 1$

Proof: Construct code with $l_i = \lceil \log \frac{1}{p_i} \rceil$

$$\sum \frac{1}{2^{l_i}}$$

$$= \sum \frac{1}{2^{\lceil \log \frac{1}{p_i} \rceil}}$$

$$\leq \sum p_i \leq 1$$

$\therefore \exists$ prefix-free code

$$l_i \leq \log \frac{1}{p_i} + 1$$

$$\therefore \mathbb{E} |C(x)| = \sum p_i l_i \leq H(x) + 1$$

Kraft's inequality

Let $|X| = n$. \exists prefix-free $C: X \rightarrow \{0,1\}^*$ with lengths
 $l_1 \dots \dots l_n$

if and only if $\sum_{i=1}^n \frac{1}{2^{l_i}} \leq 1$

"if" given $l_1 \dots l_n$ s.t. $\sum \frac{1}{2^{l_i}} \leq 1 \exists$ code

"only if" for all prefix-free $C \sum \frac{1}{2^{l_i}} \leq 1$

"only if" for all prefix-free $C \quad \sum \frac{1}{2^{\ell_i}} \leq 1$

Proof: Bingo!

Bingo tickets with codewords $C(x_1) \dots C(x_n)$

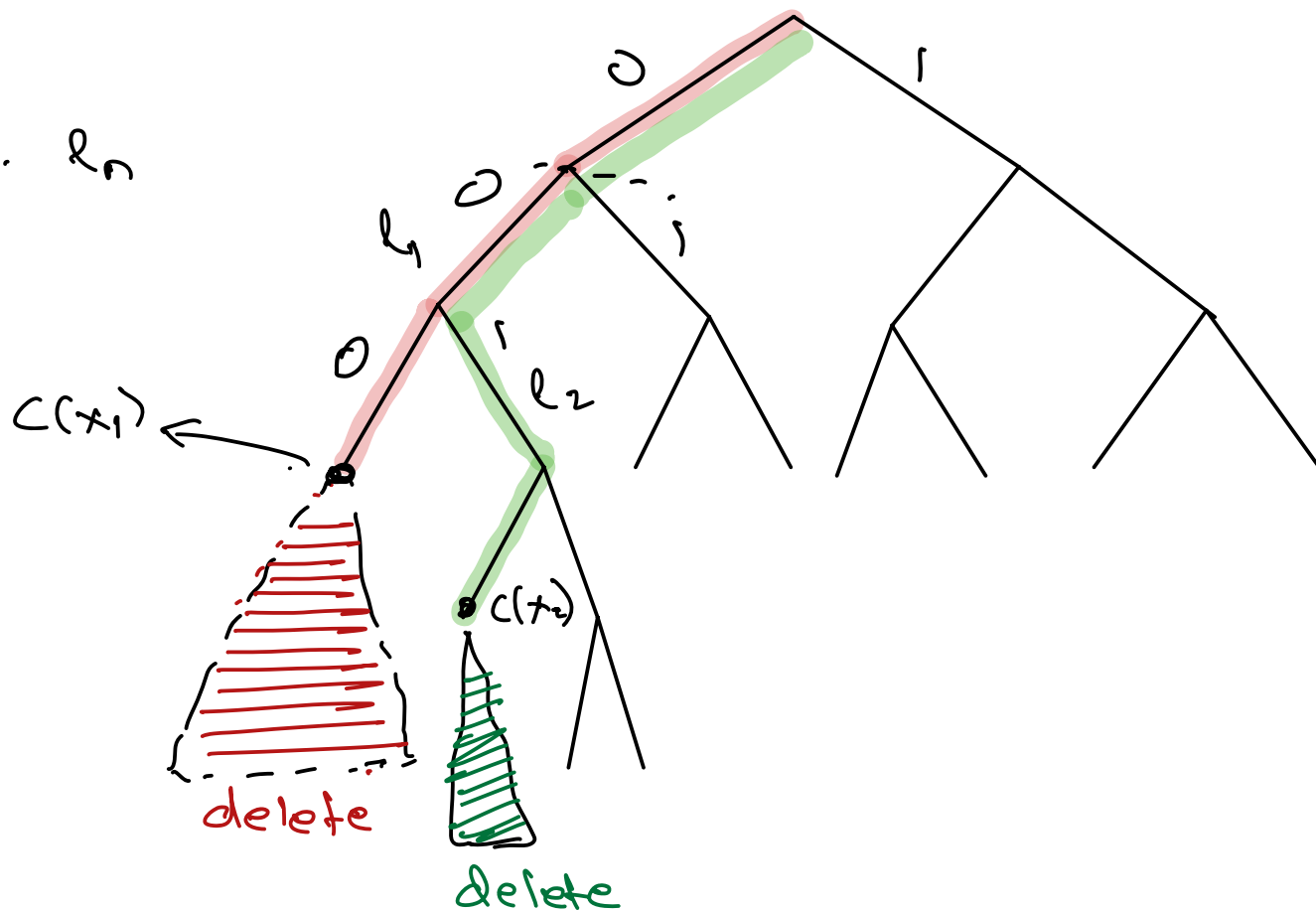
$P[\textit{i}^{\text{th}} \text{ ticket wins}] = \frac{1}{2^{\ell_i}}$ (need prefix-free)

disjoint events

$$\therefore \sum \frac{1}{2^{\ell_i}} \leq 1$$

"if" given l_1, \dots, l_n s.t. $\sum \frac{1}{2^{l_i}} \leq 1 \exists$ code

Sort l_1, \dots, l_n



- Can always pick next codeword if leaves left
- Deleted fraction of leaves $\frac{1}{2^{l_1}} + \dots + \frac{1}{2^{l_{i-1}}} < 1 \forall i$