

Lecture 20 - Information Theory

①

[Scribe - Cover-Thomas, Braverman survey...]

[Shannon '48] [Invented everything. Applications in coding theory, combinatorics, comm. theory...]

Let X be a (discrete) random variable. E.g. $X = \begin{cases} \text{red} & \text{w.p. } 1/2 \\ \text{green} & 1/4 \\ \text{blue} & 1/8 \\ \text{yellow} & 1/8 \end{cases}$

Entropy of X , denoted $H(X)$, is - intuitively -

- "amount of randomness in X " [in bits]
- ~~min~~ # of rand bits needed to generate a draw of X , on average

e.g.: $H \rightarrow \text{red}$

$T \rightarrow H \rightarrow \text{green}$

$T \rightarrow H \rightarrow \text{blue}$

$T \rightarrow \text{yellow}$

: w.p. $\frac{1}{2}$ used 1

$\frac{1}{4}$ " 2

$\frac{1}{8}$ " 3

$\frac{1}{8}$ " 3

$= \frac{14}{8} = 1.75$

- # of bits needed to store a draw from X using best compression scheme.

on avg.: ~~red~~ red $\leftrightarrow H$
green $\leftrightarrow TH$
blue $\leftrightarrow TTH$
yellow $\leftrightarrow TTT$

Rem: "prefix-free" encoding [no codeword is prefix of another]

$\Rightarrow$ can store n draws w/o "punctuation"

$\hookrightarrow THTHHH = \text{green, blue, red, red}$

- # of bits needed (on avg.) for

Alice to communicate one draw from X to Bob

[storing on a hard drive is just like transmitting (coming to your future self)]

- # of YES/NO q's Bob needs to ask (on avg.) to learn a draw of X .

[20 questions]

"Is it red? Green? Blue?" Kinda trv...]

All exactly correct if X 's probabilities are 2^{-k} , $k \in \mathbb{N}$. def: $H(X) = \sum_{x \in \text{range}(X)} p_X(x) \log_2 \frac{1}{p_X(x)}$

But...

$$X = \begin{cases} a & \text{w.p. } \frac{1}{2} \\ b & \text{w.p. } \frac{1}{3} \\ c & \text{w.p. } \frac{1}{6} \end{cases}$$

Communicating X : w.p. $\frac{1}{2} \rightarrow 1$ bit.

w.p. $\frac{1}{3} \rightarrow \log_2 3$ bits ?? (not an integer)

$\frac{1}{6} \rightarrow \log_2 6$

Elegant to just ~~define~~ $H(X)$ thus anyway, ~~skip down~~

fact: If $|\text{range}(X)| \leq N$, $0 \leq H(X) \leq \log_2 N$ eq. iff X is univ. dist. for N outcomes

eq. iff X const.

Takes " $\log_2 N$ " bits to gen. univ. dist. on $[N]$

pf: exercise using Jensen!

[How can we get around this slightly annoying gap between $H(X)$ and the actual compression/comm./generation cost? Amortization]

trad. to write as a string: XY

Say X, Y are indep. r.v.'s. The pair (X, Y) can be viewed as a single r.v.

fact: $H(\underbrace{XY}_{X,Y}) = H(X) + H(Y)$. [To optimally gen. a draw of (X, Y) , optimally do it for X , and Y . No savings poss. if indep.]

[one time we'll do a proof; rest of the follow your intuition]

$$\begin{aligned} \text{pf: } H(XY) &= \sum_{(x,y) \in \text{range}(X,Y)} p_{XY}(x,y) \log_2 \frac{1}{p_{XY}(x,y)} \\ &= \sum_x \sum_y \underbrace{p_X(x) p_Y(y)}_{\text{indep.}} \left(\log_2 \frac{1}{p_X(x)} + \log_2 \frac{1}{p_Y(y)} \right) = H(X) \sum_y p_Y(y) + H(Y) \sum_x p_X(x) \quad \square \end{aligned}$$

[*] Aside: can store/gen. using $\sum_x p(x) \lceil \log_2 \frac{1}{p(x)} \rceil$ bits: $\leq \sum p(x) (\log_2 \frac{1}{p(x)} + 1) = H(X) + 1$

$\log_2$ ~~and~~ $H \rightarrow a$

$\lceil \log_2 \rceil$ $TH \rightarrow b$

$\lceil \log_2 \rceil$ $TTT \rightarrow c$

(ex: even prefix-free!)

"Shannon-Fano code"

[not optimal - "Huffman code" is]

[Instantaneous conversion of coin flips to draws...

TTT is "wasted"!

[Of course you can't beat $H(X)$, either. Ex via Kraft...]

∴ if $X_1, \dots, X_n$ are indep copies of X , $H(X_1 \dots X_n) = n H(X)$

To communicate/generate, could use $(H(X)+1) \cdot n = nH(X) + n$ bits,
but wasteful. Eg: if $X = \begin{cases} 0 & \text{w.p. } 1/3 \\ 1 & \text{w.p. } 2/3 \end{cases}$, $H(X) = \log_2 3 \approx 1.6$,
uses 2.6n bits

But also, could use $H(X_1 X_2 \dots X_n) + 1$ bits $\parallel$ It's like, interp as # betw. 0 & 3 $\rightarrow$
 $\parallel$ send it in binary. $\parallel$
Shannon - Fano.
 $\parallel$
 $nH(X) + 1$
 $= H(X) + \frac{1}{n}$ amortized bits per draw.

$\hookrightarrow 0$ as $n \rightarrow \infty$.

∴ $H(X) =$ amortized $\parallel$ over many indep. copies $\parallel$ bits to gen/comm. draw of X

Mutual Information: Say X, Y are possibly dep. r.v.'s
How many rand bits to generate? ("Ideally"):
 $H(XY)$. If X, Y indep., it's $H(X) + H(Y)$.
If not, can get savings.

def. $\rightarrow I(X; Y) =$ savings when gen/comm. X, Y jointly
 $= H(X) + H(Y) - H(XY)$ $(= \sum_{x,y} p(x,y) (\log \frac{1}{p(x)} + \log \frac{1}{p(y)} - \log \frac{1}{p(x,y)}))$
 $\log \frac{p(x,y)}{p(x)p(y)}$

eg:

X	Y	prob
*	*	$\frac{1}{2}$
0	a	$\frac{1}{16}$
0	b	$\frac{1}{16}$
0	c	
0	d	
1	a	
1	b	
1	c	
1	d	$\frac{1}{16}$

$$H(X) = H\left(\begin{matrix} * & \text{w.p. } 1/2 \\ 0 & \text{w.p. } 1/4 \\ 1 & \text{w.p. } 1/4 \end{matrix}\right) = \frac{1}{2} \cdot 1 + \frac{1}{2} \cdot 2 = 1.5,$$

$$H(Y) = \frac{1}{2} \cdot 1 + \frac{1}{2} \cdot 3 = 2.$$

$\parallel$ not indep, b/c $X=* \Rightarrow Y=* \parallel$ $H(X) + H(Y) = 3.5$

$$H(X, Y) = \frac{1}{2} \cdot 1 + \frac{1}{2} \cdot 4 = 2.5.$$

$$\therefore I(X, Y) = 3.5 - 2.5 = 1.$$

$\hookrightarrow$ # shared bits used to gen. them / reduce to indep.
generations $\parallel$ again, amortized over many I .

Facts: $0 \leq I(X;Y) \leq H(X) + H(Y)$

↑
eq. if indep.

↑
eq. if both const, $H=0!$

↑
better $\leq \min \{H(X), H(Y)\}$ [most you could save is everything]

Equality $I(X;Y) = H(Y) \Leftrightarrow Y$ is determined by X (a for of)

eg: $I(X;X) = H(X)$

OR: Let X = rand n bit string

Y = parity of all X 's bits

$H(X) = n$ $H(Y) = 1$ $I(X;Y) = 1$

[Y let by X . Savings is 1]

Though mutual info. is nice & symmetric in X, Y , sometimes you need to look @ things asymmetrically.

Say you need to generate X, Y together. You could be like, "first I'll gen X ; then I'll see how many more bits I need to gen Y "

$H(XY) = H(X) + \# \text{ bits (on avg) needed to gen } Y \text{ given } X$

def: $H(Y|X)$ "condit. entropy of Y given X "

$= H(XY) - H(X)$

(please remember, tho it starts w/ H , it's an expect, $\mathbb{E}$)
 $= \mathbb{E} [H(Y|X=x)]$

Fact: $H(Y|X) \geq 0$ w/ eq. iff Y is determined by X

[You gotta gen X to gen XY .
Formally, need Jensen.]

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

$$(Y|X)H - (Y|X) = - (Y|X)H - (Y|X) = -2(Y|X)H$$

1. $\frac{d}{dt} \ln p(x_t) = -\frac{1}{p(x_t)} \frac{dp(x_t)}{dt}$
2. $\frac{d}{dt} \ln p(x_t) = -\frac{1}{p(x_t)} \frac{dp(x_t)}{dt}$
3. $\frac{d}{dt} \ln p(x_t) = -\frac{1}{p(x_t)} \frac{dp(x_t)}{dt}$
4. $\frac{d}{dt} \ln p(x_t) = -\frac{1}{p(x_t)} \frac{dp(x_t)}{dt}$

[[One more interp.]] : $I(X; Y) = \underbrace{H(Y) - H(Y|X)}_{\text{how much info you gain about } X \text{ by knowing } Y} \stackrel{\text{symmetry}}{=} \underbrace{H(X) - H(X|Y)}$

A noisy channel B

$X^{(n)}$ corrupts $Y^{(n)}$

$I(X; Y)$ is info gained by Bob about X.

"channel capacity" [[Shannon "operationalized"; showed there's a messaging scheme which lets you "transmit info @ this rate"!!]]

Say X = rand n -bit string, Y = (parity of X 's bits, 1 more indep rand bit)

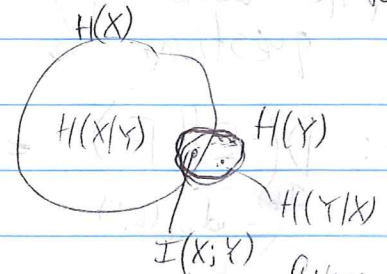
(5)

~~mutual info~~ Say you're jointly generating X, Y . Could first spend the "common part" $I(X; Y)$ bits. Then you still have to do the deficit from $H(X)$ to make X , deficit from Y to make Y .

[In example, first gen $\oplus X$: 1 bit. Then I have to make Y , 1-1 more for X]

def: $H(Y|X) = H(Y) - I(X; Y) \geq 0$

Conditional entropy of Y given X . equiv. [given "mutual info of X, Y "]



[Venn diag misleading if 3+ rvs!]

[given I 's formula $\Rightarrow$]

formula: $H(Y|X) = H(XY) - H(X)$

$\Leftrightarrow$ "chainrule": $H(XY) = H(X) + H(Y|X)$

[To gen X, Y , first gotta gen X , then gen "Y given X"]

fact: [Notation is very suggestive/bad. Tho it starts with an H , $H(Y|X)$ really an E]

$H(Y|X) = E_{x \sim X} [H(Y|X=x)]$

$$\begin{aligned} pfs: H(XY) - H(X) &= E_{(x,y) \sim XY} \left[\log \frac{1}{p(x,y)} \right] - E_{x \sim X} \left[\log \frac{1}{p(x)} \right] \\ &= E_{x,y \sim XY} \left[\log \frac{1}{p(x)p(y|x)} - \log \frac{1}{p(x)} \right] \\ &= E_{x \sim X} \left[E_{y \sim Y|X=x} \left[\log \frac{1}{p(y|x)} \right] \right]. \quad \square \end{aligned}$$

def: condit. mut. info:
 $I(X; Y|Z)$
 $= H(X|Z) + H(Y|Z) - H(XY|Z)$

[Are more def, then some apps.]

(*) GOTO PAGE (5.5)

KL-Divergence [Kullback-Leibler, AKA relative entropy]

[Let E denote] rand. letter fr. English [novel]

$H(E) = \# \text{ yes no questions to guess}$

≈ 4 [in ~~over~~ Latin alpha langs]

[1st q: is it in {CHUMP GIFTS?}]

|| Pierre comes to America and wants to play French Scrabble expert, doesn't know English. Take a shot with his French 20 g's strat. ||

F: rand French letter,

|| 1st Q: en "PEU MOL"? ||

|| He would pay more than $H(E)$ questions on average. What is the "penalty"? || "a little bit soft" ||

def: $D(E||F)$ = avg. # extra bits needed when storing/guessing draw fr. E using scheme for F. || E.g. long time for Pierre to guess w ||

$D(E||F) = D_{KL}(E, F)$

$$D(X||Y) = E_{x \sim X} \left[\log \frac{1}{p_Y(x)} - \log \frac{1}{p_X(x)} \right]$$

|| E.g. excess per bit if you zip an English novel using encoder designed for French ||

fact: Let X, Y be joint r.v.s, let $X \times Y$ denote drawing them indep.

$D(XY || X \times Y) = I(X; Y)$ || penalty for not using savings ||

rem: $D(E||F)$ like a distance between E, F , but NOT SYMMETRIC!

$\neq D(F||E)$

E.g.: Hawaiian has 12 letters || i.e. actually 13, one being "glottal stop, never mind" ||

$D(H||E) \approx 2$, modest

$D(E||H) = \infty$

Rec. 8: || moderately annoying to prove || || Gibbs-Su ||

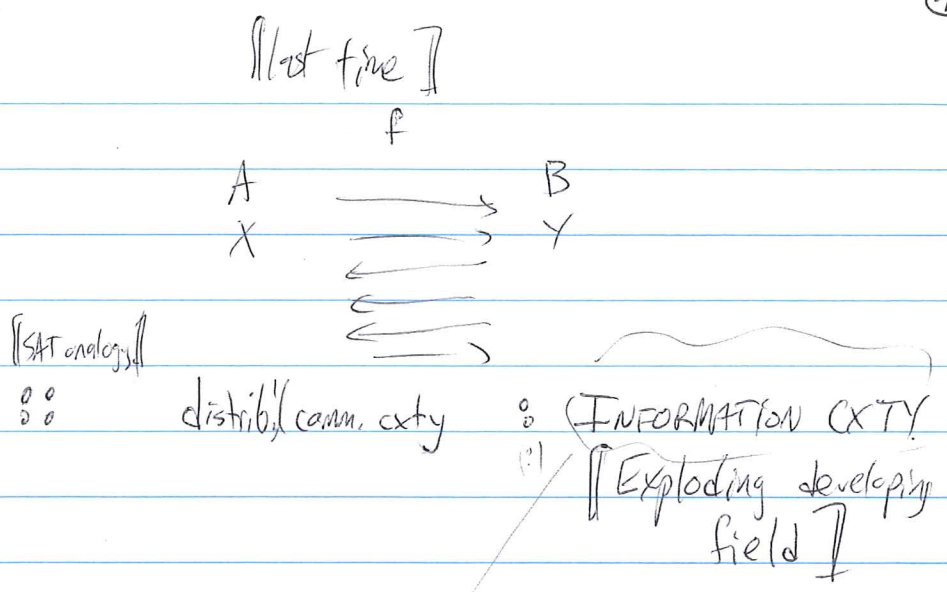
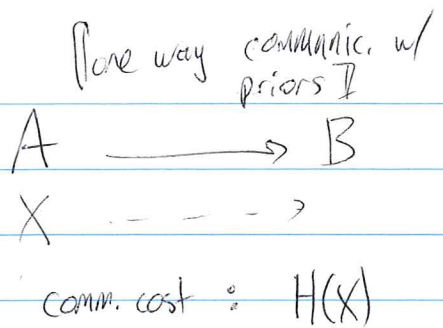
$d_{TV}(X, Y) \leq \sqrt{\frac{1}{2} d_{KL}(X, Y)}$

$d_H(X, Y)^2$

$\leq d_H(X, Y) \leq$

Hellinger

Can be sharp: $D(\text{Bern}(\frac{1}{2} + \epsilon) || \text{Bern}(\frac{1}{2})) = \Theta(\epsilon^2)$



$\propto$ amt of "info" A, B need to exchange to compute $f(x, y)$ [maybe with error]

def. Let Π be a (bit) comm prot.

$$IC_{\mu_{XY}}(\Pi) := I(\Pi; Y | X) + I(\Pi; X | Y)$$

$\uparrow$ amt. of info conversation teaches Alice abt. Bob's input
 $\uparrow$ vice versa.

$$IC_{\mu_{XY}}^{\epsilon}(f) = \inf_{\Pi} \{ IC_{\mu_{XY}}(\Pi) : \Pr[\Pi(x, y) \neq f(x, y)] \leq \epsilon \}$$

fact: $D_{\mu}^{\epsilon}(f) \geq IC_{\mu}^{\epsilon}(f)$ [Converse??]

thm [BYDKS04]: $I_{\mu_{XY}}^{\epsilon}(\text{DIST}_n) \geq n \cdot I_{\mu_{XY}}^{\epsilon}(\text{NAND}) \Rightarrow R_{\epsilon}(\text{DIST}_n) \geq \Omega(n)$

$\uparrow$ appropriate $\uparrow$ NAND(x, y) $\uparrow$ Yes! $\Omega(\epsilon^{\frac{1}{2}})$, tho slightly tough.

thm: $IC_{\mu^n}^{\epsilon}(f^n) = n \cdot IC_{\mu}^{\epsilon}(f)$

$\Rightarrow$ amortized comm cty, $\lim_{n \rightarrow \infty} \frac{IC_{\mu^n}^{\epsilon}(f^n)}{n} \geq IC_{\mu}^{\epsilon}(f)$

must get ϵ/n error on each coord.

thm [BR10]:

Proof requires compression of conversation: convo w/ little info compressible to short one.

Many open probs still

