

## 1.5 Conditional versions of basic properties

Def. •  $X, Y, Z, X_i$  disc. r.v.

•  $E$  some event

•  $f$  a deterministic function

•  $\text{range}(X|E)$  = set of values  $X|E$  takes

• Maximality  $H(X|E) \leq \log |\text{range}(X|E)|$

"=" holds iff  $X|E \sim \text{range}(X|E)$

• Chain rule  $H(X_1, \dots, X_n | Y) = H(X_1 | Y) + H(X_2 | X_1, Y) + \dots + H(X_n | X_1, \dots, X_{n-1}, Y)$

• Monotonicity • if  $X = f(Y)$ , then  $H(X) \leq H(Y)$

• if  $X = f(Y, Z) \Rightarrow H(X|Z) \leq H(Y|Z)$

• if  $Y = f(Z) \Rightarrow H(X|Z) \leq H(X|Y)$

• Dropping Conditioning  $H(Z|Y, X) \leq H(Z|Y)$

• Subadditivity  $H(X_1, \dots, X_n | Y)$

$\leq H(X_1|Y) + H(X_2|Y) + \dots + H(X_n|Y)$

• Shearer's lemma Given random  $F \subseteq [n]$  w.

$\Pr(i \in F) \geq \mu$ , for  $\forall i \in [n]$ . then

$$H(X_1, \dots, X_n | Y) \leq \frac{1}{\mu} \mathbb{E}_F H(X_F | Y)$$

• Gibbs inequality  $X, Y$  take value over same set

$A$ , write  $p_a = \Pr(X=a)$ ,  $q_a = \Pr(Y=a)$

$$\Rightarrow H(X) = \sum_{a \in A} p_a \log \frac{1}{p_a} \leq \sum_{a \in A} p_a \log \frac{1}{q_a}$$

where "=" holds iff  $p_a = q_a \quad \forall a \in A$ .

Equivalently, Kullback-Leibler divergence or relative entropy is non-negative:

$$D_{KL}(X \| Y) = \sum_{a \in A} p_a \log \frac{p_a}{q_a} \geq 0$$

## § 2 Applications

### 2.1 Volume of Hamming balls:

Consider vectors in  $\{0, 1\}^n$ , Hamming distance between  $\underline{x}, \underline{y}$  is # coord. where  $x_i \neq y_i$ .

Hamming ball of radius  $r = \{ \text{all } \underline{y} \in \{0, 1\}^n : \Delta(\underline{x}, \underline{y}) \leq r \}$   
around  $\underline{x}$

Hamming ball <sup>radius r</sup> around  $\underline{0} = \{ \underline{x} : |\underline{x}| \leq r \}$

$$\text{vol}(B(r)) = \sum_{i=0}^r \binom{n}{i}$$

↑  
# non-zero coord.

Def The **binary entropy function**  $h: [0,1] \rightarrow \mathbb{R}$

$$h(p) = p \log \frac{1}{p} + (1-p) \log \frac{1}{1-p}$$

•  $h(p) = H(X)$ ,  $X \sim \text{Binom}(1, p)$

•  $h(p) \nearrow (0, \frac{1}{2})$ ,  $h(\frac{1}{2}) = \max = 1$ ,  $\searrow (\frac{1}{2}, 1)$

Stirling's formula:  $n! \sim \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$

$$\Rightarrow \binom{n}{pn} \sim \frac{2^{h(p) \cdot n}}{\sqrt{2\pi n p(1-p)}}$$

Thm 2.2 Let  $0 < p \leq \frac{1}{2}$ . Then  $\forall n \in \mathbb{N}$ ,

$$\sum_{i \leq pn} \binom{n}{i} \leq 2^{h(p) \cdot n}$$

quite tight

Pf: • Let  $A = \text{collect}^n$  of all subsets of  $[n]$  of

size  $\leq pn$ ,  $|A| = \sum_{i \leq pn} \binom{n}{i}$

• Take  $X \sim A \Rightarrow$  <sup>maximality</sup>  $H(X) = \log |A|$

Suffices to show

$$H(X) \leq h(p) \cdot n$$

• Now think of  $X = (X_1, \dots, X_n)$  where

$$X_i = \mathbb{1}_{\{i \in X\}}$$

NIS

$$H(X_i) \leq h(p)$$

$$H(X) = H(X_1, \dots, X_n) \stackrel{\text{subadd.}}{\leq} \sum_{i \in [n]} H(X_i) = n \underline{H(X_i)}$$

Let  $\alpha_i = \Pr(1 \in X) \leq p \leq \frac{1}{2}$

$X_i \sim \text{Binom}(1, \alpha_i)$

$$H(X_i) = h(\alpha_i) \leq h(p) \quad \blacksquare$$

Exer 2.3 (Chernoff using Thm 2.2)

Let  $X \sim \text{Binom}(n, \frac{1}{2})$  w/ standard deviation  $\sigma = \frac{\sqrt{n}}{2}$

$$\Rightarrow \forall c > 0, \Pr(|X - \frac{n}{2}| \geq c\sigma) \leq 2^{-\frac{c^2}{2} + 1}$$

2.3 Triangle maximisation

Q: Given that a graph  $G$  has  $m$  edges,

What is max #  $\Delta_s$  it can have?

Intuitively more compact  $\Rightarrow$  more  $\Delta$ s.

Say  $e(G) = m = \binom{k}{2}$ ,  $k \in \mathbb{N}$




$k$  vs

$$\#\Delta_s = \binom{k}{3} \approx \frac{(2m)^{3/2}}{6}$$

Thm 2.5 An  $m$ -edge graph has at most

$$\frac{(2m)^{3/2}}{6} \text{ triangles.}$$

Rmk. Thm 2.5 is an easy consequence of Kruskal-Katona.

Pf: • Let  $T = \{ \text{all labeled } \Delta \text{ s in } G \}$  

$$t = |T|, \text{ NTS } t \leq (2m)^{3/2}$$

Take  $X = (X_1, X_2, X_3) \sim T$ , where  $X_i$ s are  
vxs of the random  $\Delta X$ .

$$\stackrel{\text{max.}}{\Rightarrow} H(X) = \log t \stackrel{\text{NTS}}{=} \leq \frac{3}{2} \log(2m)$$

• Let  $F \sim \binom{[3]}{2} \Rightarrow \Pr(i \in F) = \frac{2}{3} = \mu$   
 $i \in \{1, 2, 3\}$

Shearer's lem

$$\Rightarrow H(X) \leq \frac{3}{2} \mathbb{E}_F H(X_F)$$

$\forall$  choice of value for  $F$ ,  $X_F$  takes values in the set.

of all labeled (ordered) edges

$$\max \Rightarrow H(X_F) \leq \log(2m) \quad \text{▨}$$

## § 2.4 $\Delta$ -intersecting family

Def A family of graphs  $\mathcal{F}$  on  $[n]$  is  $\Delta$ -intersecting

if  $\forall G, G' \in \mathcal{F}$ ,  $G \& G'$  shares  $\geq \Delta$ .

• Ellis-Films-Friedgut:  $\Delta$ -int. fam. size  $\leq 2^{\binom{n}{2} - 3}$   
on  $[n]$

tight: lower bd constr. fix  $\Delta_2^3$ , take all graphs containing this triangle.

Thm 2.6  $\forall \Delta$ -int. fam  $\mathcal{F}$  on  $[n]$  size  $\leq 2^{\binom{n}{2} - 2}$

PF: • Let  $X \sim \mathcal{F}$ ,

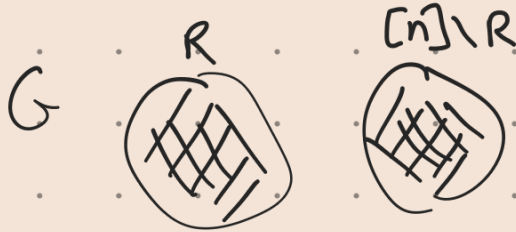
NIS  $H(X) = \log |\mathcal{F}| \leq \binom{n}{2} - 2$ .

• Think of  $X$  as the random vector

$(X_e : e \in \binom{[n]}{2})$ , where  $X_e = \mathbb{1}_{\{e \in X\}}$

• Consider  $R \sim 2^{[n]}$  (unif random subset of  $[n]$ )

- Let  $G$  be the random graph (the set of coord., w/ which we apply Shearer's lem.)



$$\mathbb{E}_G(e(G)) = \frac{1}{2} \binom{n}{2} \quad \& \quad \Pr(e \in G) = \mu = \frac{1}{2}$$

$$\forall e \in \binom{[n]}{2}$$

Shearer's lem.

$$\Rightarrow H(X) = H(X_e : e \in \binom{[n]}{2})$$

$$\leq \frac{1}{2} \mathbb{E}_G H(X_G) \quad ?$$

- $\mathcal{F}|_G$  is an intersecting family.

$\forall F, F' \in \mathcal{F}, E(F) \cap E(F')$  contains a  $\Delta$ .  
& this  $\Delta$  intersects  $G$ .



- $X_G$  takes values in  $\mathcal{F}|_G$ .

$$|\mathcal{F}|_G| \leq_{\text{intersecting}} 2^{e(G)-1}$$

$$\xrightarrow{\text{max.}} H(X_G) \leq \log |\mathcal{F}|_G| = e(G) - 1$$

- Finally  $H(X) \leq 2 \mathbb{E}_G H(X_G)$

$$\leq 2 \mathbb{E}_G (e(G) - 1) = \binom{n}{2} - 2$$

## § 1.4 Axioms determine entropy funct. uniquely

We show that

Lem 1.17 Let  $X$  be a r.v. defined over  $A$ .

If  $H(X)$  satisfies all six axioms.

$$\Rightarrow H(X) = \sum_{a \in A} p_a \log \frac{1}{p_a}, \quad \text{where } p_a = \Pr(X=a).$$

Lem 1.16 If  $X \sim A \Rightarrow H(X) = \log |A|$ .

Pf. • Consider the special case  $n = |A| = 2^k$

Take  $Y \sim [2] \xrightarrow{\text{normalisation}} H(Y) = 1$

Take i.i.d  $Y_1, \dots, Y_k$  of  $Y$ .

indep  $\Rightarrow H(Y_1, \dots, Y_k) = k H(Y) = k$

$(Y_1, \dots, Y_k) \sim [2]^k \cong [n] \sim X$

invariance  $\Rightarrow H(X) = H(Y_1, \dots, Y_k) = k = \log n = \log |A|$

• General case, let  $\delta = H(X) - \log n$   
 $X \sim A$   
 $|A| = n$  WTS  $\delta = 0$



Take i.i.d. copies  $X_1, \dots, X_r$  of  $X$ .

indep  
 $\Rightarrow$

$$H(X_1, \dots, X_r) = r H(X)$$

• if  $2^k \leq n^r \leq 2^{k+1}$

Len 1.3

& the special  
 $2^k$ -case

$$k \leq r \cdot H(X) \leq k+1$$

$$\Rightarrow \underbrace{\frac{k}{r} - \log n}_{\ominus} \leq \int \leq \underbrace{\frac{k+1}{r} - \log n}_{\oplus}$$

$$\Rightarrow |\delta| \leq \oplus - \ominus = \frac{1}{r} \rightarrow 0 \text{ as } r \rightarrow \infty$$