

①

* Bregman's Theorem

$$A = (a_{ij}) \in \{0,1\}^{n \times n}$$

$$\text{Perm}(A) := \sum_{\substack{\sigma: [n] \rightarrow [n] \\ \text{bijection}}} \prod_{i \in [n]} a_{i, \sigma(i)}$$

A can be thought of as the adjacency matrix of a bipartite graph $G = (L, R, E)$, where $L = R = [n]$, and $(i, j) \in E$ iff $a_{ij} = 1$.

A perfect matching in G is a ~~bijection set~~
 ~~$M \subseteq E$ s.t. every vertex is incident to exactly one edge in M .~~
 $M \subseteq E$ s.t. every vertex is incident to exactly one edge in M .

Equivalently, a perfect matching is a bijection $\sigma: [n] \rightarrow [n]$ s.t. $\forall i \in [n]$, $(i, \sigma(i)) \in E$.

$$\Rightarrow \text{Perm}(A) = \# \text{ perfect matchings in } G.$$

Def: $d_i := \text{deg. of the left vertex } i$.

$$\text{Obviously, } \text{Perm}(A) \leq \prod_{i \in [n]} d_i.$$

Bregman's Theorem: $\text{Perm}(A) \leq \prod_{i \in [n]} (d_i!)^{1/d_i}.$

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* Entropy proof by Jaikumar Radhakrishnan:

Let $\sigma: [n] \rightarrow [n]$ be a uniform random perfect matching.

Then Bregman's Theorem is equivalent to:

$$H(\sigma) \leq \sum_{i \in [n]} \frac{\log(d_i!)}{d_i} \quad (*)$$

Of course: $H(\sigma) = H(\sigma(1)) + H(\sigma(2) | \sigma(1)) + \dots +$
 $H(\sigma(n) | \sigma(1), \dots, \sigma(n-1))$

$$\leq \sum_{i \in [n]} H(\sigma(i)) \leq \sum_{i \in [n]} \log d_i,$$

which proves $\text{Perm}(A) \leq 2^{\sum \log d_i} = \prod_{i \in [n]} d_i,$

the trivial bound!

Trick: Introduce more randomness.

Let $\tau: [n] \rightarrow [n]$ be ~~any~~ ~~uniformly random~~ ~~any~~ given bijection. Then,

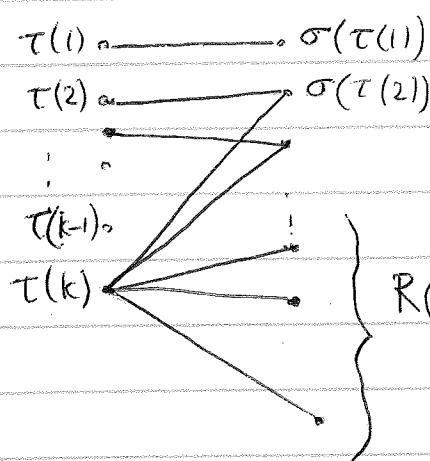
$$H(\sigma) = H(\sigma \cdot \tau), \text{ where } \sigma \tau(i) \triangleq \sigma(\tau(i)).$$

③

Suppose we have revealed

$$\sigma\tau(1), \sigma\tau(2), \dots, \sigma\tau(k-1),$$

How much information remains in $\sigma\tau(k)$?



[at most $\log$ of

Support size of $R(\sigma, \tau, k)$

under uniform σ and
Conditioned on observations.]

Lemma: For random vars X, Y , suppose

support(X) can be partitioned into sets

$$A_1, \dots, A_d \quad \text{s.t.} \quad \forall i, |\text{Supp}(Y|X \in A_i)| \leq i.$$

Then, $H(Y|X) \leq \sum_{i \in [d]} \Pr(X \in A_i) \log i$

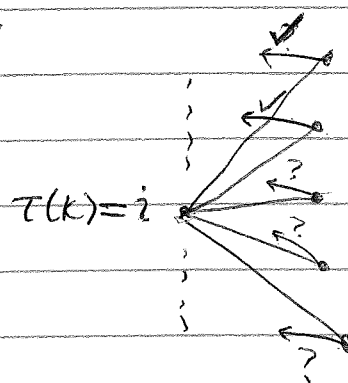
④

$$\begin{aligned}
 H(\sigma) &= \sum_{k \in [n]} H(\sigma(k) | \sigma(\tau(1)), \dots, \sigma(\tau(k-1))) \\
 &= \mathbb{E}_T \sum_k \quad " \\
 &= \sum_k \mathbb{E}_T \quad "(i = \tau(k)) \\
 &= \sum_{i \in [n]} \mathbb{E}_T H(\sigma(i) | \sigma(\tau(1)), \dots, \sigma(\tau(\tau^{-1}(i)-1))) \\
 &\stackrel{\text{lemma}}{\leq} \sum_{i \in [n]} \mathbb{E}_T \sum_{j \in [d_i]} \Pr_{\sigma}(|R(\sigma, \tau, \tau^{-1}(i))| = j) \log j \\
 &= \sum_{i, j} \mathbb{E}_T \quad " \\
 &= \sum_{i, j} \Pr_{\sigma, \tau}(|R| = j) \log j \quad (**)
 \end{aligned}$$

Claim: $\Pr_{\sigma, \tau}(|R| = j) = \frac{1}{d_i}$
 $(\forall j)(\forall \sigma)$

~~T random~~
 $\sigma(\tau(1)), \dots, \sigma(\tau(k-1))$ revealed
 T random $\Rightarrow k$ random.

$|R| = \#$ of ~~neighbors~~ unrevealed neighbors of i .



R only depends on the ordering of neighbors of i under T .

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$|R| = d_i - l + 1$, $l = \text{order of } \sigma(i)$
~~under~~ among $\Gamma(i)$
 under τ
 $= \text{uniform random.}$

$\Rightarrow |R| = \text{uniform random}$

$$\Rightarrow P_r(|R| = j) = \frac{1}{d_i} \quad (\forall \sigma, j)$$

□

By Claim $\Rightarrow (**) = \sum_{\substack{i, j \\ i \in [n] \rightarrow j \in [d_i]}} \frac{1}{d_i} \log j$

$$= \sum_{i \in [n]} \frac{1}{d_i} \log(d_i!) \Rightarrow (*)$$

□

$$\left(H(\sigma) \leq \sum_{i \in [n]} \frac{1}{d_i} \log(d_i!) \right)$$

⑥

* Shearer's Lemma:

$X = (X_1, \dots, X_n)$ S a r.v. on subsets of $[n]$,

s.t. $\forall i, \Pr(i \in S) \geq \mu$ Then,

$$\mathbb{E}_S H(X_S) \geq \mu \cdot H(X).$$

e.g., S uniform on singletons $\Rightarrow \mu = \frac{1}{n} \Rightarrow \mathbb{E}_S H(X_S) \geq \frac{H(X)}{n}$

$$\Leftrightarrow \sum H(X_i) \geq H(X).$$

(sub additivity of entropy)

* Proof (due to Jaikumar Radhakrishnan):

Fix $T = \{i_1, \dots, i_k\}$, $i_1 < i_2 < \dots < i_k$

$$\begin{aligned} H(X_T) &= H(X_{i_1}) + H(X_{i_2} | X_{i_1}) + \dots + H(X_{i_k} | X_{i_1}, \dots, X_{i_{k-1}}) \\ &\geq H(X_{i_1} | X_{<i_1}) + H(X_{i_2} | X_{<i_2}) + \dots + H(X_{i_k} | X_{<i_k}). \end{aligned}$$

[(more conditioning)]

$$\text{Now, } \mathbb{E}_S (H(X_S)) \geq \mathbb{E}_S \sum_{i \in S} H(X_i | X_{<i})$$

$$= \sum_{i \in [n]} \Pr(i \in S) \cdot H(X_i | X_{<i})$$

$$\geq \mu \sum_{i \in [n]} H(X_i | X_{<i}) = \mu H(X). \quad \square$$

⑦

* Application: Counting embeddings of Graphs.

let T be a fixed graph with m edges.

What is the max. copies of T that can be ~~embedded~~ ^{packed} in a graph with l edges?

Def: $\wedge$ Embedding of T in G :

$$\sigma: V(T) \rightarrow V(G)$$

$$\{i, j\} \in E(T) \Leftrightarrow \{\sigma(i), \sigma(j)\} \in E(G).$$

Def: $N(T, l) = \max$ # of embedding of T in a graph with l edges.

Theorem (~~Alon, Freidgut, Kahn~~)

$$\frac{N(T, l)}{l^m} \leq \frac{1}{2^m}$$

* Fractional indep. sets:

$$\psi: V(T) \rightarrow [0, 1] \text{ s.t. } \forall \overset{\{u, v\}}{\uparrow} e \in E(T),$$

$$\psi(u) + \psi(v) \leq 1.$$

$$|\psi| := \sum_{v \in V} \psi(v).$$

$$\psi^* := \psi \text{ that maximizes } |\psi|.$$

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* Fractional cover:

$$\phi : E(G) \rightarrow [0,1], \forall v \in V(T),$$

$$\sum_{\substack{e \in E(T) \\ v \in e}} \phi(e) \geq 1.$$

$$|\phi| = \sum_{e \in E(T)} \phi(e).$$

* $\phi^* := \phi$ that minimizes $|\phi|$.

* By LP-duality, $|\phi^*| = |\psi^*|$.

* Example Complete graph with k vertices,

ψ^* assigns $\frac{1}{2}$ to every vertex &

$\phi^* \approx \frac{1}{k-1}$ to every edge.

*

(a)

Theorem: (Alon, Freidgut-Kahn)

$$\left(\frac{l}{m}\right)^{|\psi^*|} \leq N(T, l) \leq (2l)^{|\psi^*|}$$

Proof: (upper bound)

Let σ be a uniform random embedding of $T \rightarrow G$, where G is any graph with l edges.

All we need to bound is $H(\sigma)$.

* Let $S \in E(T)$ be random with

$$\Pr(S=e) \triangleq \frac{\phi^*(e)}{|\phi^*|}$$

Apply Shearer's Lemma on r.v.'s

$\{\sigma(v) : v \in V(T)\}$ with respect to S .

$\Rightarrow$ N.B. $\forall v, \Pr(v \in S) \geq \sum_{e \in E(T)} \frac{\phi^*(e)}{|\phi^*|} \stackrel{[\text{def. cover}]}{\geq} \frac{1}{|\phi^*|}$

Shearer $\Rightarrow \mathbb{E} \sum_{e \in E(T)} H(\sigma_e) \geq \frac{H(\sigma)}{|\phi^*|}$ [an edge of G]

On the other hand, $\forall e \in E(T), H(\sigma_e) \leq \log(2l)$

$\Rightarrow \mathbb{E} \sum_{e \in E(T)} H(\sigma_e) \leq \log(2l)$

(10)

$$\Rightarrow N(T, l) \leq (2l)^{|\phi^*|} = (2l)^{|\psi^*|}.$$

Lower bound: (assuming $(l/m)^{|\psi^*|}$ is integer)

Construct G from T as follows:

1) replace every $v \in T$ with an indep. set of $\binom{l}{m}^{\psi^*(v)}$ vertices.

2) Connect every vertex in indep. set for u to v iff $\{u, v\} \in E(T)$.

* Every edge of T contributes to

$$\binom{l}{m}^{\psi^*(u) + \psi^*(v)} \text{ edges in } G \leq \binom{l}{m}.$$

$\Rightarrow G$ has $\leq l$ edges.

* An embedding can choose any of the vertices in each indep. set $\Rightarrow$

$$\# \text{ embeddings} \geq \binom{l}{m}^{\sum_v \psi^*(v)} = \binom{l}{m}^{|\psi^*|}.$$

□

⑪

Corollary 1) $N(K_k, l) \leq (2l)^{k/2}$

Corollary 2) Any graph with l edges
contains ~~less~~ $\leq \frac{1}{6}(2l)^{3/2}$ triangles.
