

An entropy proof of the Erdős-Kleitman-Rothschild theorem

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Definition. A graph G is H -free if $H \not\subseteq G$
not necessarily induced subgraph

$$\mathcal{F}_n(H) := \left\{ G \subseteq K_n : H \not\subseteq G \right\}$$

complete graph with vertex set $\{1, \dots, n\}$

Turán's problem. Determine

$$ex(n, H) := \max \{ e(G) : G \in \mathcal{F}_n(H) \}$$

This talk. What does a typical H -free graph look like?

Theorem (EKR 1976) Draw $G \in \mathcal{F}_n(K_3)$ uniformly at random. Then

$$\lim_{n \rightarrow \infty} \mathbb{P}(G \text{ is bipartite}) = 1.$$

This is a counting statement: $|\{G \in \mathcal{F}_n(K_3) : \chi(G) > 2\}| \ll |\{G \subseteq K_n : \chi(G) \leq 2\}|$.

Moreover, for every fixed $r \geq 3$,

$$2^{\text{ex}(n, K_r)} \leq |\mathcal{F}_n(K_r)| \leq 2^{\text{ex}(n, K_r) + o(n^2)}.$$

↑
trivial

Theorem (Erdős-Frankl-Rödl 1986) For every H ,

$$|\mathcal{F}_n(H)| = 2^{\text{ex}(n, H) + o(n^2)}$$

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Proof methods:

EKR: Kövári-Sós-Turán / Erdős + induction on n & r

EFR: regularity lemma

BMS+ST: hypergraph containers + supersaturated Turán's theorem

KMPS: Shannon's entropy + supersaturated Turán's theorem

Entropy. X : random var. taking values in a finite set \mathcal{X} .

$$H(X) := - \sum_{x \in \mathcal{X}} P(X=x) \log_2 P(X=x)$$

chosen for convenience today

Informally, $H(X) = \mathbb{E}[\text{surprise}(X)]$.

Conditional entropy. X, Y : random vars taking values in finite sets \mathcal{X}, \mathcal{Y} , respectively.

$$H(X|Y) := \mathbb{E}[H(X_{\substack{\downarrow \\ \{Y=y\}}})]$$

X conditioned on $\{Y=y\}$

$$= - \sum_{y \in \mathcal{Y}} P(Y=y) \sum_{x \in \mathcal{X}} P(X=x|Y=y) \log_2 P(X=x|Y=y)$$

Properties of entropy:

$$\textcircled{1} \quad 0 \leq H(X) \leq \log_2 |\mathcal{X}|$$

'=' iff X is uniform on \mathcal{X}

$$\textcircled{2} \quad H(X|Y) = H(X,Y) - H(Y) \stackrel{'=' \text{ iff } X \text{ and } Y \text{ are independent}}{\leq} H(X)$$

Corollary: $H(X_1, \dots, X_n|Y) \leq \sum_{i=1}^n H(X_i|Y)$

$$\textcircled{3} \quad H(X|Y,Z) \leq H(X|Y)$$

\textcircled{4} Pinsker's inequality:

$$d_{TV}((X,Y), X \times Y) \leq \sqrt{2 \cdot H(X) - H(X|Y)}$$

total variation distance ↑
joint dist. ↑ ↗ random vector with independent coordinates

An entropy proof of: $\log_2 |\mathcal{F}_n(K_3)| = \text{ex}(n, K_3) + o(n^2)$.

Choose $X \in \mathcal{F}_n(K_3)$ u.a.r. so that $H(X) \stackrel{(1)}{=} \log_2 |\mathcal{F}_n(K_3)|$.

Think of X as the vector $(X_e)_{e \in K_n}$.

Key Lemma. There exists $F \subseteq K_n$ with $e(F) \leq n^{2-\varepsilon}$ s.t., letting $p_e := \mathbb{E}[X_e | (X_f)_{f \in F}]$, we have

$$\sum_{i,j,k} \mathbb{E}[p_{ij} \cdot p_{ik} \cdot p_{jk}] \leq n^{3-5\varepsilon}.$$

$(x_f)_{f \in F}$ -measurable r.v.

When $F = \emptyset$, then $p_e \approx \frac{1}{4}$ for each $e \in K_n$.

When $F = K_n$, then $p_{ij} \cdot p_{ik} \cdot p_{jk} = 0$ for all i, j, k .

Define $G := \{e \in K_n : p_e \geq n^{-\varepsilon}\}$. $(X_F)_F$ -measurable random graph

$$N_{K_3}(G) \cdot n^{-3\varepsilon} \leq \sum_{i,j,k} p_{ij} p_{ik} p_{jk} \quad \begin{matrix} \text{Key Lemma} \\ \downarrow \\ F \subseteq K_n \end{matrix}$$

Let $B := \left\{ \sum_{i,j,k} p_{ij} p_{ik} p_{jk} > n^{3-4\varepsilon} \right\}$ and note that
 $(X_F)_F$ -measurable event

$$B^c \Rightarrow N_{K_3}(G) \leq n^{3-\varepsilon} \Rightarrow e(G) \leq \underline{\text{ex}(n, K_3)} + O(n^{2-\delta}).$$

↑
 Erdős-Simonovits ('supersaturation' – a simple 'averaging' argument)

Key Lemma + Markov's ineq. $\Rightarrow \underline{P(B) \leq n^{-\varepsilon}}$

For every $F' \subseteq F$, let $X_e^{F'} := X_e|_{(X_f)_F = F'}$ $p_e(F') = \mathbb{E}[X_e^{F'}]$

$$\begin{aligned}
 H(X) &\stackrel{(2)}{=} H((X_f)_F) + H((X_e)_{K_n \setminus F} | (X_f)_F) \\
 &\leq H((X_f)_F) + P(B) \cdot \max \{ H((X_e^{F'})_{K_n \setminus F}) : F' \in B \} \\
 &\quad + P(B^c) \cdot \max \{ H((X_e^{F'})_{K_n \setminus F}) : F' \notin B \} \\
 &\stackrel{(1)}{\leq} c_F + P(B) \cdot e(K_n \setminus F) \\
 &\quad + P(B^c) \cdot \max \{ \sum_{e \in K_n \setminus F} H(X_e^{F'}) : F' \notin B \} \\
 &\stackrel{KL}{\leq} \underbrace{n^{2-\varepsilon} + n^{-\varepsilon} \cdot \binom{n}{2}}_{\leq 2n^{2-\varepsilon}} + \underbrace{\max \{ \sum_{e \in K_n} H(X_e^{F'}) : F' \notin B \}}_{\text{main term}}
 \end{aligned}$$

$$\log_2 |\mathcal{F}_n(K_3)| \leq \max \left\{ \sum_{e \in K_n} H(X_e^{F'}) : F' \notin \mathcal{B} \right\} + 2n^{2-\varepsilon}$$

Finally,

$$\sum_{e \in K_n} H(X_e^{F'}) = \sum_{e \in K_n} H(\text{Ber}(p_e))$$

$$\begin{aligned} &\leq e(G^{F'}) + e(K_n \setminus G^{F'}) \cdot \max_{p \leq n^{-\varepsilon}} H(\text{Ber}(p)) \\ &\leq e(G^{F'}) + O(n^{2-\varepsilon} \log n) \end{aligned}$$

and

$$\max \{ e(G^{F'}) : F' \notin \mathcal{B} \} \leq \underbrace{\text{ex}(n, K_3)}_{\substack{N_{K_3}(G) \leq n^{3-\varepsilon} \\ \uparrow \\ \downarrow \text{supersaturation}}} + O(n^{2-\delta}).$$

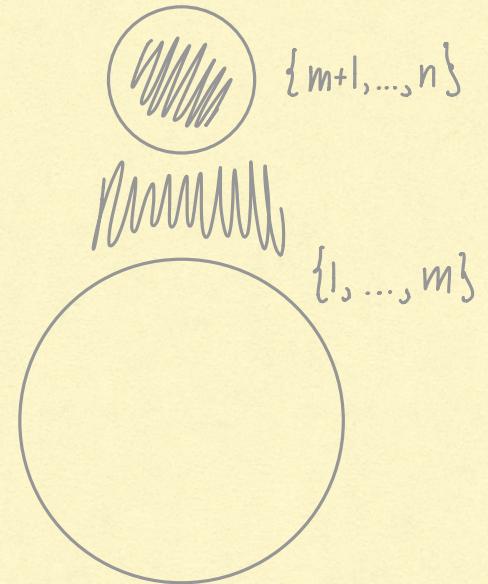
Key Lemma. There exists $F \subseteq K_n$ with $e(F) \leq n^{2-\varepsilon}$ s.t.,
 letting $p_e := \mathbb{E}[X_e | (X_f)_{f \in F}]$, we have

$(X_f)_F$ -measurable r.v.

$$\sum_{i,j,k} \mathbb{E}[p_{ij} \cdot p_{ik} \cdot p_{jk}] \leq n^{3-5\varepsilon}.$$

Proof:

Define $F_m \ni ij \Leftrightarrow \max\{i, j\} > m$ s.t.



- $\emptyset = F_n \subseteq F_{n-1} \subseteq \dots \subseteq F_1 = K_n$
- $e(F_m) \leq n(n-m)$
- $F_{m-1} \setminus F_m \ni ij \Leftrightarrow \max\{i, j\} = m$

Let

$$h_m := H(X_{12} | (X_f)_{F_m}).$$

$$0 \leq H(X) \leq \log_2 |X|$$

$$H(X|Y, Z) \leq H(X|Y)$$

By ① & ③, since $F_m \subseteq F_{m-1}$,

$$0 \geq h_n \geq h_{n-1} \geq \dots \geq h_1 \geq 0.$$

In particular, there must be $m \in \{n-\sqrt{n}, \dots, n\}$ s.t.

$$h_m - h_{m-1} \leq 1/\sqrt{n}.$$

We set $F := F_m$.

$$\underbrace{H(X_{12}|(X_f)_F)}_{h_m} - \underbrace{H(X_{12}|(X_f)_{F_{m-1}})}_{h_{m-1}} \leq \frac{1}{\sqrt{n}}$$

③ $\Downarrow F \setminus F_{m-1} = F_m \setminus F_{m-1} \ni \{1, m\}, \{2, m\}$

$$\underbrace{H(X_{12}|(X_f)_F)}_{h_m''} - \underbrace{H(X_{12}|(X_f)_F, X_{1m}, X_{2m})}_{\geq h_{m-1}} \leq \frac{1}{\sqrt{n}}$$

$\Updownarrow F' \sim (X_f)_F$

$$\mathbb{E}[H(X_{12}^{F'}) - H(X_{12}^{F'} | X_{1m}, X_{2m})] \leq \frac{1}{\sqrt{n}}$$

$d_{TV}((X, Y), X \times Y) \leq \sqrt{2(H(X) - H(X|Y))}$

\Downarrow Pinsker \Downarrow Jensen

$$\mathbb{E}[d_{TV}((X_{12}^{F'}, X_{1m}, X_{2m}), X_{12}^{F'} \otimes (X_{1m}^{F'}, X_{2m}^{F'}))] \leq \mathbb{E}[\sqrt{2} \otimes] \leq \sqrt{2} n^{-1/4}$$

$$\mathbb{E}[d_{TV}((X_{12}^{F'}, X_{1m}^{F'}, X_{2m}^{F'}), X_{12}^{F'} \times (X_{1m}^{F'}, X_{2m}^{F'}))] = \sqrt{2} n^{-1/4}$$

By symmetry (every permutation of $\{1, \dots, m\}$ fixes F), we may replace $(1, 2, m)$ with any $(i, j, k) \in \{1, \dots, m\}^3$.

$$\mathbb{E}[p_{ij} \cdot p_{ik} \cdot p_{jk}] \leq \mathbb{E}[d_{TV}(\underbrace{(X_{ij}^{F'}, X_{ik}^{F'}, X_{jk}^{F'})}_{P_1}, \underbrace{X_{ij}^{F'} \times X_{ik}^{F'} \times X_{jk}^{F'}}_{P_2})]$$

Δ -ineq.

consider the event $T := \{X_{ij} X_{ik} X_{jk} = 1\}$:

$$P_1(T) = 0 \quad \& \quad P_2(T) = p_{ij} \cdot p_{ik} \cdot p_{jk}$$

$$2\sqrt{2} n^{-1/4}$$

If $\max\{i, j, k\} > m$, then $p_{ij} p_{ik} p_{jk} = 0$.

$$\boxed{\sum_{i,j,k} \dots \blacksquare}$$