

# The Erdős Matching Conjecture

## and related questions

Andrey Kupavskii

UNIVERSITY OF BIRMINGHAM

<https://www.birmingham.ac.uk/>

MOSCOW INSTITUTE OF PHYSICS AND TECHNOLOGY

<https://mipt.ru/english/>

## Definitions

Denote  $\binom{[n]}{k}$ : the set of all  $k$ -element subsets of  $[n]$ .

A subset  $\mathcal{F} \subset \binom{[n]}{k}$  is called a *family*.

A *matching of size  $s$*  in  $\mathcal{F}$ :  $s$  pairwise disjoint sets  $F_1, \dots, F_s \in \mathcal{F}$ .

The *matching number*  $\nu(\mathcal{F})$  of  $\mathcal{F}$ : the size of the largest matching in  $\mathcal{F}$ .

If  $\nu(\mathcal{F}) = 1$ , then  $\mathcal{F}$  is *intersecting*: any two sets in  $\mathcal{F}$  intersect.

## The extremal quantity

Define

$$e_k(n, s) := \max \left\{ |\mathcal{F}| : \mathcal{F} \subset \binom{[n]}{k}, \nu(\mathcal{F}) < s \right\}.$$

**Theorem (Erdős-Ko-Rado, 1938-1961)**

$$e_k(n, 2) = \binom{n-1}{k-1} \quad \text{for } n \geq 2k.$$

The theorem is tight. Consider the family  $\left\{ A \in \binom{[n]}{k} : 1 \in A \right\}$ .

For  $n = 2k$  the family  $\binom{[2k-1]}{k}$  has the same cardinality.

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## The Erdős Matching Conjecture

How to construct a large family  $\mathcal{F} \subset \binom{[n]}{k}$ , satisfying  $\nu(\mathcal{F}) < s$ ?

$$\mathcal{A}^{(k)}(n, s) := \left\{ A \in \binom{[n]}{k} : A \cap [s-1] \neq \emptyset \right\}, \quad \mathcal{B}^{(k)}(n, s) := \binom{[sk-1]}{k}.$$

We have  $|\mathcal{A}^{(k)}(n, s)| = \binom{n}{k} - \binom{n-s+1}{k}, \quad |\mathcal{B}^{(k)}(n, s)| = \binom{sk-1}{k}.$

### The Erdős Matching Conjecture, 1965

For  $n \geq sk$  we have

$$e_k(n, s) = \max\{|\mathcal{A}^{(k)}(n, s)|, |\mathcal{B}^{(k)}(n, s)|\}.$$

Put  $x := s/n$ . If  $k$  is fixed and  $s \rightarrow \infty$ :  $|\mathcal{A}|/\binom{n}{k} \rightarrow 1 - (1-x)^k,$   
 $|\mathcal{B}|/\binom{n}{k} \rightarrow (kx)^k.$

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## Asymptotic fractional version of the EMC.

A **fractional matching** for  $\mathcal{F} \subset 2^{[n]}$ : a function  $w : \mathcal{F} \rightarrow [0, 1]$ , such that

$$\sum_{F \in \mathcal{F}: i \in F} w(F) \leq 1 \quad \text{for every element } i \in [n].$$

**Fractional matching number**  $\nu^*(\mathcal{F})$ :

the size of the largest fractional matching in  $\mathcal{F}$ .

### Conjecture A (Alon et. al., 2012)

Let  $x \in [0, 1/k]$  be fixed and let  $\mathcal{F}_n \subset \binom{[n]}{k}$  be a sequence of families such that  $\nu^*(\mathcal{F}) \leq xn$ . Then

$$\limsup_{n \rightarrow \infty} \frac{|\mathcal{F}|}{\binom{n}{k}} \leq \max \{1 - (1 - x)^k, (kx)^k\}.$$

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## Deviation of sums of random variables

$\mathbf{X} := (X_1, \dots, X_k)$ :  $X_i \geq 0$  are i.i.d. random variables,  $E[X_i] = x$ .

$$m_k(x) := \sup_{\mathbf{X}} \Pr[X_1 + \dots + X_k \geq 1].$$

Note:  $m_k(x) = 1$  for  $x \geq 1/k$ .

**Conjecture B** (Łuczak, Mieczkowska, Šileikis, 2017)

$$m_k(x) = \max \{1 - (1 - x)^k, (kx)^k\}.$$

Case  $k = 2$  was resolved by Hoeffding and Shrikhande (1955).

Related conjectures of Samuels (1966) and Feige (2006) speak about random variables that are **not** necessarily **identically distributed**.

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# Conjectures A and B are equivalent

(Alon, Frankl, Huang, Rödl, Ruciński, Sudakov, 2012)

**Conjecture B**  $\Rightarrow$  **Conjecture A**. Take the largest  $\mathcal{F}$  with  $\nu^*(\mathcal{F}) \leq xn$ .  
By LP-duality: ( $\nu^* = \tau^*$ ) there exists  $w : [n] \rightarrow [0, 1]$ , such that

$$\sum_{i \in [n]} w(i) = xn, \quad \text{and} \quad \sum_{i \in F} w(i) \geq 1 \text{ for every } F \in \mathcal{F}.$$

Define a random variable:  $w(t)$  for a randomly chosen  $t \in [n]$ . Form  $\mathbf{v} := (w(t_1), \dots, w(t_k))$ . Then

$$m_k(x) \geq \Pr \left[ \sum_{i=1}^k w(t_i) \geq 1 \right] \gtrsim \Pr[\mathbf{v} \in \mathcal{F}] = \frac{|\mathcal{F}|}{\binom{n}{k}}.$$

## What do we know about EMC?

True for  $k = 2$  (Erdős and Gallai, 1959)

True for  $k = 3$  (Łuczak and Mieczkowska, 2014, for large  $s$ ; Frankl, 2017, for all  $s$ ).

True for  $n > n_0(s, k)$  (Erdős, 1965)

True for  $n > 2k^3s$  (Bollobás, Daykin, Erdős, 1976)

True for  $n > 100ks^2$  (Frankl, Füredi, 1987)

True for  $n > 3k^2s$  (Huang, Loh, Sudakov, 2012)

True for  $n \geq (2s - 1)k - s$  (Frankl, 2013)

Connections to large deviation bounds, frac versions (Alon et. al. 2012)

Equivalence of Conjectures A, B (Łuczak, Mieczkowska, Šileikis, 2017)

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# New results

## Theorem (AK, Frankl, 2018+)

There exists  $s_0$  such that the EMC is true for  $s \geq s_0$ , any  $k$  and  $n \geq \frac{5}{3}sk - \frac{2}{3}s$ .

Consequently, Conjectures A and B hold for  $x < \frac{3}{5k-2}$ . Previous best due to the equivalence and the result of Frankl:  $x < \frac{1}{2k-1}$ .

We also get a bound  $e_k(n, s) \leq c(s-1) \binom{n-1}{k-1}$ , where  $c < 1$  and depends on  $sk/n$ .



# Dirac-type thresholds

$m_k^d(n)$  ( $f_k^d(n)$ ): **minimum  $d$ -degree** in  $\mathcal{F} \subset \binom{[n]}{k}$  that guarantees the existence of a **perfect** (fractional) **matching**.

**Theorem (Alon et. al., 2012; Treglown and Zhao, 2016)**

If  $\limsup_{n \rightarrow \infty} f_k^d(n) / \binom{n-d}{k-d} = c^*$ , then

$$\limsup_{n \rightarrow \infty} m_k^d(n) / \binom{n-d}{k-d} = \max\{c^*, 1/2\}.$$

If  $c^* < 1/2$ , then we know  $m_k^d(n)$  **exactly** for large  $n$ . Also,

$$f_k^d(n) \leq e_{k-d}(n, n/k) + 1.$$

**Corollary (Kupavskii, Frankl, 2018+)**

Determination of  $c^*$  for  $d \geq 2k/5$ ; exact values of  $m_k^d(n)$  for  $d \geq 3k/8$ .

Previous best: for  $d \geq k/2$ : Pikhurko (2008) for  $d \leq k-2$ ;  
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# Proof ingredients

Take the approach of Frankl as a base. The original approach uses:

1. Shifting.
2. Shadows of families with small matching numbers.
3. Inequality on the sum of sizes of cross-dependent families.

We add the following ingredients:

4. Better bounds on shadows.
5. Concentration inequalities for intersections of families and matchings.
6. Induction

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## A concentration inequality

Consider a family  $\mathcal{F} \subset \binom{[n]}{k}$  for  $n = kt$  of *density*  $\alpha := |\mathcal{F}| / \binom{[n]}{k}$ .

Take a  $t$ -matching  $\mathcal{M} \subset \binom{[n]}{k}$  uniformly at random.

Define a random variable  $\eta := |\mathcal{M} \cap \mathcal{F}|$ . Then  $\mathbb{E}[\eta] = \alpha t$ .

### Theorem (AK, Frankl, 2018+)

For any  $\beta > 0$ , we have  $\Pr [|\eta - \alpha t| \geq 2\beta\sqrt{t}] \leq 2e^{-\beta^2/2}$ .

# Proof outline

Assume  $\mathcal{M} = \{M_1, \dots, M_t\}$ . We have  $\eta = \eta_1 + \dots + \eta_t$ , where  $\eta_i$  indicates if  $M_i \in \mathcal{F}$ .

Define a **martingale**  $X_0, \dots, X_t$ , where  $X_i := \mathbb{E}[\eta \mid \eta_1, \dots, \eta_i]$ .

Note that  $X_0 = \mathbb{E}[X_0]$  and  $X_t = \eta$ .

Assume  $|X_i - X_{i-1}| \leq 2$  for any  $i$ .

## Azuma-Hoeffding inequality (1963, 1967)

If  $X_0, \dots, X_t$  is a martingale and  $|X_i - X_{i-1}| \leq 2$  for any  $i \in [t]$ , then  $\Pr [|X_t - X_0| \geq 2\beta\sqrt{t}] \leq 2e^{-\beta^2/2}$ .

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## Proof of $|X_i - X_{i-1}| \leq 2$ for any $i$

$Y_{i-1} := \mathbb{E}[\eta \mid M_1, \dots, M_{i-1}]$  and  $Y_i := \mathbb{E}[\eta \mid \eta_i, M_1, \dots, M_{i-1}]$ .

It is sufficient to show  $|Y_i - Y_{i-1}| \leq 2$ .

Fix  $M_1, \dots, M_{i-1}$ , put  $S := [n] \setminus (\cup_{j=1}^{i-1} M_j)$  and consider  $\mathcal{F}' := \mathcal{F} \cap \binom{S}{k}$ .

Kneser graph  $KG_{S,k}$ : vertices —  $\binom{S}{k}$ ,  
edges — pairwise disjoint sets.

$\mathcal{F}' \subset \binom{S}{k}$  gives an induced subgraph of  $KG_{S,k}$ .

Denote  $\alpha' := |\mathcal{F}'| / \binom{|S|}{k}$ ,  $e(\mathcal{F}')$ : proportion of edges of  $KG_{S,k}$  contained inside the subgraph induced on  $\mathcal{F}'$ .

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$\frac{Y_{i-1}}{t-i+1} = \alpha'$  (“the density of  $\mathcal{F}'$ ”)

$Y_i$ : random variable with two values.

If  $\eta_i = 1$ , then  $\frac{Y_i - 1}{t-i} = \frac{2e(\mathcal{F}')}{\alpha'}$  (“the average degree of  $\mathcal{F}'$ ”)

$\lambda'$ : the second largest absolute value of an eigenvalue of  $KG_{S,k}$ .

We use the Alon-Chung bound:

$$\left| \frac{2e(\mathcal{F}')}{\alpha'} - \alpha' \right| \leq \frac{\lambda(1 - \alpha')}{d}.$$

In Kneser graphs:  $\frac{\lambda}{d} = \frac{1}{t-i}$ . Therefore,

$$|Y_i - Y_{i-1}| \leq 2.$$

Similar for  $\eta_i = 0$

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### Theorem (AK, Frankl, 2018+)

There exists  $s_0$  such that the following holds. Fix  $s \geq s_0$ ,  $k$  and  $n \geq \frac{5}{3}sk - \frac{2}{3}s$ . Then any family  $\mathcal{F} \subset \binom{[n]}{k}$  with  $\nu(\mathcal{F}) < s$  satisfies

$$|\mathcal{F}| \leq \binom{n}{k} - \binom{n-s+1}{k}.$$

### Theorem (AK, Frankl, 2018+)

Take  $k, t \in \mathbb{N}$  and  $n = kt$ . Fix  $\mathcal{F} \subset \binom{[n]}{k}$  of density  $\alpha$ . For any  $\beta > 0$  the random variable  $\eta := |\mathcal{M} \cap \mathcal{F}|$ , where  $\mathcal{M}$  is randomly chosen  $t$ -matching, satisfies

$$\Pr [|\eta - \alpha t| \geq 2\beta\sqrt{t}] \leq 2e^{-\beta^2/2}.$$