

# Packing nearly optimal Ramsey $R(3, t)$ graphs

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Joint work with Lutz Warnke

# Context of this talk

## Ramsey number $R(s, t)$

$R(s, t) :=$  minimum  $n \in \mathbb{N}$  such that every red/blue edge-coloring of complete  $n$ -vertex graph  $K_n$  contains red  $K_s$  or blue  $K_t$

- Major problem in combinatorics: determining asymptotics
- Testbed for new proof techniques/methods:  
Alteration, LLL, Concentration Ineq., Semi-Random, Differential Eq.

## Celebrated Result (Ajtai-Komlós-Szemerédi 1980 + Kim 1995)

$$R(3, t) = \Theta(t^2 / \log t)$$

- Lower bound harder: Kim received Fulkerson Prize 1997
- $R(3, t) = \Omega(t^2 / (\log t)^2)$  already by Erdős in 1961

## Topic of this talk

Extension of Kim-result (implies asymptotics of other Ramsey parameter)

## Review of previous results

Erdős (1961) + Spencer (1977) + Krivelevich (1994)

All find an  $n$ -vertex graph  $G \subseteq K_n$  such that  $G$  is  $\Delta$ -free with independence number  $\alpha(G) \leq C\sqrt{n \log n}$

- Construct  $G$  in the binomial random graph  $G_{n,p}$

Kim (1995) + Bohman (2008): one nearly optimal  $R(3, t)$  graph

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- \* Tight up to the constant: Ajtai-Komlós-Szemerédi (1980)
- \* Lead to the right order of magnitude of Ramsey number  $R(3, t)$
- Construct  $G$  by (semi-random variation of)  $\Delta$ -free process: greedily add random edges that do not create a  $\Delta$

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## Why this result is difficult?

Standard approach (**alteration**): in  $G_{n,p}$ , try to remove one edge of all  $\Delta$ 's

Facts: w.h.p.

- #edges =  $\Theta(n^2 p)$
- # $K_3$ 's =  $\Theta(n^3 p^3) = \Theta(n^2 p \cdot np^2) \ll \text{\#edges} \Rightarrow p = \varepsilon / \sqrt{n}$
- Max ISET of  $G_{n,p} \approx \frac{2 \log n}{p} \stackrel{!}{=} C\sqrt{n} \cdot \log n \gg \sqrt{n \log n}$

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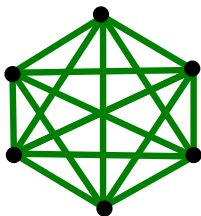
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**$\Delta$ -free process: add one random edge in each step**

open (can be added)  $\longrightarrow$



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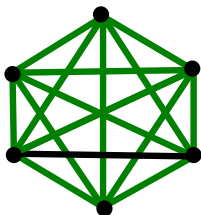
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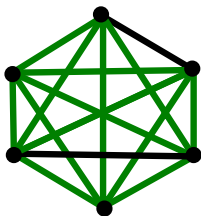
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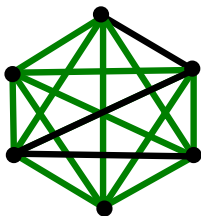
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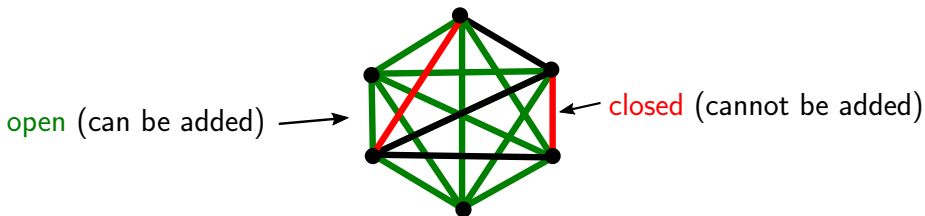
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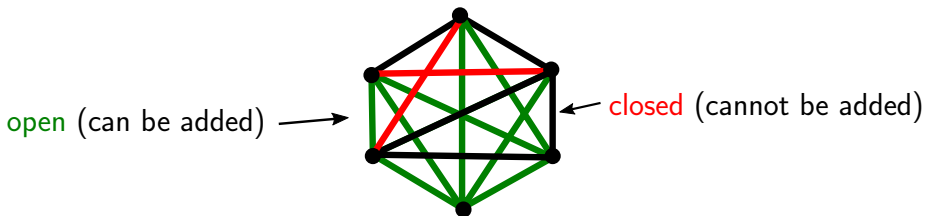
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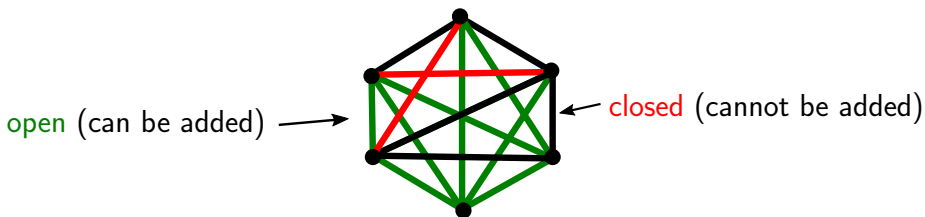
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**Semi-random variation: add many random-like edges in each step**



# Main Result: packing nearly optimal $R(3, t)$ graphs

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G., Warnke (2020): almost packing of nearly optimal  $R(3, t)$  graphs

Given  $\varepsilon > 0$ , we find edge-disjoint graphs  $(G_i)_{i \in \mathcal{I}}$  with  $G_i \subseteq K_n$  such that

- (a) each  $G_i$  is  $\Delta$ -free with  $\alpha(G_i) \leq C_\varepsilon \sqrt{n \log n}$
- (b) the union of the  $G_i$  contains  $\geq (1 - \varepsilon) \binom{n}{2}$  edges

- Using simple *polynomial-time randomized algorithm*:  
sequentially choose  $G_i$  via semi-random variation of  $\Delta$ -free process
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Motivation: why should we care?

- Natural packing extension of Kim's result
- Technical challenge: controlling errors over  $\Theta(\sqrt{n/\log n})$  iterations
- Establishes Ramsey-Theory conjecture by Fox et.al. (cf. next slides)

# Ramsey Theory with $r \geq 2$ colors

$G \rightarrow (H)_r \iff$  any  $r$ -coloring of  $E(G)$  has monochromatic copy of  $H$

Ramsey theory  $\triangleq$  studying properties of " $r$ -Ramsey minimal graphs"

$\mathcal{M}_r(H) :=$  all graphs  $G$  that are  $r$ -Ramsey minimal for  $H$   
(i.e.,  $G \rightarrow (H)_r$  and  $G' \not\rightarrow (H)_r$  for all  $G' \subsetneq G$ )

- $\min_{G \in \mathcal{M}_r(K_k)} v(G) =$  Ramsey number
- $\min_{G \in \mathcal{M}_r(K_k)} e(G) =$  Size Ramsey number

Minimum degree of  $r$ -Ramsey minimal graphs (Burr, Erdős, Lovász 1976)

$s_r(H) := \min_{G \in \mathcal{M}_r(H)} \delta(G)$

- $s_2(K_k) = (k-1)^2$ : Burr, Erdős, Lovász (1976)
- $s_2(H) = 2\delta(H) - 1$ : for many bipartite  $H$  (trees,  $K_{a,b}$ , etc)  
Fox, Lin (2006) + Szabó, Zumstein, Zürcher (2010)
- $s_r(K_k) = \tilde{\Theta}_k(r^2)$ : Fox, Grinshpun, Liebenau, Person, Szabó (2015)

# Ramsey Conjecture of Fox et.al.

Minimum degree of minimal  $r$ -Ramsey graphs (Burr, Erdős, Lovász 1976)

$$s_r(K_k) := \min_{G \in \mathcal{M}_r(K_k)} \delta(G)$$

- $cr^2 \log r \leq s_r(K_3) \leq Cr^2(\log r)^2$  by FGLPS (2015)

Conjecture (Fox, Grinshpun, Liebenau, Person, Szabo, 2015)

$$s_r(K_3) = O(r^2 \log r)$$

- They suggested to pack  $G_i$  sequentially via  $\Delta$ -free process (their weaker upper bound relies on sequential LLL-argument)

Conj. True (G., Warnke, 2020): corollary of our main packing result

$$\text{Implies } s_r(K_3) = \Theta(r^2 \log r)$$

- For technical reasons: use *semi-random variation* of  $\Delta$ -free process

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# Glimpse of the proof

**Main-Technical-Result:** find random-like  $\Delta$ -free subgraph  $G \subseteq H$

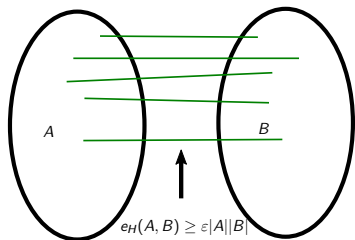
Let  $\varrho := \sqrt{\beta(\log n)/n}$  and  $s := C_\varepsilon \sqrt{n \log n}$ . If  $H \subseteq K_n$  is such that

$$e_H(A, B) \geq \varepsilon |A||B|$$

for all disjoint sets  $A, B$  of size  $s$ , then we can find  $\Delta$ -free  $G \subseteq H$  with

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**Implies packing result:** (maintaining  $e_{H_i}(A, B)$  bounds inductively)

- Start with  $H_0 = K_n$
- Sequentially choose  $G_i \subseteq H_i$  and set  $H_{i+1} = H_i \setminus G_i$

$$e_{H_i}(A, B) = (1 - (1 \pm \delta)\varrho)^i |A||B|$$

- Stop when  $e_{H_i}(A, B) \approx \varepsilon |A||B|$  holds

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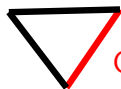
**Proof based on semi-random variation of  $\Delta$ -free process:**

- Do *not* require degree/codegree regularity of  $H$
- ‘Self-stabilization’ mechanism built into process (to control errors)
- Tools: Bounded-Differences-Ineq. and Upper-Tail-Ineq. of Warneke

# Semi-random construction of $\Delta$ -free subgraph

To construct triangle-free  $T_j$ , we iteratively keep track of

- $E_j$ : “random” set of edges
- $T_j \subseteq E_j$ :  $\Delta$ -free and  $|T_j| \approx |E_j|$
- $O_j \subseteq \{\text{all } e \notin E_j \text{ that don't form a } \Delta \text{ with any two edges of } E_j\}$



Closed edge

Idea of each step

- (1) Generate few random edges  $\Gamma_{j+1} \subseteq O_j$
- (2) Alteration: find  $\Gamma'_{j+1} \subseteq \Gamma_{j+1}$  s.t.  $T_{j+1} = T_j \cup \Gamma'_{j+1}$  remains  $\Delta$ -free
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- Start with  $O_0 = E(H)$  for the dense host graph  $H$ .  $E_0 = T_0 = \emptyset$

## Definition of $\Gamma_{j+1}$ and $E_{j+1}$

- $\Gamma_{j+1} \subseteq O_j$ :  $p$ -random subset of  $O_j$
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## Why can we ensure $|\Gamma'_{j+1}| \approx |\Gamma_{j+1}|$ ?

- $\Gamma_{j+1}$  small  $\Rightarrow$  very few new  $\Delta$ 's created in  $E_j \cup \Gamma_{j+1}$
- hence removal of few edges destroys all new  $\Delta$ 's

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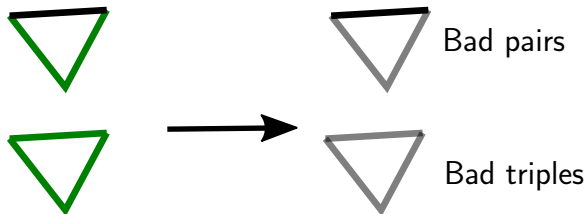


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Alteration to destroy new  $\Delta$ 's:  $\Gamma'_{j+1} = \Gamma_{j+1} \setminus \mathcal{D}_{j+1}$

$\mathcal{D}_{j+1}$  = edges of a maximal edge-disjoint collection of bad pairs/triples

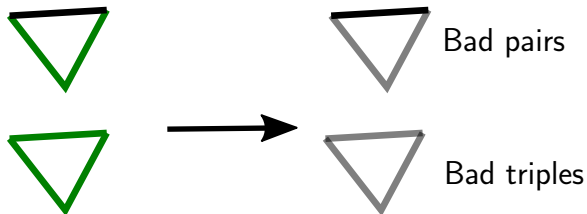
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# Open edges: effect of closed edges

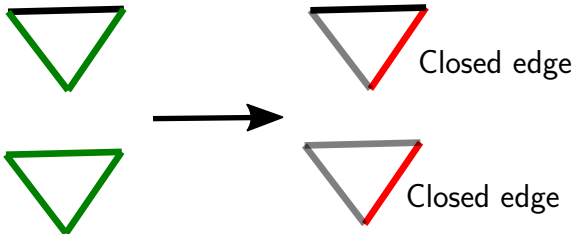
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## Updating "open edges" that can still be added

$$O_{j+1} = O_j \setminus (\Gamma_{j+1} \cup \{\text{"closed edges"}\}) \cup \{\text{extra edges for technical reasons}\}.$$

"Closed edge" forms a triangle with two edges in  $E_{j+1} = E_j \cup \Gamma_{j+1}$ .



## Open edges: self-stabilization mechanism

Updating "open edges" that can still be added

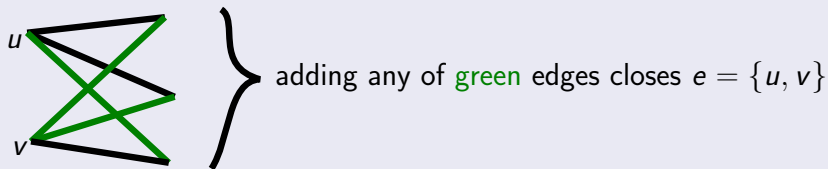
$$O_{j+1} = O_j \setminus (\Gamma_{j+1} \cup \{\text{"closed edges"}\}) \cup \{\text{extra random edges}\}.$$

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$Y_e(j) = \#$  edges whose addition to  $E_{j+1}$  will close  $e$



Self-stabilization: make  $\mathbb{P}(\text{closed})$  equal for all  $e$  (independent of history)

$$\mathbb{P}(e \text{ not closed in next step of iteration}) \approx (1 - p)^{|Y_e(j)|}$$

$$\mathbb{P}(e \text{ not (closed or extra edge)}) \approx (1 - p)^{|Y_e(j)|} (1 - q_e) \stackrel{!}{=} \text{same for all } e$$

# Semi-random construction of $\Delta$ -free subgraph

To construct triangle-free  $T_j$ , we iteratively keep track of

- $E_j$ : "random" set of edges
- $T_j \subseteq E_j$ :  $\Delta$ -free and  $|T_j| \approx |E_j|$
- $O_j \subseteq \{\text{all } e \notin E_j \text{ that don't form a } \Delta \text{ with any two edges of } E_j\}$

## Idea of each step

- (1) Generate few random edges  $\Gamma_{j+1} \subseteq O_j$
- (2) Alteration: find  $\Gamma'_{j+1} \subseteq \Gamma_{j+1}$  s.t.  $T_{j+1} = T_j \cup \Gamma'_{j+1}$  remains  $\Delta$ -free
- (3) Update  $O_{j+1} \subseteq O_j \setminus \Gamma_{j+1}$

# Number of edges between two large sets

Assume we can show

$$|O_j(A, B)| \approx q_j |A| |B|, \text{ where } q_j = \Psi'(j\sigma), \text{ for } O_0 = H = K_n.$$

Use  $p = \sigma/\sqrt{n}$ , then we can approximate  $|T_j(A, B)|$

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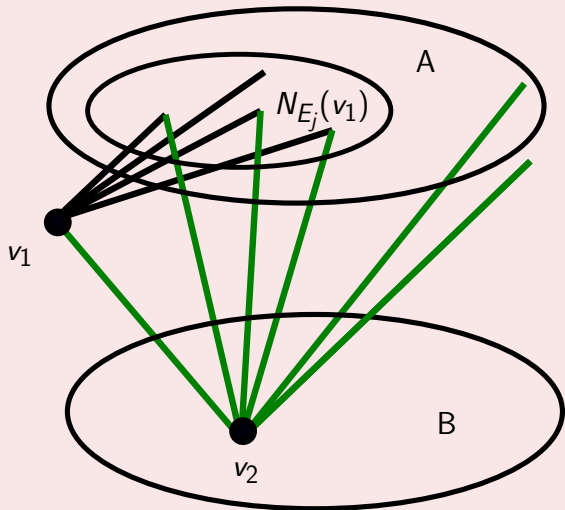
$$\begin{aligned} |T_J(A, B)| &= \sum_{0 \leq j < J} |T_{j+1}(A, B) \setminus T_j| \approx \sum_{0 \leq j < J} |\Gamma_{j+1}(A, B)| \\ &\approx \sum_{0 \leq j < J} p |O_j(A, B)| \approx \frac{1}{\sqrt{n}} \sum_{0 \leq j < J} \sigma q_j \cdot |A| |B| \\ &\approx \frac{1}{\sqrt{n}} \int_0^{J\sigma} \Psi'(x) dx \cdot |A| |B| \approx \frac{\Psi(J\sigma)}{\sqrt{n}} |A| |B| \\ &\approx \frac{\sqrt{\beta(\log n)}}{\sqrt{n}} |A| |B| = \varrho |A| |B| \end{aligned}$$



# A technical difficulty

Difficulty of tracking  $|O_j(A, B)|$

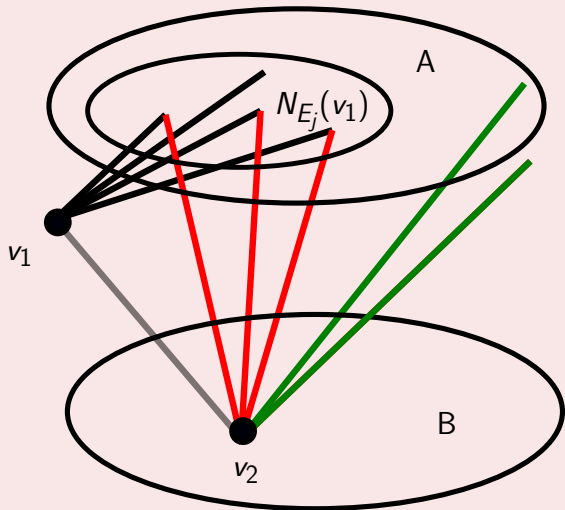
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# Summary

G., Warnke (2020): almost packing of nearly optimal  $R(3, t)$  graphs

Given  $\varepsilon > 0$ , we find edge-disjoint graphs  $(G_i)_{i \in \mathcal{I}}$  with  $G_i \subseteq K_n$  such that

- (a) each  $G_i$  is  $\Delta$ -free with  $\alpha(G_i) \leq C_\varepsilon \sqrt{n \log n}$
- (b) the union of the  $G_i$  contains  $\geq (1 - \varepsilon) \binom{n}{2}$  edges

## Remarks

- Natural algorithmic packing version of Kim's  $R(3, t)$  construction
- Establishes  $s_r(K_3) = \Theta(r^2 \log r)$  asymptotics conjectured by Fox et.al.

## Questions

- Further applications of the  $K_3$ -free packing result?
- Generalization of packing-result to  $K_k$ -free graphs worth effort?

## Reference

- He Guo, Lutz Warnke, *Packing nearly optimal Ramsey  $R(3, t)$  graphs*, *Combinatorica* **40**, 63–103 (2020)