

# Cumulants and triangles in Erdős-Rényi random graphs

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partially joint work with Pierre-Loïc Méliot (Orsay)  
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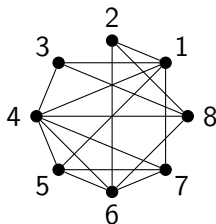


Universität  
Zürich<sup>UZH</sup>

## A problem in random graphs

Erdős-Rényi model of random graphs  $G(n, p)$ :

- $G$  has  $n$  vertices labelled  $1, \dots, n$ ;
- each edge  $\{i, j\}$  is taken independently with probability  $p$ ;

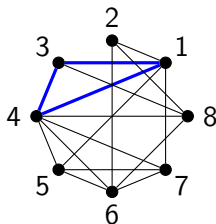


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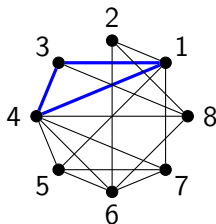
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Describe asymptotically the fluctuations of the number  $T_n$  of triangles.

Answer (Ruciński, 1988)

The fluctuations are asymptotically Gaussian.

# Outline

- 1 Intro: cumulant method for number of triangles in  $G(n, p)$
- 2 First extension: stronger conclusion
- 3 Second extension: weaker hypothesis
- 4 Ideas of proof

## A good tool for that: mixed cumulants

- the  $r$ -th mixed cumulant  $\kappa_r$  of  $r$  random variables is a specific  $r$ -linear symmetric polynomial in joint moments. Examples:

$$\begin{aligned}\kappa_1(X) &:= \mathbb{E}(X), & \kappa_2(X, Y) &:= \text{Cov}(X, Y) = \mathbb{E}(XY) - \mathbb{E}(X)\mathbb{E}(Y) \\ \kappa_3(X, Y, Z) &:= \mathbb{E}(XYZ) - \mathbb{E}(XY)\mathbb{E}(Z) - \mathbb{E}(XZ)\mathbb{E}(Y) \\ &\quad - \mathbb{E}(YZ)\mathbb{E}(X) + 2\mathbb{E}(X)\mathbb{E}(Y)\mathbb{E}(Z).\end{aligned}$$

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- if the variables can be split in two mutually independent sets, then the cumulant vanishes.
- if, for each  $r \neq 2$ , the sequence  $\kappa_r(X_n)$  converges towards 0 and if  $\text{Var}(X_n)$  has a limit, then  $X_n$  converges in distribution towards a Gaussian law.



## Application to the number of triangles

$$T_n = \sum_{\Delta=\{i,j,k\}\subset[n]} B_{\Delta}, \text{ where } B_{\Delta}(G) = \begin{cases} 1 & \text{if } G \text{ contains the triangle } \Delta; \\ 0 & \text{otherwise.} \end{cases}$$

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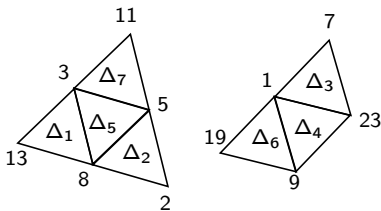
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Example:

$\{\Delta_1, \Delta_2, \Delta_5, \Delta_7\}$  is independent from  $\{\Delta_3, \Delta_4, \Delta_6\}$ .

Reminder: presence of different edges are independent events.

$$\kappa_{\ell}(B_{\Delta_1}, \dots, B_{\Delta_7}) = 0.$$

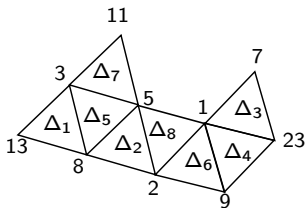


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Example:

$$\kappa_{\ell}(B_{\Delta_1}, \dots, B_{\Delta_8}) \neq 0.$$

This configuration contributes to the sum. Call it **configuration of dependent triangles**.

**Lemma:** Such a configuration has at most  $\ell + 2$  vertices (here  $\ell = 8$ ).

## Bound on the cumulant

$$\kappa_\ell(T_n) = \sum_{\Delta_1, \dots, \Delta_\ell} \kappa_\ell(B_{\Delta_1}, \dots, B_{\Delta_\ell}).$$

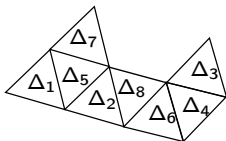
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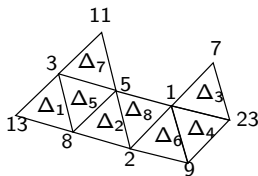


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- only configurations of dependent triangles contribute to the sum ;
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- each configuration can be labelled in at most  $n^{\ell+2}$  ways.

## Bound on the cumulant

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Fact 1: number of non-zero terms is smaller than  $C_\ell n^{\ell+2}$ .

Fact 2 (easy): each non-zero terms is bounded by  $C'_\ell$ .

Conclusion:

$$|\kappa_\ell(T_n)| = O_\ell(n^{\ell+2})$$

# The central limit theorem for triangles

Proposition (Leonov, Shirryaev, 1955)

If  $X_1, \dots, X_\ell$  can be split into two sets of mutually independent variables, then

$$\kappa_\ell(X_1, \dots, X_\ell) = 0$$

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Corollary (Ruciński, 1988)

$$\widetilde{T}_n := \frac{T_n - \mathbb{E}(T_n)}{\sqrt{\text{Var}(T_n)}} \rightarrow \mathcal{N}(0, 1)$$

Proof:  $\text{Var}(T_n) \approx n^4$  thus,  $\kappa_\ell(\widetilde{T}_n) = \kappa_\ell(T_n) / \text{Var}(T_n)^{\ell/2} = O_\ell(n^{2-\ell})$ .

# Transition

First extension: stronger conclusion

# Statement

Theorem (F., Méliot, Nighekbali, 2014)

Let  $X_1, \dots, X_\ell$  be random variables with finite moments of order  $\ell$ ,

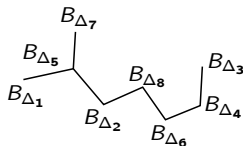
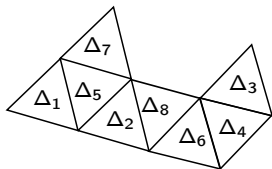
$$|\kappa_\ell(X_1, \dots, X_\ell)| \leq 2^{\ell-1} \|X_1\|_\ell \cdots \|X_\ell\|_\ell \cdot \text{ST}(G_{\text{dep}}(X_1, \dots, X_\ell)),$$

where  $\text{ST}(G_{\text{dep}}(X_1, \dots, X_\ell))$  is the number of **spanning trees** of a **dependency graph** of  $X_1, \dots, X_\ell$ .

A dependency graph for the list  $(B_{\Delta_1}, \dots, B_{\Delta_\ell})$ :

$$B_{\Delta_i} \sim B_{\Delta_j} \Leftrightarrow \Delta_i \text{ and } \Delta_j \text{ share an edge}$$

Example:



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Corollary (FMN, 2014)

There exists an absolute constant  $C$  such that

$$|\kappa_\ell(T_n)| \leq (C\ell)^\ell n^{\ell+2}$$

Naive bound:  $(C\ell)^{3\ell} n^{\ell+2}$

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Let  $X_n = (T_n - \mathbb{E}(T_n))/n^{5/3}$ . Then (uniformly on compacts of  $\mathbb{C}$ ),

$$\mathbb{E}\left(e^{z X_n}\right) = \exp\left(n^{2/3} z^2/2\right) \exp(L_p z^3/6)(1 + o(1)).$$

( $L_p$  is an explicit constant that depends only on  $p$ ).



# Mod-Gaussian convergence and consequences

## Corollary (FMN,2014)

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This type of estimates for the Laplace transform is called **mod-Gaussian convergence** (Kowalski, Nikeghbali). It implies:

- a central limit theorem (here, we recover the result of Ruciński);
- description of the **normality zone** and asymmetry of deviations at the edge of this zone;
- **speed of convergence** in the central limit theorem (here of order  $O(1/n)$ ; we recover a result of Krokowski, Reichenbachs and Thaele, 2015).

# Discussion

Our result applies to **sum of mostly independent variables** (*i.e.* most of the variables are independent)

- Number of copies of a given subgraph;
- Number of arithmetic progression in a random subset of  $\{1, \dots, n\}$ ;
- Number of descents/inversions in random permutations. . .

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The **normality zone** and **speed of convergence** results bound applies in the general context of mod-Gaussian convergence. Again, lots of examples:

- determinant of unitary random matrices,
- number of zeros of a complex analytic function with random coefficients,
- Curie-Weiss model in statistical physics. . .

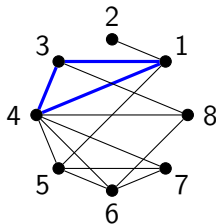
# Transition

Second extension: weaker hypothesis  
(work in progress)

# Erdős-Rényi model $G(n, M)$

- $G$  has  $n$  vertices labelled  $1, \dots, n$ ;
- The edge-set of  $G$  is taken uniformly among all possible edge-sets of **cardinality  $M$** .

Example with  $n = 8$  and  $M = 14$

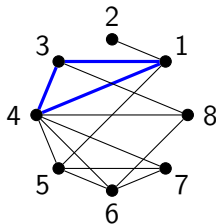


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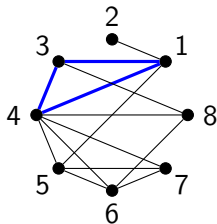
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## Question

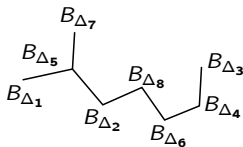
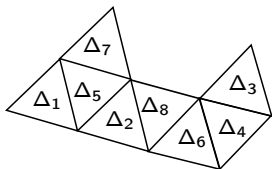
Let  $M_n = \lfloor p\binom{n}{2} \rfloor$ , with  $p$  fixed.

Describe asymptotically the fluctuations of the number  $T_n$  of triangles in  $G(n, M_n)$ .

### Proposition (F., > 2015)

Let  $\Delta_1, \dots, \Delta_\ell$  be triangles. Define  $G_{\Delta_1, \dots, \Delta_\ell}$  as before and denote  $r$  its number of connected components. Then

$$\kappa_\ell(B_{\Delta_1}, \dots, B_{\Delta_\ell}) \leq \frac{C_\ell}{M_n^{r-1}}.$$



This graph is **not a dependency graph any more**, but the more connected components it has, the smaller the cumulant is.



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### Corollary

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### Corollary

$$\frac{T_n - \mathbb{E}(T_n)}{\sqrt{\text{Var}(T_n)}} \rightarrow \mathcal{N}(0, 1)$$

First proved by Janson in 1994 using a coupling with  $G(n, p_n)$ .

## “Weak dependency graph”: a general theory?

Other examples, where the order of magnitude of joint cumulants depends on the number of components of some underlying graph:

- 1 Patterns in **random words** with a fixed number of occurrences of each letter;
- 2 Images of distinct integers in a **random permutation** of size  $n$  are  $1/n$ -dependent;
- 3 Indicators of particles that can jump in the steady state of the **symmetric simple exclusion process**;
- 4 Entries in **Haar-distributed orthogonal matrices**;
- 5 Relations in **random set partitions**.

Yields various central limit theorems in all cases (work in progress!)

# Transition

Ideas of proof

## Moment-cumulant relation

Mixed cumulants can be expressed in terms of mixed moments:

$$\kappa(X_1, \dots, X_r) = \sum_{\pi} \mu(\pi) M_{\pi},$$

where

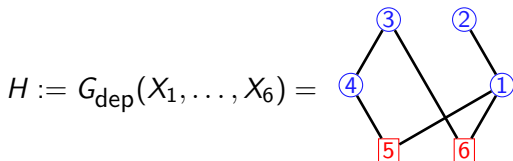
- $\pi$  runs over **set-partitions** of  $[\ell]$ ,
- $\mu(\pi) = \mu(\pi, \{[\ell]\})$  is the Möbius function of the set-partition poset (it is explicit but we will only use  $\sum_{\text{interval}} \mu(\pi) = 0$ ),
- $M_{\pi} = \prod_{B \in \pi} \mathbb{E}[\prod_{i \in B} X_i]$ .

Example:

$$\begin{aligned} M_{\{\{1,3\},\{2,4\}\}} &= \mathbb{E}(X_1 X_3) \mathbb{E}(X_2 X_4) \\ \kappa_3(X, Y, Z) &= \mathbb{E}(XYZ) - \mathbb{E}(XY)\mathbb{E}(Z) - \mathbb{E}(XZ)\mathbb{E}(Y) \\ &\quad - \mathbb{E}(YZ)\mathbb{E}(X) + 2\mathbb{E}(X)\mathbb{E}(Y)\mathbb{E}(Z). \end{aligned}$$

# Using independence to simplify $M_\pi$

Example: take  $\pi = \{\{1, 2, 3, 4\}, \{5, 6\}\}$  and



$$\begin{aligned} \text{Then } M_\pi &:= \mathbb{E}(X_1 X_2 X_3 X_4) \mathbb{E}(X_5 X_6) \\ &= \mathbb{E}(X_1 X_2) \mathbb{E}(X_3 X_4) \mathbb{E}(X_5) \mathbb{E}(X_6) \\ &= M_{\{\{1,2\}, \{3,4\}, \{5\}, \{6\}\}}. \end{aligned}$$

In general,  $M_\pi = M_{\phi_H(\pi)}$ , with obvious definition of  $\phi_H(\pi)$ :  
 “replace each part  $\pi_i$  of  $\pi$  by the connected components of  $H[\pi_i]$ ”.

# Rewriting the summation

$$\begin{aligned}\kappa(X_1, \dots, X_r) &= \sum_{\pi} \mu(\pi) M_{\pi} = \sum_{\pi} \mu(\pi) M_{\phi_{\mathbf{H}}(\pi)} \\ &= \sum_{\pi'} M_{\pi'} \left( \sum_{\substack{\pi \text{ s.t.} \\ \phi_{\mathbf{H}}(\pi) = \pi'}} \mu(\pi) \right)\end{aligned}$$

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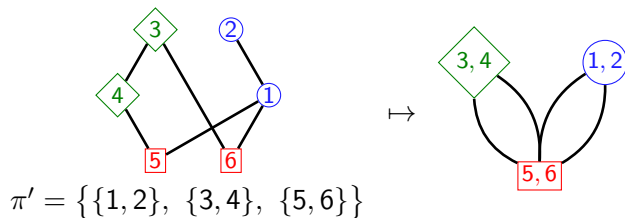
- $\phi_H(\pi) = \pi' \Rightarrow$  for all part  $\pi'_i$  of  $\pi'$ , the induced graph  $H[\pi'_i]$  is connected.
- if so, we have to compute

$$\alpha_H^{\pi'} := \sum_{\substack{\pi \text{ s.t.} \\ \phi_H(\pi) = \pi'}} \mu(\pi).$$



# Bounding $\alpha_H^{\pi'}$

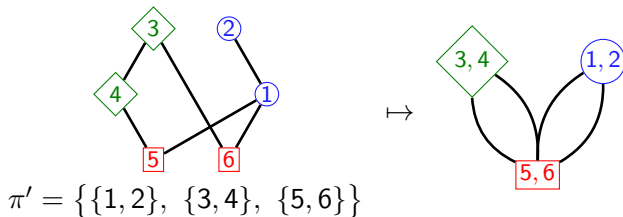
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**Lemma**

$$\left| \alpha_H^{\pi'} \right| \leq \text{ST}(H/\pi').$$

In the example:  $\text{ST}(H/\pi') = 4$ .

# Bounding everything

Reminder:

$$\kappa(X_1, \dots, X_\ell) = \sum_{\pi'} M_{\pi'} \alpha_H^{\pi'}$$

where the sum runs over set-partition  $\pi'$  such that the induced graphs  $H[\pi'_i]$  are connected.

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We have the following inequalities

$$|M'_{\pi}| \leq \|X_1\|_{\ell} \cdots \|X_{\ell}\|_{\ell} \quad (\text{Hölder inequality});$$

$$|\alpha_H^{\pi'}| \leq \text{ST}(H/\pi');$$

$$\mathbf{1}_{H[\pi'_i] \text{ connected}} \leq \text{ST}(H[\pi'_i])$$

# Bounding everything

Reminder:

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Thus

$$|\kappa(X_1, \dots, X_\ell)| \leq \|X_1\|_\ell \cdots \|X_\ell\|_\ell \left[ \sum_{\pi'} \text{ST}(H/\pi') \left( \prod_i \text{ST}(H[\pi'_i]) \right) \right]$$

# A combinatorial identity

## Lemma

$$2^{\ell-1} \text{ST}(H) = \sum_{\pi'} \text{ST}(H/\pi') \left( \prod_i \text{ST}(H[\pi'_i]) \right),$$

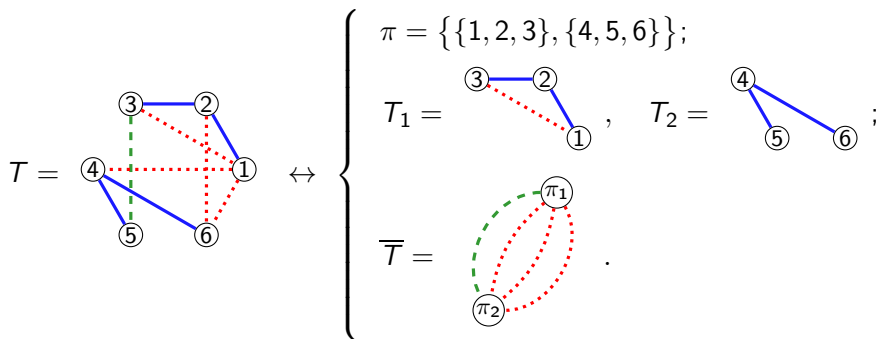
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## Progress report

We have “proved”:

### Theorem

Let  $X_1, \dots, X_\ell$  be random variables with finite moments of order  $\ell$ ,

$$|\kappa_\ell(X_1, \dots, X_\ell)| \leq 2^{\ell-1} \|X_1\|_\ell \cdots \|X_\ell\|_\ell \cdot \text{ST}(G_{\text{dep}}(X_1, \dots, X_\ell)).$$

Next step:

### Corollary

There exists an absolute constant  $C$  such that

$$|\kappa_\ell(T_n)| \leq (C\ell)^\ell n^{\ell+2}.$$

## A sharp bound on cumulants of $T_n$

Recall that  $\kappa_\ell(T_n) = \sum_{\Delta_1, \dots, \Delta_\ell} \kappa_\ell(B_{\Delta_1}, \dots, B_{\Delta_\ell})$ .

Thus

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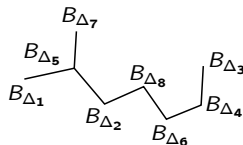
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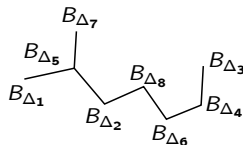
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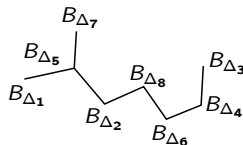
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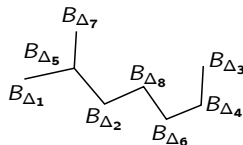
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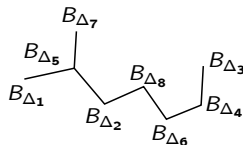
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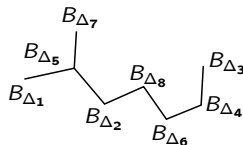
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• ...  $|\kappa_\ell(T_n)| \leq 2^{\ell-1} \ell^{\ell-2} \binom{n}{3} (3n-6)^{\ell-1} \leq (6\ell)^{\ell} n^{\ell+2}$

## Mod-Gaussian convergence

Let  $X_n = (T_n - \mathbb{E}(T_n))/n^{5/3}$ , then

$$\begin{aligned} \log \mathbb{E}(\exp(zX_n)) &= \sum_{\ell \geq 2} \kappa_\ell(X_n) z^\ell / \ell! \\ &= n^{2/3} \sigma^2 z^2 / 2 + L z^3 / 6 + \underbrace{\sum_{\ell \geq 4} n^{5/3} \kappa_\ell(T_n) z^\ell / \ell!}_{\text{call it } R} \end{aligned}$$

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almost as if  $X_n$  would be the sum of  $n^{2/3}$  independent standard Gaussian.  
 $\rightarrow$  results on normality zone and speed of convergence follow by adaptation of standard techniques (Berry-Esseen lemma, change of probability measure).

## A word on the proof of the second extension (1/2)

First consider **edge-disjoint triangles**  $\Delta_1, \dots, \Delta_\ell$ .

Then, for  $A \subset \{1, \dots, \ell\}$ ,

$$M_A := \mathbb{E} \left( \prod_{i \in A} B_{\Delta_i} \right) = \frac{(M_n)_{3\ell}}{\binom{n}{2}_{3\ell}}$$

Notation:  $(x)_k = x(x-1)(x-2)\dots(x-k+1)$ .

We want to prove that

$$\kappa(B_{\Delta_1}, \dots, B_{\Delta_\ell}) = O(M_n^{-\ell+1})$$

This case is already difficult!

## A word on the proof of the second extension (2/2)

### Lemma

Define  $T_B$  by  $M_A = \prod_{B \subset A} (1 + T_B)$ , i.e.  $T_B = -1 + \prod_{A \subset B} M_A^{(-1)^{|B|-|A|}}$ .

$$\text{Then } T_B = O(M_n^{-|B|+1}).$$

Proof: elementary analysis.

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$$\kappa_\ell(X_1, \dots, X_\ell) = \sum_{\pi} \mu(\pi) \left[ \prod_{A \in \pi} \prod_{B \subset A} (1 + T_B) \right].$$

Expand and exchange summation

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Fact: the sum [...] vanish unless the monomial is  $O(M_n^{\ell-1})$ . □



# Open questions

- What about  $G(n, p_n)$  with  $p_n \rightarrow 0$  and  $np_n \rightarrow \infty$ . One has:

$$\kappa_\ell(T_n) \leq C_\ell n^3 p_n^2 \max(np_n^2, 1)^{\ell-1} \quad (\text{Mikhailov, 1991});$$

$$\kappa_\ell(T_n) \leq (C\ell)^\ell n^{\ell+2} p_n^3 \quad (\text{FMN, 2014}).$$

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- Future work: Stein's method/Lovász local lemma with weak dependency graphs?