

# On a phase transition of the random intersection graph: supercritical region

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2015 Combinatorics Workshop  
NIMS  
July 16, 2015

This is joint work with [Jeong Han Kim](#) and [Joochan Na](#) (KIAS).

# Motivation

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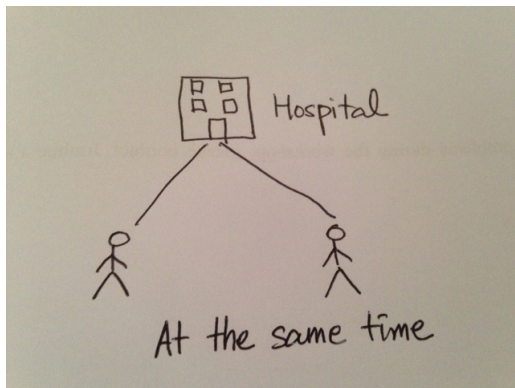
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What is the proper network (or graph) model explaining epidemic of Mers?

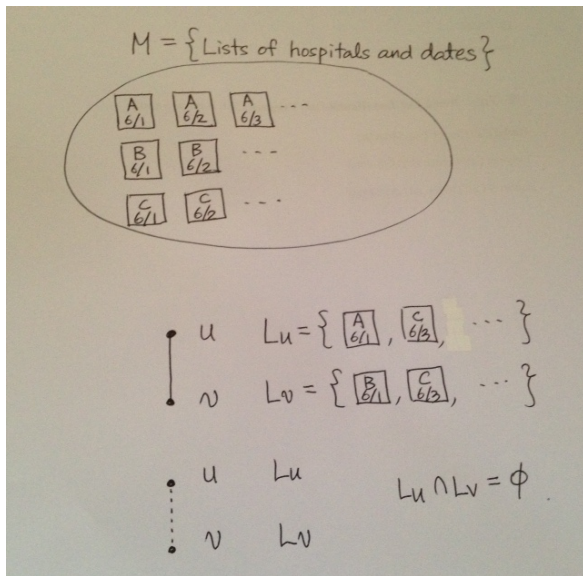
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## Feature of Mers in Korea





## Graph model about epidemic of Mers



## Sec 1) Definition

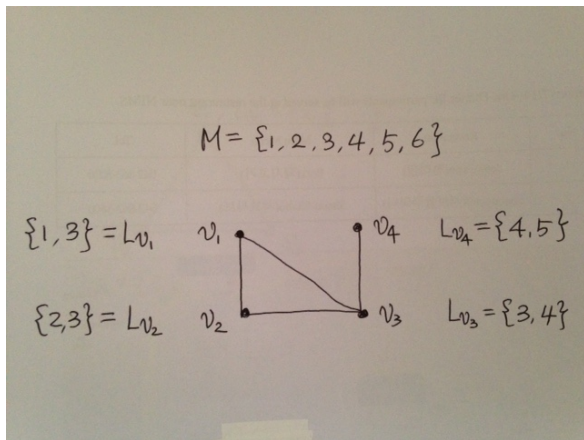
- $V := \{v_1, \dots, v_n\}$
- $\{L_1, \dots, L_n\}$ : a collection of sets

### Definition (Intersection graph)

The *intersection graph* on  $V$  generated by  $\{L_1, \dots, L_n\}$  is the graph on  $V$  in which

$$v_i \sim v_j \quad \text{if and only if} \quad L_i \cap L_j \neq \emptyset.$$

# Example



# Random intersection graph

Definition (Random Intersection graph  $G(n, m; p)$ )

- $M$ : a set of size  $m$ .
- $L_i$ : a random subset obtained by choosing each element in  $M$  independently with probability  $p$ .

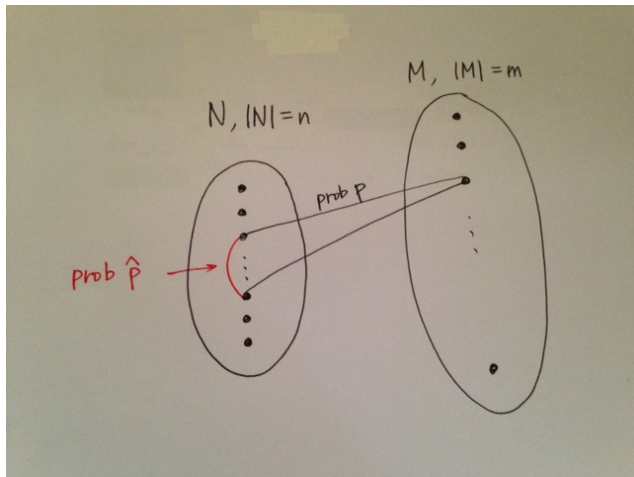
# Random intersection graph

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- $M$ : a set of size  $m$ .
- $L_i$ : a random subset obtained by choosing each element in  $M$  independently with probability  $p$ .
- The *random intersection graph*  $G(n, m; p)$  is the intersection graph generated by i.i.d.  $L_i$  as above.

It was defined by [Karoński, Scheinerman, and Singer-Cohen \(1999\)](#).

# Visualization: Random bipartite graph



# Application

- 1 A random intersection graph has received a lot of attention because of a great diversity of applications:
  - Epidemic
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- 1 A random intersection graph has received a lot of attention because of a great diversity of applications:
  - Epidemic
  - Network user profiling
  - Analysis of complex networks.
- 2 The special case when  $L_i$ 's are uniformly distributed as subsets of  $M$  of the same size has been applied to security of wireless sensor networks.



## Question

When is  $G(n, m; p)$  essentially the same as the binomial random graph  $G(n, \hat{p})$  with the same expected number of edges?

**Remark:**  $\hat{p} := 1 - (1 - p^2)^m$ .

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Distance between two random graphs: **Total variation.**

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## Notion

Distance between two random graphs: **Total variation**.

## Definition

The **total variation** between two random graphs  $X$  and  $Y$  is defined by

$$\text{TV}(X, Y) := \frac{1}{2} \sum_G \left| \Pr[X = G] - \Pr[Y = G] \right|,$$

where the sum is over all possible graphs  $G$  of  $X$  and  $Y$ .

## Sec 2) Previous results

### Observation

Let  $\omega \rightarrow \infty$  as  $n \rightarrow \infty$ .

① If  $p \leq \frac{1}{\omega n \sqrt{m}}$ ,

then two random graphs are the **empty graph** with high probability.

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### Assumption

$$\frac{1}{\omega n \sqrt{m}} \leq p \leq \sqrt{\frac{2 \ln n + \omega}{m}}.$$

## Proposition

If  $m \ll n^3$ , then

$$\text{TV}\left(G(n, m; p), G(n, \hat{p})\right) \rightarrow 1.$$

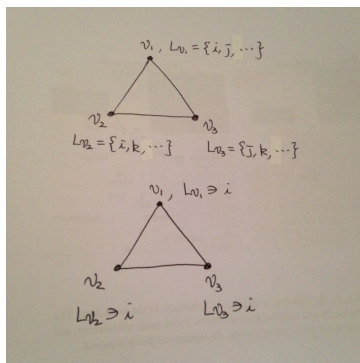
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Theorem (Fill, Scheinerman and Singer-Cohen (2000))

If  $m = n^\alpha$  and  $\alpha > 6$ , then

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## Theorem (Rybarczyk (2011))

If  $m = n^\alpha$  and  $3 \leq \alpha \leq 6$ , for any monotone property  $\mathcal{P}$ ,  
 $\Pr[G(n, m; p) \in \mathcal{P}]$  is similar to  $\Pr[G(n, \hat{p}) \in \mathcal{P}]$   
 (with a technical statement).

## Question

What is the **smallest** constant  $\alpha$  such that  
for  $m = n^\alpha$  and any  $p = p(n, m)$ ,

$$\text{TV}\left(G(n, m; p), G(n, \hat{p})\right) = o(1)?$$

## Previous result

$$3 \leq \alpha \leq 6.$$

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What is the **smallest** constant  $\alpha$  such that for  $m = n^\alpha$  and any  $p = p(n, m)$ ,

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$$3 \leq \alpha \leq 6.$$

## Main Theorem (Kim, Lee, Na (2015+))

For  $m \gg n^4$  and  $0 \leq p \leq 1$ ,

$$\text{TV}\left(G(n, m; p), G(n, \hat{p})\right) = o(1).$$

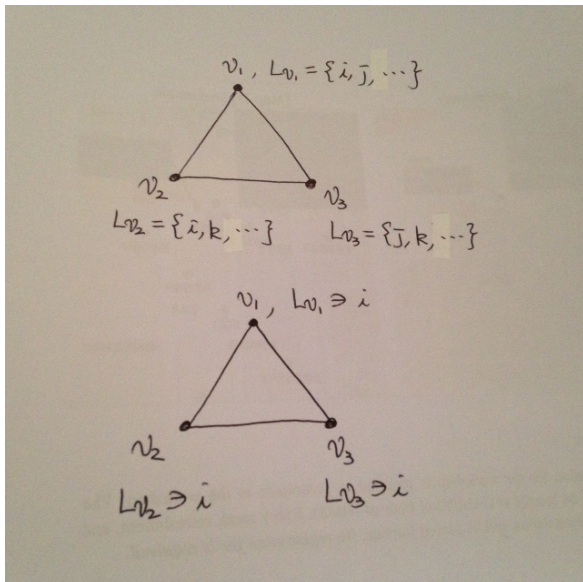
\* We believe that **4** in the exponent is tight.

# Artifact triangles

## Definition

An **artifact triangle** is a triangle formed by the same element in  $M$ .

- 1 Fill, Scheinerman and Singer-Cohen (2000):  
the case when there is **no artifact triangle**.
- 2 Kim, Lee and Na (2015+):  
the case when there are **not so many artifact triangles**.

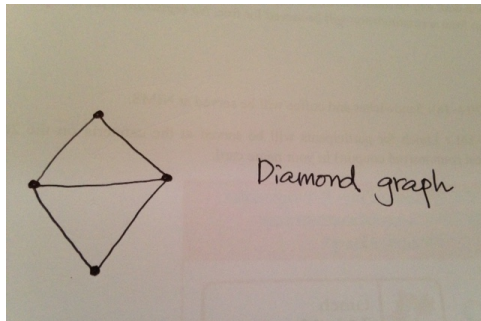


# Key object

Key object: Diamond graph

- A **diamond graph** =  $K_4$  minus one edge.
- The **number of diamond graphs with two artifact triangles** in  $G(n, m; p)$  is **small** iff

$$G(n, m; p) \sim G(n, \hat{p}).$$



## Sec 3) Outline of Proof of Main Theorem

Recall: Main Theorem (Kim, Lee, Na (2015+))

For  $m \gg n^4$  and  $0 \leq p \leq 1$ ,

$$\mathrm{TV}\left(G(n, m; p), G(n, \hat{p})\right) = o(1).$$

Remark (essentially by Rybarczyk)

$G(n, m; p)$  is approximated by a random graph  $G(n, (p_2, p_3, p_4))$ .



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**Remark:** Why the  $p_k$ ?

- For  $a \in M$ ,  $V_a := \{v : a \in L_v\}$ .
- For a fixed  $k$ -subset  $U \subset V$ ,

$$\Pr \left[ \exists a \in M \text{ s.t. } V_a = U \right] = 1 - (1 - p^k(1-p)^{n-k})^m.$$

Recall: Main Theorem (Kim, Lee, Na (2015+))

For  $m \gg n^4$  and  $0 \leq p \leq 1$ ,

$$\text{TV}\left(G(n, m; p), G(n, \hat{p})\right) = o(1).$$

Key Lemma (Kim, Lee, Na (2015+))

For  $m \gg n^4$  and  $0 \leq p \leq \left(\frac{3 \log n}{m}\right)^{1/2}$ ,

$$\text{TV}\left(G(n, (p_2, p_3, p_4)), G(n, p_2)\right) = o(1).$$

## Sec 4) Proof of Lemma

$$\begin{aligned}
& \text{TV}\left(G(n, (p_2, p_3, p_4)), G(n, p_2)\right) \\
& := \frac{1}{2} \sum_G \left| \Pr[X = G] - \Pr[Y = G] \right| \\
& = \sum_{G \in \mathcal{G}} \left( \Pr[G(n, p_2) = G] - \min \left\{ \Pr[G(n, (p_2, p_3, p_4)) = G], \Pr[G(n, p_2) = G] \right\} \right).
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 \end{aligned}$$

In order to show that  $\text{TV}\left(G(n, (p_2, p_3, p_4)), G(n, p_2)\right)$  is small, it suffices to show that

$$\Pr[G(n, (p_2, p_3, p_4)) = G] \geq (1 - O(\epsilon)) \Pr[G(n, p_2) = G].$$



$$\begin{aligned}
& \Pr[G(n, (p_2, p_3, p_4)) = G] \\
&= \sum_{\substack{Q \subseteq \mathcal{H}_4(G) \\ T \subseteq \mathcal{H}_3(G)}} \Pr[\mathcal{H}_4(n, p_4) = Q, \mathcal{H}_3(n, p_3) = T, G(n, p_2) = G] \\
&= \sum_{\substack{Q \subseteq \mathcal{H}_4(G) \\ T \subseteq \mathcal{H}_3(G)}} p_4^{|Q|} (1-p_4)^{\binom{n}{4} - |Q|} p_3^{|T|} (1-p_3)^{\binom{n}{3} - |T|} p_2^{|G| - |K(Q) \cup K(T)|} (1-p_2)^{\binom{n}{2} - |G|} \\
&= \Pr[G(n, p_2) = G] \sum_{\substack{Q \subseteq \mathcal{H}_4(G) \\ T \subseteq \mathcal{H}_3(G)}} p_4^{|Q|} (1-p_4)^{\binom{n}{4} - |Q|} p_3^{|T|} (1-p_3)^{\binom{n}{3} - |T|} p_2^{-|K(Q) \cup K(T)|}.
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$$\begin{aligned}
\frac{\Pr[G(n, (p_2, p_3, p_4)) = G]}{\Pr[G(n, p_2) = G]} &\geq \sum_{Q \subseteq \mathcal{H}_4(G)} p_4^{|Q|} (1-p_4)^{\binom{n}{4} - |Q|} p_2^{-|K(Q)|} \\
&\times \sum_{T \subseteq \mathcal{H}_3(G \setminus K(Q))} p_3^{|T|} (1-p_3)^{\binom{n}{3} - |T|} p_2^{-|K(T)|}.
\end{aligned}$$

# Three cases

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- Case I: no artifact triangles
- Case II:  $\exists$  artifact triangles and no artifact quadruples
- Case III:  $\exists$  artifact quadruples

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- Hence,

$$\begin{aligned} &\text{TV}\left(G(n, (p_2, p_3, p_4)), G(n, p_2)\right) \\ &= \sum_{G \in \mathcal{G}} \left( \Pr[G(n, p_2) = G] - \min \left\{ \Pr[G(n, (p_2, p_3, p_4)) = G], \Pr[G(n, p_2) = G] \right\} \right) \\ &\leq O(\varepsilon). \end{aligned}$$

**Remark:** It gives Fill–Scheinerman–Singer–Cohen (2000).

## Case II: $\exists$ artifact triangles and no artifact quadruples

**In this case**, the expected number of artifact triangles is not small, but the expected number of artifact quadruples is small, that is,

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**Remark:** It is not possible for an arbitrary  $G$  to show

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**Idea:** We consider properties of **typical**  $G \in G(n, \hat{p})$ .

For any family  $\mathcal{G}_3$  of **typical** graphs on  $V$ ,

$$\begin{aligned}
 & \text{TV}\left(G(n, (p_2, p_3, p_4)), G(n, p_2)\right) \\
 & \leq \Pr[G(n, p_2) \notin \mathcal{G}_3] \\
 & + \sum_{G \in \mathcal{G}_3} \left( \Pr[G(n, p_2) = G] - \min \left\{ \Pr[G(n, (p_2, p_3, p_4)) = G], \Pr[G(n, p_2) = G] \right\} \right) \\
 & \leq O(\varepsilon) + \sum_{G \in \mathcal{G}_3} \left( \Pr[G(n, p_2) = G] - \min \left\{ \Pr[G(n, (p_2, p_3, p_4)) = G], \Pr[G(n, p_2) = G] \right\} \right).
 \end{aligned}$$

For any family  $\mathcal{G}_3$  of typical graphs on  $V$ ,

$$\begin{aligned} & \text{TV}\left(G(n, (p_2, p_3, p_4)), G(n, p_2)\right) \\ & \leq \Pr[G(n, p_2) \notin \mathcal{G}_3] \\ & + \sum_{G \in \mathcal{G}_3} \left( \Pr[G(n, p_2) = G] - \min \left\{ \Pr[G(n, (p_2, p_3, p_4)) = G], \Pr[G(n, p_2) = G] \right\} \right) \\ & \leq O(\varepsilon) + \sum_{G \in \mathcal{G}_3} \left( \Pr[G(n, p_2) = G] - \min \left\{ \Pr[G(n, (p_2, p_3, p_4)) = G], \Pr[G(n, p_2) = G] \right\} \right). \end{aligned}$$

### Goal

For any  $G \in \mathcal{G}_3$ ,

$$\Pr[G(n, (p_2, p_3, p_4)) = G] \geq (1 - O(\varepsilon)) \Pr[G(n, p_2) = G].$$

- $|\mathcal{H}_3(G)|$  : the number of triangles in  $G$ .
- $I(G)$  : the number of diamond graphs, i.e.,  $K_4$  minus one edge.

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### Lemma

Let  $\mathcal{G}_3$  be the set of all graphs  $G$  on  $V$  satisfying

$$|\mathcal{H}_3(G)| \geq (1 - \delta) \binom{n}{3} p_2^3 \quad \text{and} \quad I(G) \leq n^4 p_2^5 / \varepsilon.$$

Then, for  $\frac{\varepsilon}{nm^{1/3}} < p \leq \left(\frac{3 \log n}{m}\right)^{1/2}$ ,

$$\Pr[G(n, p_2) \in \mathcal{G}_3] = 1 - O(\varepsilon).$$

Taking  $Q = \emptyset$ , we have

$$\begin{aligned} & \frac{\Pr[G(n, (p_2, p_3, p_4)) = G]}{\Pr[G(n, p_2) = G]} \\ & \geq (1 - p_4)^{\binom{n}{4}} \sum_{T \subseteq \mathcal{H}_3(G)} p_3^{|T|} (1 - p_3)^{\binom{n}{3} - |T|} p_2^{-|K(T)|} \\ & \geq (1 - O(\varepsilon)) \sum_{t=0}^{t_0} \sum_{\substack{T \subseteq \mathcal{H}_3(G) \\ |T|=t, |K(T)|=3t}} p_3^t (1 - p_3)^{\binom{n}{3} - t} p_2^{-3t}, \end{aligned}$$

where

$$t_0 := \frac{n^3 m p^3}{\varepsilon} = \Theta\left(\frac{n^3 p_3}{\varepsilon}\right).$$

$$\begin{aligned}
& \frac{\Pr[G(n, (p_2, p_3, p_4)) = G]}{\Pr[G(n, p_2) = G]} \\
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& = (1 - O(\varepsilon)) \sum_{t=0}^{t_0} (1 - O(\varepsilon)) \binom{\binom{n}{3}}{t} p_2^{3t} p_3^t (1 - p_3)^{\binom{n}{3}-t} p_2^{-3t} \\
& \geq (1 - O(\varepsilon)) \sum_{t=0}^{t_0} \binom{\binom{n}{3}}{t} p_3^t (1 - p_3)^{\binom{n}{3}-t} \\
& = (1 - O(\varepsilon)) \left( 1 - \Pr \left[ \text{Bin} \left( \binom{n}{3}, p_3 \right) > t_0 \right] \right) = 1 - O(\varepsilon).
\end{aligned}$$

It implies that

$$\text{TV} \left( G(n, (p_2, p_3, p_4)), G(n, p_2) \right) = O(\varepsilon).$$

## Case III: $\exists$ artifact triangles and quadruples

**In this case**, the expected number of artifact quadruples is not small, that is,

$$\frac{\varepsilon}{n^{2/3}m^{1/3}} < p \leq \left(\frac{3 \log n}{m}\right)^{1/2}.$$



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- $|\mathcal{H}_4(G)|$  : the number of quadruples in  $G$ .

### Lemma

Let  $\mathcal{G}_4 \subset \mathcal{G}_3$  be the set of all graphs  $G$  on  $V$  satisfying

$$|\mathcal{H}_4(G)| \geq \left(1 - \frac{1}{\varepsilon n}\right) \binom{n}{4} p_2^6.$$

Then, for  $\frac{\varepsilon}{n^{2/3}m^{1/3}} < p \leq \left(\frac{3 \log n}{m}\right)^{1/2}$ ,

$$\Pr[G(n, p_2) \in \mathcal{G}_4] = 1 - O(\varepsilon).$$

## Goal

For any  $G \in \mathcal{G}_4$ ,

$$\Pr[G(n, (p_2, p_3, p_4)) = G] \geq (1 - O(\varepsilon)) \Pr[G(n, p_2) = G].$$

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$$\begin{aligned} & \frac{\Pr[G(n, (p_2, p_3, p_4)) = G]}{\Pr[G(n, p_2) = G]} \\ & \geq \sum_{Q \subseteq \mathcal{H}_4(G)} p_4^{|Q|} (1-p_4)^{\binom{n}{4} - |Q|} p_2^{-|K(Q)|} \cdot \sum_{T \subseteq \mathcal{H}_3(G \setminus K(Q))} p_3^{|T|} (1-p_3)^{\binom{n}{3} - |T|} p_2^{-|K(T)|} \\ & \geq (1 - O(\varepsilon)) \cdot \min_{\substack{Q \subseteq \mathcal{H}_4(G) \\ |Q| \leq q_0}} \sum_{T \subseteq \mathcal{H}_3(G \setminus K(Q))} p_3^{|T|} (1-p_3)^{\binom{n}{3} - |T|} p_2^{-|K(T)|}, \end{aligned}$$

where  $q_0 := \frac{n^4 m p^4}{\varepsilon} = \Theta\left(\frac{n^4 p_4}{\varepsilon}\right)$ .

$$\text{Let } t_0 := \frac{n^3 m p^3}{\varepsilon} = \Theta\left(\frac{n^3 p_3}{\varepsilon}\right) \text{ and } r := \frac{n^4 m^2 p^6}{\varepsilon^3} = \Theta\left(\frac{n^4 p_3^2}{\varepsilon^3}\right).$$

We have that

$$\begin{aligned} & \sum_{T \subseteq \mathcal{H}_3(G \setminus K(Q))} p_3^{|T|} (1 - p_3)^{\binom{n}{3} - |T|} p_2^{-|K(T)|} \\ & \geq \sum_{t=0}^{t_0} \sum_{\substack{T \subseteq \mathcal{H}_3(G \setminus Q) \\ |T|=t, |K(T)| \leq r}} p_3^t (1 - p_3)^{\binom{n}{3} - t} p_2^{-6t+r} \\ & \geq \sum_{t=0}^{t_0} (1 - O(\varepsilon)) \binom{\binom{n}{3}}{t} p_2^{6t} \cdot p_3^t (1 - p_3)^{\binom{n}{3} - t} p_2^{-6t+r} \\ & \geq (1 - O(\varepsilon)) p_2^r \cdot \sum_{t=0}^{t_0} p_3^t (1 - p_3)^{\binom{n}{3} - t} \geq (1 - O(\varepsilon)), \end{aligned}$$

$$\text{since } p_2^r = (1 - e^{-mp^2(1-p)^{n-2}})^r \geq 1 - O(re^{-mp^2}) = 1 - O(\varepsilon).$$

# Problem

## Problem

Fix  $3 < \alpha < 6$ , and let  $m = n^\alpha$ .

Find a probability  $p^* = p^*(n, m)$  such that

- If  $p \ll p^*$ , then  $\text{TV}(G(n, m; p), G(n, \hat{p})) = o(1)$ .
- If  $p \gg p^*$ , then  $\text{TV}(G(n, m; p), G(n, \hat{p})) \geq c$ ,  
for some positive constant  $c > 0$ .

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## Question

When  $G(n, m; p) \not\sim G(n, \hat{p})$ ,

what are interesting properties and structures of  $G(n, m; p)$ ?

*Thank you for your attention!*

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# Proof of Lemma 1

## Lemma

- 1  $\text{TV}(G(n, m; p), G(n, (p_k))) = o(1)$ . (essentially by Rybarczyk)
- 2  $\text{TV}(G(n, p_2, p_3, p_4), G(n, p_2)) = o(1)$ . (Main part)
- 3  $\text{TV}(G(n, p_2), G(n, \hat{p})) = o(1)$ . (Not hard)

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## Idea of Proof

- 1 Coupling argument
- 2 Property of Poisson distribution

## Coupling argument

### Definition

For two random variables  $X$  and  $Y$ , the **coupling**  $(X', Y')$  of  $X$  and  $Y$  is a random variable on the product of the sample spaces of  $X$  and  $Y$  such that the marginal distributions of  $X'$  and  $Y'$  are the distributions of  $X$  and  $Y$ , respectively.

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### Lemma

$X, Y$  : random variables.

- 1 Any coupling  $(X', Y')$  of  $X$  and  $Y$  satisfies

$$\text{TV}(X, Y) \leq \Pr[X' \neq Y'].$$

- 2 There exists a coupling such that

$$\text{TV}(X, Y) = \Pr[X' \neq Y'].$$

# Proof of Lemma (1)

- $X$  := the number of columns of the matrix  $R(n, m; p)$  with two or more 1's.
- $\Pr[|V_a| = k] = \binom{n}{k} p^k (1-p)^{n-k} =: r_k$
- $X = \text{Binom}(m, q_2)$  where  $q_2 := \sum_{k \geq 2} r_k$ .

$G(n, m; p)$  can be constructed as follows:

- 1  $K^{(1)}, \dots, K^{(h)}, \dots$ : i.i.d. random complete graphs on subsets of  $V$ 
  - the number of vertices in  $K^{(1)}$  is  $k (\geq 2)$  with probability  $r_k/q_2$
  - Then, once the number is given to be  $k$ , every  $k$ -subset of  $V$  is equally likely to be the vertex set of  $K^{(1)}$ .
  - In other words, for a  $k$ -subset  $U$  of  $V$  with  $k \geq 2$ , the probability of  $U$  being the vertex set of  $K^{(1)}$  is  $\frac{r_k}{q_2} \binom{n}{k}^{-1}$ .
- 2  $G(n, m; p)$  is the edge union of  $X$  random complete graphs  $K^{(1)}, \dots, K^{(X)}$ .

## Definition ( $G_Y$ )

- $Y := \text{Poisson}(mq_2)$  that is coupled with  $X$  so that

$$\Pr[X \neq Y] = \text{TV}(X, Y).$$

- Let  $G_Y$  be the graph whose edge set is the (edge) union of  $K^{(1)}, \dots, K^{(Y)}$ .

## Property

- 1  $G_Y$  has the same distribution as  $G(n, (p_k))$ .
- 2

$$\begin{aligned} \text{TV}(G(n, m; p), G_Y) &\leq \Pr[G(n, m; p) \neq G_Y] \\ &\leq \Pr[X \neq Y] = \text{TV}(X, Y). \end{aligned}$$

## Lemma (Barbour and Holst (1989))

Let  $X := \text{Binom}(m, q_2)$  and  $Y := \text{Poisson}(mq_2)$ . Then

$$\text{TV}(X, Y) \leq q_2.$$

$$\begin{aligned} \text{TV}(X, Y) \leq q_2 &= \sum_{k \geq 2} \binom{n}{k} p^k (1-p)^{n-k} \leq \sum_{k \geq 2} n^k p^k = O(n^2 p^2) \\ &= O\left(\frac{n^2 \log n}{m}\right) = o(1). \end{aligned}$$



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