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Presented paper by Noga Alon

## Bipartite decomposition of random graphs

(<http://arxiv.org/abs/1402.6466>)

### Definitions

**Definition** (Maximal size complete bipartite induced subgraph).

$\beta(G)$  := size of maximal complete bipartite induced subgraph of  $G$ .

**Definition** (Minimal bipartite decomposition number).

$\tau(G)$  := minimal number of complete edge disjoint covering bipartite subgraphs of  $G$ .

**Definition** (Minimal nontrivial bipartite decomposition number).

$\tau'(G)$  := minimal number of complete edge disjoint covering nontrivial (non-star) bipartite subgraphs of  $G$ . ( $\infty$  if didn't exist)

**Definition** .

$\gamma(G)$  :=  $|V|$ –number of  $C_4$  forming  $G$ .

### Main theorems

**Conjecture** (*Erdős disproved conjecture*)

For  $G \in G(n, \frac{1}{2})$  whp

$$\tau(G) = n - \alpha(G).$$

**Theorem (1).**

$$f(k) := \binom{n}{k} 2^{-\binom{k}{2}},$$

$$k_0 := \max k \text{ st. } f(k) \geq 1.$$

For  $G \in G(n, \frac{1}{2})$

(i) If  $f(k_0 + 1) \ll 1 \ll f(k_0)$  then  $\alpha(G) = k_0$  and  $\beta(G) = k_0 + 2$  whp.

(ii) If  $f(k_0) \in \Theta(1)$  then whp one of the following, each of them with probability bounded away from 0 and 1.

(a)  $\alpha(G) = k_0$  and  $\beta(G) = k_0 + 2$

(b)  $\alpha(G) = k_0$  and  $\beta(G) = k_0 + 1$

(c)  $\alpha(G) = k_0 - 1$  and  $\beta(G) = k_0 + 2$

(d)  $\alpha(G) = k_0 - 1$  and  $\beta(G) = k_0 + 1$

(iii) If  $f(k_0 + 1) \in \Theta(1)$  then whp one of the following, each of them with probability bounded away from 0 and 1.

Same as previous, set there  $k_0 := k_0 + 1$

**Theorem (2).**

$\exists c > 0$  st.  $\forall p$  st.  $\frac{2}{n} \leq p \leq c$  then  $G \in G(n, p)$  whp satisfies

$$\tau(G) = n - \Theta\left(\frac{\log(np)}{p}\right).$$

**Theorem** (3).

If  $p \ll n^{-\frac{7}{8}}$  then  $G \in G(n, p)$  whp satisfies

$$\tau(G) = n - \max_{G \supseteq H \text{ induced, st. components are } C_4 \text{ or } v \in V} (\gamma(H)).$$

### Some lemmas

**Theorem** (Stein-Chen method).

Let  $\{X_\alpha\}_{\alpha \in \mathcal{F}}$  be a finite family of indicator variables with dependency graph  $L$  and  $X := \sum_{\alpha \in \mathcal{F}} X_\alpha$  having  $E[X] = \sum_{\alpha \in \mathcal{F}} E[X_\alpha] = \lambda$ . Let  $PO(X)$  be a Poisson random variable with expectation  $\lambda$ , then:

$$\sup_{\alpha \in \mathcal{F}} |P(X_\alpha) - PO(X_\alpha)| \leq \min(\lambda^{-1}, 1) \left( \sum_{\alpha \in \mathcal{F}} E[X_\alpha]^2 + \sum_{\alpha, \beta \in \mathcal{F}, (\alpha, \beta) \in E(L)} (E[X_\alpha X_\beta] + E[X_\alpha]E[X_\beta]) \right).$$

**Lemma** (Trivial vs. non-trivial).

$\forall G(v, E) \exists U \subseteq V$  st.

$$\tau(G) = |V| - |U| + \tau'(G).$$

**Theorem** (Čebyšev inequality).

$$P(|X - E[X]| \geq E[X]) \leq \frac{\text{Var}[X]}{(E[X])^2}$$

**Lemma** (About variance).

$$\begin{aligned} \text{Var}[X] &= E[X^2] - (E[X])^2, \\ \text{Cov}[X, Y] &= E[XY] - E[X]E[Y], \\ \text{Var}\left[\sum_i X_i\right] &= \sum_i \text{Var}[X_i] + \sum_{i \neq j} \text{Cov}[X_i, X_j]. \end{aligned}$$

**Lemma** (Technical lemma).

$\exists a, c, C; \forall p$  st.  $0 \leq p \leq c$  and  $np \geq C \log n$  then

$$P(\tau'(G) \leq 2n) \leq 2^{-apn^2}.$$

**Lemma** (relation of expectation of IS and BG).

Define

$$g(k) := \binom{n}{k+2} (2^{k+1} - 1) 2^{-\binom{k+2}{2}}.$$

Then  $g(k) \in \Theta(f(k))$ .

### Useful estimates

**Lemma**

$$\begin{aligned} 1 &\geq \binom{n}{k_0+1} 2^{-\binom{k_0+1}{2}} \geq \left(\frac{n}{k_0+1}\right)^{k_0+1} 2^{-\binom{k_0+1}{2}}, \\ 1 &\leq \binom{n}{k_0} 2^{-\binom{k_0}{2}} \leq \left(\frac{ne}{k_0}\right)^{k_0} 2^{-\binom{k_0}{2}}, \\ n &= \Theta(k_0 2^{\frac{k_0}{2}}). \end{aligned}$$