

Colóquio Brasileiro de Matemática

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**Counting**  
**CONNECTED**  
**Graphs**  
**using**  
**ERDŐS MAGIC**

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Joint with

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Complexity =  $E - V + 1$

$C(k, l)$  = Number of

CONNECTED Labelled Graphs

$k$  Vertices

Complexity  $l$

$C(k, 0) = k^{k-2}$  Cayley

$C(k, l) \sim c_l k^{3l/2} k^{k-2}$  Wright

$l > (\frac{1}{2} + \epsilon)k \ln k$  Erdős-Rényi

General  $k, l \rightarrow \infty$  Bender, Canfield, McKay (1990)

via 2-core: Pittel, Wormald (2005)

## Areas

- Discrete Mathematics
- Theoretical Computer Science

### Breadth First Search Algorithms

- Probability
- Percolation

## Tilted Balls in Bins

$k - 1$  balls,  $k$  bins,  $p \in (0, 1]$

Truncated Geometric

Ball  $j$  in Bin  $T_j$

$$\Pr[T_j = i] = \frac{p(1-p)^{i-1}}{1 - (1-p)^k}$$

$Z_i$  balls in bin  $i$

$Y_0 = 1$ ,  $Y_i = Y_{i-1} + Z_i - 1$  (so  $Y_k = 0$ )

TREE:  $Y_t > 0$ ,  $0 \leq t < k$  (=Parking Function)

$$M := \sum_{i=0}^k (Y_i - 1) = \binom{k}{2} - \sum_{j=1}^k T_j$$

$G(k, p)$  Random Graph

Vertices  $0, 1, \dots, k - 1$

Adjacency Prob  $p$

THM: Prob  $G(n, p)$  Connected  $= A_1 A_2$

THM: Prob  $G(k, p)$  Connected & Complexity  $l$   
 $= C(k, l) p^{k+l-1} (1-p)^{k(k-1)/2-k-l+1} = A_1 A_2 A_3$

with

$$A_1 = (1 - (1 - p)^k)^{k-1}$$

$$A_2 = \Pr[\text{TREE}]$$

$$A_3 = \Pr[\text{BIN}[M, p] = l | \text{TREE}]$$

Strategy:  $A_2, A_3$  determine  $C(k, l)$

## Breadth First Search

1	2	3	4	5
N	N	Y	Y	N
N	N	-	-	N
Y	N	-	-	N
-	Y	-	-	Y
-	-	-	-	-
-	-	-	-	-

$$T_3 = T_4 = 1, T_1 = 3, T_2 = T_5 = 4$$

$A_1$ : All  $T_j$  defined

$$\vec{Z} = (2, 0, 1, 2, 0, 0)$$

Walk  $\vec{Y} = (1, 2, 1, 1, 2, 1, 0)$

TREE: BFS doesn't terminate early

Tree Edges 03, 04, 41, 12, 15

$M = 2$  Unexposed 34, 25

## Setting the Tilt $p$

$$\mu := E[M], \sigma^2 := Var[M]$$

$$p\mu = l$$

### Three Regimes

Small  $l = o(k)$ ,  $k^{-3/2} \ll p \ll k^{-1}$

Large  $l = \Theta(k)$ ,  $p = \Theta(k^{-1})$

Very Large  $l \gg k$ ,  $p \gg k^{-1}$

( $l > ck \ln k$ ;  $p > c' \frac{\ln k}{k}$  Erdős-Rényi)

( $l \neq 0$ , fixed);  $p = \Theta(k^{-3/2})$ ; complicated)

Small:  $k^{-3/2} \ll p \ll k^{-1}$

$$\epsilon = \frac{1}{2}pk$$

Left  $Z_i$  Poisson  $1 + \epsilon$

Galton-Watson  $\Pr[\text{ESC}] \sim 2\epsilon$

Right  $Z_i^* = Z_{k-i}; Y_i^* = Y_{k-i}$

$$Y_0^* = 0, Y_i^* = Y_{i-1}^* + 1 - Z_i^*$$

$Z_i^*$  Poisson  $1 - \epsilon$

$\Pr[\text{ESC}^*] \sim \epsilon$

\*\*\* Scaling for ESC, ESC\* is  $\epsilon^{-2} \ll k$

$\Pr[\text{TREE}] \sim \Pr[\text{ESC}] \Pr[\text{ESC}^*] \sim 2\epsilon^2$

Note:  $p \ll k^{-3/2} \Rightarrow \Pr[\text{TREE}] \sim k^{-1}$

$p = \Theta(k^{-3/2}) \Rightarrow \Pr[\text{TREE}]$  complicated



$$\text{Large: } p \sim \frac{c}{k}$$

$$\text{Left } Z_i \text{ Poisson } \frac{c}{1-e^{-c}}$$

$$\text{Galton-Watson } \Pr[\text{ESC}] \sim 1 - e^{-c}$$

$$\text{Right } Z_i^* = Z_{k-i}; Y_i^* = Y_{k-i}$$

$$Y_0^* = 0, Y_i^* = Y_{i-1}^* + 1 - Z_i^*$$

$$Z_i^* \text{ Poisson } \frac{ce^{-c}}{1-e^{-c}}$$

$$\Pr[\text{ESC}^*] \sim 1 - \frac{ce^{-c}}{1-e^{-c}}$$

Chernoff:  $Y_i > 0$  in middle

$$\Pr[\text{TREE}] \sim \Pr[\text{ESC}] \Pr[\text{ESC}^*] \rightarrow 1 - (c+1)e^{-c}$$

$$\text{Very Large } p \gg k^{-1}$$

Chernoff:  $Y_i > 0$  all  $i$

$$\Pr[\text{TREE}] \rightarrow 1$$

## Gaussian $M$

$$M := \binom{k}{2} - \sum_{j=1}^k T_j$$

Esseen:  $\Pr[M < \mu + u\sigma] \rightarrow \Pr[N(0, 1) < u]$

\*\*\* Still holds *conditional on TREE*

Hardest when  $p$  barely  $\gg k^{-3/2}$

Easy when  $p$  Large

Trivial when  $p$  Very Large

## From CLT to Local Stats

$$E[W] = \mu, \text{Var}[W] = \sigma^2$$

$$V = \text{BIN}[W, p], l = \mu p = E[V]$$

Formally: Infinite sequence of  $W_k, p_k, \sigma_k, l_k$

Assume **\*\***  $p^2 \sigma^2 = O(p\mu)$  **\*\***

$$(\sigma^+)^2 := p^2 \sigma^2 + p\mu$$

Assume  $\sigma^{-1}(W - \mu) \rightarrow N(0, 1)$

THM:  $V$  Local CLT, Mean  $l$ , Var  $(\sigma^+)^2$

$$\Pr[V = l] \rightarrow \frac{1}{\sqrt{2\pi\sigma^+}}$$

Apply to  $M|\text{TREE}$

$$A_3 = \Pr[BIN[M, p] = l | TREE]$$

Small:  $p\mu = l = p^2\sigma^2$ ,  $A_3 \sim (4\pi l)^{-1/2}$

Very Large:  $p\mu = l \gg p^2\sigma^2$ ,  $A_3 \sim (2\pi l)^{-1/2}$

Large:  $p \sim \frac{c}{k}$ .  $p^2\sigma^2 = \Theta(l)$ ,  $A_3 \sim g(c)l^{-1/2}$

## The Giant/Dominant Component

$G(n, p)$

$p = \frac{c}{n}$ ,  $c > 1$  Erdős-Rényi Giant

$p = \frac{1}{n} + \lambda n^{-4/3}$ ,  $\lambda \rightarrow +\infty$

Supercritical: Dominant Component

THM: Probability  $C(v)$  has  $k$  vertices, complexity  $l$  is  $\sim A_1^* A_2 A_3$  with

$$A_1^* = \Pr[\text{BIN}[n-1, 1 - (1-p)^k] = k-1]$$

Corollary: Local Stats for  $k, l$  of Giant/Dominant Component. Correlated Gaussians

# Generating Random Connected Graph with Nitin Arora

Time  $\Theta(K + L)$  (!! ) for  $L = \Omega(\ln K)$

$p$  with  $p\mu = L$

Tilted Balls into Boxes

$L = \Omega(K)$  get BFS Tree with prob.  $\Omega(1)$

$L = o(K)$  use Fast Abort.

Add precisely  $L$  of  $M$  unexposed with prob.

$$\frac{\Pr[\text{BIN}[M, p] = L]}{\max_m \Pr[\text{BIN}[m, p] = L]}$$

Any new possibility that existence acquires,  
even the least likely,  
transforms everything about existence.

– Milan Kundera