

# On Randomly Weighted, Randomly Perturbed Dense Graphs

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# On Randomly Weighted, Randomly Perturbed Dense Graphs

We let  $G_0$  be a dense graph or digraph and we add random edges  $R$  to create a graph or digraph  $G = G_0 + R$ .

Introduced by Bohman, F, Martin in 2003.

Many recent papers.

Anastos, F in 2019 randomly colored the edges and considered (i) rainbow Hamilton cycles and (ii) rainbow connectivity.

Aigner-Horev and Hefetz 2020 gave improvements to (i) and Balogh, Finlay and Palmer 2021 gave improvements to (ii).

# On Randomly Weighted, Randomly Perturbed Dense Graphs

In this talk we also give the edges of  $G$  independent random weights.

Randomly adding randomly weighted, randomly colored edges, not yet considered.

# On Randomly Weighted, Randomly Perturbed Dense Graphs

In this talk we also give the edges of  $G$  independent random weights.

We consider the case where  $G_0$  is  $\alpha n$ -regular.

- 1 Minimum Spanning Trees.
- 2 Shortest Paths.
- 3 Perfect matchings in bipartite graphs.
- 4 Asymmetric TSP (with Peleg Michaeli).

# Minimum Spanning Trees

Let  $G$  be an asymptotically  $r$ -regular and suppose each edge has an independent uniform  $[0, 1]$  random weight.

Let  $L_n$  be the (random) minimum weight of a spanning tree.

Theorem (Beveridge, F, McDiarmid)

Suppose that  $r \rightarrow \infty$ . Given a moderate connectivity condition **MCC**,  $L_n \sim \frac{n}{r} \zeta(3)$ .

- 1  $G = K_n : L_n \sim \zeta(3)$ .
- 2  $G = K_{n,n} : L_n \sim 2\zeta(3)$ .
- 3  $G = Q_n : L_n \sim \frac{2^n}{n} \zeta(3)$ .

Adding  $o(n^2)$  randomly weighted random edges to an asymptotically  $\alpha n$ -regular graph satisfies **MCC** w.h.p.  
So, w.h.p.  $L_n \sim \alpha^{-1} \zeta(3)$ .

# Shortest Paths

The edges of  $G_0 + R$  are given independent  $EXP(1)$  random lengths –  $\Pr(EXP(1) \geq \lambda) = e^{-\lambda}$ .

Let  $d_{i,j}$  denote the length of a shortest path from  $i$  to  $j$ .

## Theorem

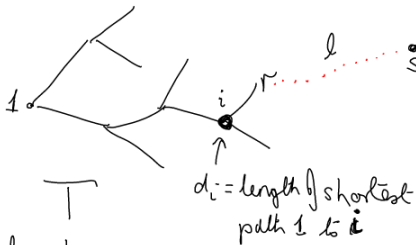
- ①  $d_{1,2} \sim \frac{\log n}{\alpha n}$  w.h.p.
- ②  $\max_j d_{1,j} \sim \frac{2 \log n}{\alpha n}$  w.h.p.
- ③  $\max_{i,j} d_{i,j} \sim \frac{3 \log n}{\alpha n}$  w.h.p.

Janson 1999 proved this when  $\alpha = 1, R = \emptyset$ .

We adapt his argument.

# Shortest Paths

## Dijkstra Algorithm



$k$  vertices  
 $0 = d_1 \leq d_2 \leq \dots \leq d_k$

$\mathcal{V}_k$  edges from  
 $\mathcal{T} \text{ to } \overline{\mathcal{T}}$

$d_s = \text{length of shortest path}$

$$d_r + l \geq d_k$$

$$\Rightarrow l = d_k - d_r + E(i)$$

$$\Rightarrow d_{k+1} - d_k = \min_{\mathcal{V}_k \text{ indep. } E(i)'s} E(i)$$

$$E(d_{k+1} - d_k) = \frac{1}{\mathcal{V}_k}$$

# Shortest Paths

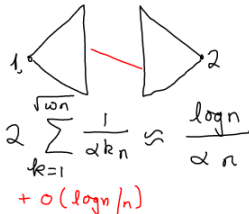
$$\text{So } E(d_m) = \frac{1}{v_1} + \frac{1}{v_2} + \dots + \frac{1}{v_m}$$

Complete Graph:  $v_k = k(n-k)$

H + R:  $v_k \approx \alpha k n$  for  $k = o(n)$

$d_{1,2}$ : Upper Bound

Lower Bound  
Similar



$$2 \sum_{k=1}^{\sqrt{wn}} \frac{1}{\alpha k n} \approx \frac{\log n}{\alpha n} + o(\log n / n)$$

$m = \sqrt{wn}$  vertices  
in each tree  
Cheap random edge



# Perfect Matchings

Perfect matchings in  $K_{n,n}$ :  $M_n$  is the minimum weight matching.

- 1 Walkup 1979:  $\mathbf{E}(w(M_n)) \leq 3$ ; Uniform  $[0, 1]$  edge weights.

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- ⑤ Parisi 1998: Conjecture  $\mathbf{E}(w(M_n)) = \sum_{i=1}^n \frac{1}{i^2}$ .

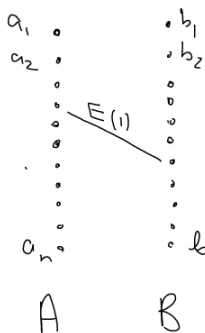
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- ⑥ Linusson, Wästlund 2004 and Nair, Prabhakar, Sharma 2005 verified the Parisi conjecture.

# Perfect Matchings

Assume edge weights are exponential mean 1.



$M_r = \text{min. wt matching of}$   
 $A_r = \{a_1, a_2, \dots, a_r\}$  into  $B$

Parisi conjecture equivalent to

$$E[\omega(M) - \omega(M_{r-1})] = \frac{1}{r} \left[ \frac{1}{n} + \frac{1}{n-1} + \dots + \frac{1}{n-r+1} \right]$$

Lovely "proof" next.

# Perfect Matchings – Wästlund

$X = w(M_r)$  and  $Y_i$  is the minimum weight of a matching from  $A_r \setminus \{a_i\}$  into  $B$ .

Add a new vertex  $b^*$  and edges of length  $EXP(\lambda)$  from  $A_r$  to  $B^* = B \cup \{b^*\}$ .

$$\begin{aligned} P(r) &= \Pr(b^* \in M_r^*) = \mathbf{E} \left( \sum_{i=1}^r \Pr(w(a_i, b^*) < X - Y_i) \right) \\ &= \mathbf{E} \left( \sum_{i=1}^r (1 - e^{-\lambda(X - Y_i)}) \right) \\ &= \lambda \sum_{i=1}^r \mathbf{E}(X - Y_i) + O(\lambda^2) \\ &= \lambda r (\mathbf{E}(w(M_r)) - \mathbf{E}(w(M_{r-1}))) + O(\lambda^2). \end{aligned}$$

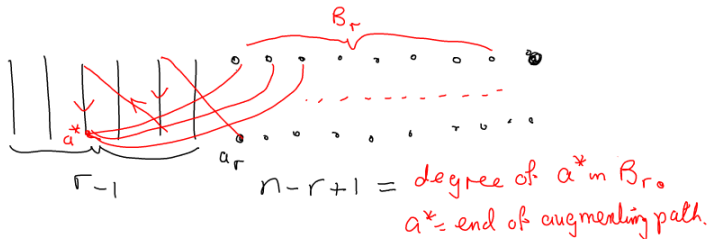


# Perfect Matchings – Wästlund

$$\underline{H + R}$$

$$P(b^* \in M_r^* \mid b^* \notin M_{r-1}^*) = \frac{\lambda}{n-r+1+\lambda}$$

$K_{n,n}$



$$\begin{aligned} P(r) &= \lambda r (\mathbf{E}(w(M_r)) - \mathbf{E}(w(M_{r-1}))) + O(\lambda^2) \\ &= \Pr(b^* \in M_r^*) = 1 - \prod_{i=0}^{r-1} \Pr(b^* \in M_{r-i}^* \mid b^* \notin M_{r-i-1}^*) \end{aligned}$$

In  $K_{n,n}$  we have

$$P(r) = 1 - \prod_{i=0}^{r-1} \frac{n-i}{n-i+\lambda} = \lambda \sum_{i=0}^{r-1} \frac{1}{n-i} + O(\lambda^2).$$

In  $G_0 + R$  we have

$$P(r) = \lambda \sum_{i=0}^{r-1} \frac{1}{\delta_i} + O(\lambda^2)$$

where  $\delta_r$  is the degree of  $a^*$  in  $B_r$ .

# Perfect Matchings

We should have  $\delta_r \sim \alpha(n - r + 1)$  implying that

$$\mathbf{E}(w(M_n)) \sim \frac{\pi^2}{6\alpha}.$$

We can only prove this when  $G_0$  is pseudo-random in the sense of Thomason 1989.

The proof is a bit technical.

# Asymmetric TSP

Let  $C(i, j)$  be independent  $EXP(1)$  for  $1 \leq i \neq j \leq n$ .

We let the  $C(i, j)$  be the costs for an instance of the Asymmetric Travelling Salesperson Problem (ATSP). I.e. the minimum cost of a Hamilton cycle in the complete digraph  $\vec{K}_n$ .

The Assignment Problem (AP) is the problem of finding the minimum cost perfect matching in  $K_{n,n}$  with costs  $C(i, j)$ .

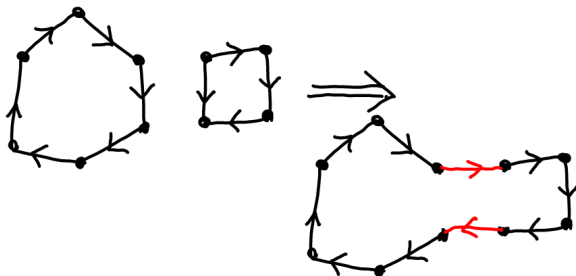
AP is equivalent to finding the minimum cost of a collection of vertex disjoint cycles in  $K_{n,n}$  that cover all vertices.

It follows that in terms of optimal values  $val(AP) \leq val(ATSP)$ .

# Asymmetric TSP

Karp 1979 descibed a *patching* algorithm that in  $O(n^3)$  time w.h.p. produces a Hamilton cycle  $H_{karp}$  of weight  $(1 + o(1))val(AP) \leq (1 + o(1))val(ATSP)$ .

The algorithm solves  $AP$  and then patches the associated cycles together as cheaply as possible.



# Asymmetric TSP

We replace  $\vec{K}_n$  with  $D = D_0 + R$  where  $D_0$  has minimum in- and out-degree at least  $\alpha n$ .

## Theorem

*Suppose that  $|R| = n^{2-\epsilon}$  and that each edge of  $D$  is given an independent  $\text{EXP}(1)$  cost. Then w.h.p.  $\text{val}(\text{ATSP}) = (1 + o(1))\text{val}(\text{AP})$  and Karp's patching algorithm finds a tour of the claimed cost in polynomial time.*

The proof rests on the following lemma:

## Lemma

- (a) *W.h.p., the solution to AP contains only edges of cost  $C(i, j) \leq \gamma_n = n^{-(1-2\epsilon)}$ .*
- (b) *W.h.p., after solving AP, the number  $\nu_C$  of cycles is at most  $r_0 \log n$  where  $r_0 = n^{1-3\epsilon}$ .*

# Asymmetric TSP

Given the lemma, the proof is simple. Let  $\mathcal{C} = C_1, C_2, \dots, C_\ell$  be a cycle cover and let  $k_i = |C_i|$  where  $k_1 \leq k_2 \leq \dots \leq k_\ell$ ,  $2 \leq \ell \leq r_0$ . Different edges in  $C_i$  give rise to disjoint patching pairs. We only consider the random edges  $R$  when looking for a patch. The number of possible patching pairs  $\nu_{\mathcal{C}}$  satisfies

$$\begin{aligned}\nu_{\mathcal{C}} &\geq \sum_{i \neq j} k_i k_j = \frac{1}{2} \left( n^2 - \sum_{i=1}^{\ell} k_i^2 \right) \\ &\geq \frac{1}{2} \left( n^2 - ((n - \ell + 1)^2 + \ell - 1) \right) \geq \frac{\ell n}{2}.\end{aligned}$$

Each of these  $\nu_{\mathcal{C}}$  pairs uses a disjoint set of edges.

# Asymmetric TSP

We define the sets

$$R_\ell = \left\{ e \in R : C(e) \leq \gamma_n + \left( \frac{\log n}{\ell n^{1-5\epsilon/2}} \right)^{1/2} \right\}, \quad 1 \leq \ell \leq r_0.$$

Let  $\mathcal{E}_\ell$  be the event that  $|\mathcal{C}| = \ell$  and that there is no patch using only edges in  $R_\ell$ . If  $\mathcal{E}_\ell$  does not occur then we reduce the number of cycles by at least one. We have

$$\Pr(\mathcal{E}_\ell) \leq \exp \left\{ -\frac{n^{4-2\epsilon} \log n}{18n^{4-5\epsilon/2}} \right\} = o(n^{-1}).$$

It follows that  $\Pr(\exists i : \mathcal{E}_i) = o(1)$  and then the union bound implies that w.h.p. the patches involved in these cases add at most the following to the cost of the assignment:

$$\sum_{\ell=1}^{r_0 \log n} \left( \gamma_n + \left( \frac{\log n}{\ell n^{1-5\epsilon/2}} \right)^{1/2} \right) \leq r_0 \gamma_n \log n + \left( \frac{2r_0 \log^2 n}{n^{1-5\epsilon/2}} \right)^{1/2} = o(1).$$



# THANK YOU