

The Parisi and Coppersmith-Sorkin 'Conjectures'

- **Setting** — $m \times n$ matrix A of iid non-negative r.v.s
 - Want to pick k entries which form a **matching** (i.e., at most one entry per row/column)
 - $C_{k,m,n} \equiv$ weight of minimum k -matching
($C_n \equiv$ wt of min matching when $m=n$)

- **Results**
 - If $A_{ij} \sim U[0,1]$, then $\mathbb{E}[C_n]$ bounded as $n \rightarrow \infty$ (Walkup '79)
 - If $A_{ij} \sim \text{Exp}(1)$, then $\mathbb{E}[C_n] \rightarrow \pi^2/6$ as $n \rightarrow \infty$
(conjecture by Parisi & Mezard '87, proved by Aldous '00)

- **(The Parisi Conjecture)** If $A_{ij} \sim \text{Exp}(1)$, then

$$\mathbb{E}[C_n] = \sum_{i=1}^n 1/i^2$$

- **(The Coppersmith-Sorkin Conjecture)** If $A_{ij} \sim \text{Exp}(1)$, then

$$\mathbb{E}[C_{k,m,n}] = \sum_{\substack{i,j \geq 0 \\ i+j \leq k}} \frac{1}{(m-i)(n-j)}$$

(conjecture by Parisi '98, CS '99; proved by Linusson-Wästlund and Nair-Priabakaran-Sharma '03)

- Moreover, $C_n \rightarrow \pi^2/6$ in probability (ie, $\text{Var}(C_n) \searrow 0$)

- We will see an argument by J. Wästlund from "An Easy Proof of the $\zeta(2)$ Limit in the Random Assignment Problem" (2009)

Background and some heuristic calculations

- If $X \sim \text{Exp}(\lambda)$, then we have the following
 - i) $f(x) = \lambda e^{-\lambda x}$, $F(x) = 1 - e^{-\lambda x} \quad \forall x \geq 0$
 - ii) $E[X] = 1/\lambda$, $\text{Var}(X) = 1/\lambda^2$
 - iii) $P[X \leq s+t | X \geq s] = P[X \leq t]$ (memoryless)
- Given $X_1, X_2, \dots, X_n$ independent, with $X_i \sim \text{Exp}(\lambda_i)$
 - $\Rightarrow \min_{i \in [n]} \{X_i\} \sim \text{Exp}(\sum_{i=1}^n \lambda_i)$, $\arg\min_{i \in [n]} \{X_i\} \sim i \text{ w.p. } \frac{\lambda_i}{\sum_{i=1}^n \lambda_i}$

Returning to $A \sim m \times n$ matrix of iid $\text{Exp}(1)$ wts

- $\min_{(i,j) \in [m] \times [n]} \{A_{ij}\} \sim \text{Exp}(mn)$ (and hence $E[C_{1,m,n}] = 1/mn$)

- For a **lower bound**, consider picking the k lowest entries

$$\Rightarrow E[C_{k,m,n}] \geq \frac{1}{mn} + \left(\frac{1}{mn} + \frac{1}{mn-1} \right) + \dots + \left(\frac{1}{mn} + \frac{1}{mn-1} + \dots + \frac{1}{mn-k+1} \right)$$

by memorylessness, $E[\text{second smallest } A_{ij} | \text{smallest} = s] = s + 1/mn-1$

For $m=n=k$, we get $E[C_{n,n,n}] \geq \sum_{i=0}^{n-1} \frac{n-i}{n^2-i} \in \left[\frac{1}{n^2}, \frac{1}{n^2-n} \right] \cdot \left(\sum_{i=1}^n i \right)$

$\in \left[\frac{1}{2} + \frac{1}{2n}, \frac{1}{2} + \frac{1}{n-1} \right] \xrightarrow{n \uparrow \infty} \frac{1}{2}$

- For an **upper bound**, consider a greedy policy

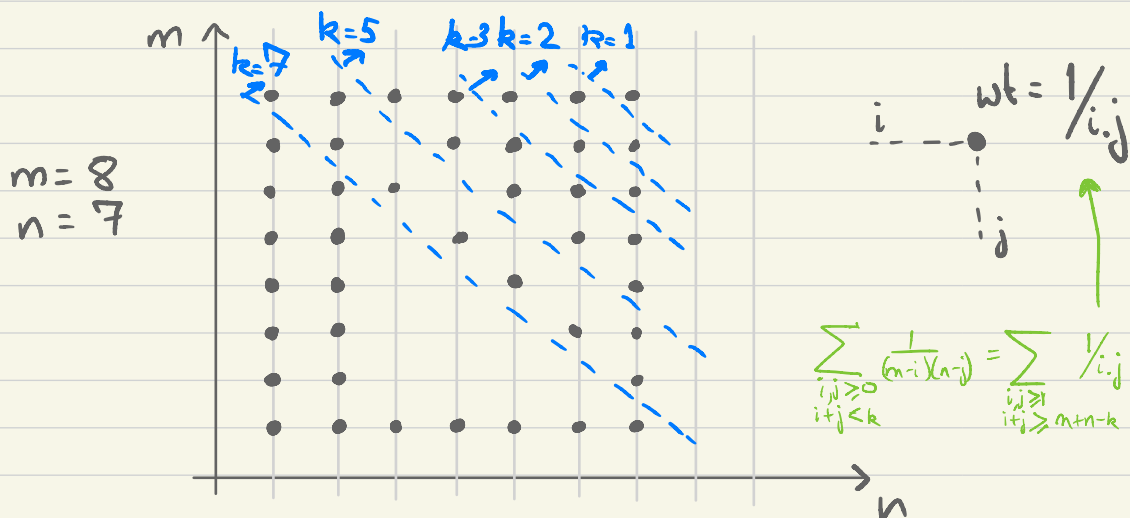
$$\Rightarrow E[C_{k,m,n}] \leq \frac{1}{mn} + \left(\frac{1}{mn} + \frac{1}{(n-1)(n-1)} \right) + \dots + \left(\frac{1}{mn} + \frac{1}{(n-1)(n-1)} + \dots + \frac{1}{(n-k+1)(n-k+1)} \right)$$

We delete the minimum edge, remove the row and column, and find min in remaining $(n-1)(n-1)$ graph

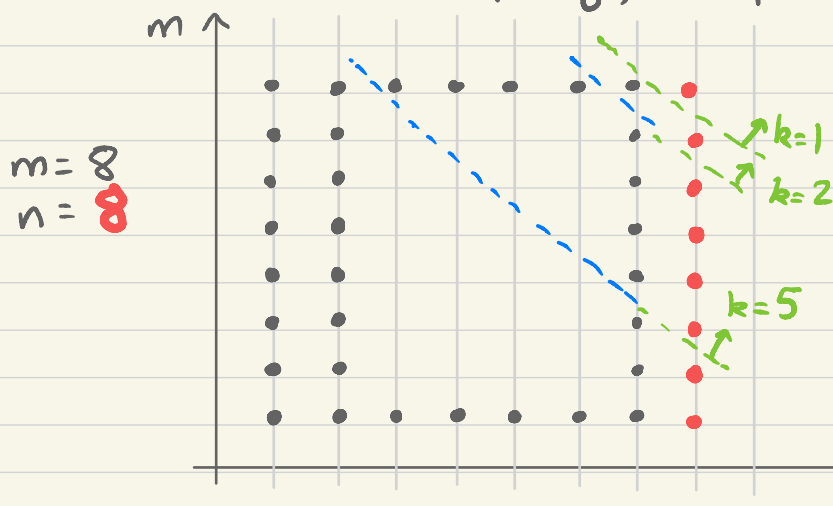
Thus $E[C_{n,n,n}] \leq \sum_{i=0}^{n-1} \frac{n-i}{(n-i)^2} = H_n \in [\ln(n+1), \ln(n)+1]$ ($\uparrow$ as $n \uparrow \infty$)

- Thus, the greedy policy here is very poor! This is in contrast to max wt matching, where greedy is a 2-approximation.
- In fact, getting any constant upper bound here is somewhat non-trivial.

A graphical view of the C-S formula



This leads to a (surprising!) decomposition for induction



$$\mathbb{E}[C_{k,m,n}] = \mathbb{E}[C_{k-1,m,n-1}] + \frac{1}{n} \sum_{i=1}^m \frac{1}{i}$$

The Participation Lemma (Buck-Chan-Robbins)

Given $A \equiv m \times n$ with $A_{ij} \sim \text{Exp}(1)$ iid
 let $\hat{A} \triangleq \begin{bmatrix} A \\ \bar{B} \end{bmatrix}$, where $B \equiv 1 \times n$, $B_j \sim \text{Exp}(\lambda)$ iid
 some parameter

- Since entries are from a continuous r.v. $\Rightarrow$ unique min matchings (ie, matrix $A, \hat{A}$ are 'generic')

Denote $M_k^* \equiv$ unique min k -matching of $\hat{A}$

Lemma - Let $\hat{A}$ be an $m \times n$ matrix, $k < \min(m, n)$ and let M_k^* denote the min-wt k -matching. Then every row which participates in M_k^* also participates in M_{k+1}^*

Pf - Augmenting paths!

In particular, argue that the edges in the symmetric difference of M_k^*, M_{k+1}^* form a single alternating path (no cycles)

Lemma - For any $k \leq \min(m, n)$

$$\mathbb{P}[B \text{ participates in } M_{k+1}^* \mid B \text{ does not participate in } M_k^*] = \frac{\lambda}{\lambda + m - k}$$

Pf - Suppose wlog the first k rows participate in M_k^*

- By previous lemma, one more row in $\{A_{k+1}, A_{k+2}, \dots, A_m, B\}$ ends up participating in M_{k+1}^* . Assume the new element is in column j .

$$\begin{aligned} \mathbb{P}[B \in M_{k+1}^* \mid B \notin M_k^*] &= \mathbb{P}[B_j < A_{i,j} \quad \forall i \in \{k+1, \dots, m\}] \\ &= \frac{\lambda}{\lambda + m - k} \end{aligned}$$

$\xleftarrow{\text{Exp}(\lambda)} \min_i (A_{ij}) \sim \text{Exp}(m-k)$

Proof of the C-S formula

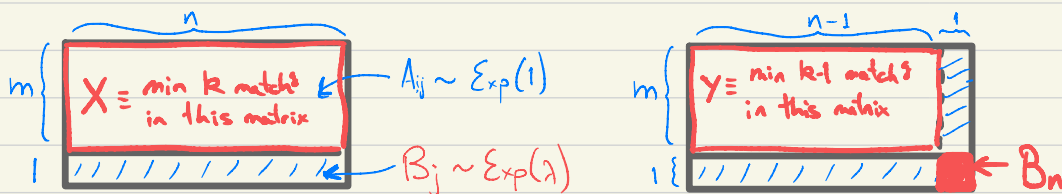
- We want to show by induction that

denote this as $\Delta_{k,m,n}$

$$\mathbb{E}[C_{k,m,n}] - \mathbb{E}[C_{k-1,m,n-1}] = \sum_{i=0}^{k-1} \frac{1}{n-i} \quad (*)$$

- Recall $\hat{A} = A$ augmented with an extra row B with $A_{ij} \sim \text{Exp}(1)$, $B_j \sim \text{Exp}(\lambda)$ iid

Let $X = \text{wt of min } k\text{-matching in } A = \{A_{ij}\}_{i \in m, j \in n}$
 $Y = \text{wt of min } (k-1)\text{-matching in } \{A_{ij}\}_{i \in m, j \in n-1}$



Claim 1 - $X \sim C_{k,m,n}$, $Y \sim C_{k-1,m,n-1}$ ($\sim$ = 'distributed as')

- This is by definition of $C_{k,m,n}$ and X, Y
- As a result, $\mathbb{E}[\Delta_{k,m,n}] = \mathbb{E}[X - Y]$

Also, iterating over the participation lemma, we get

Claim 2 - $\mathbb{P}[B \in M_k^*] = 1 - \prod_{i=0}^{k-1} \left(\frac{n-i}{n-i+\lambda} \right) = 1 - \left(\sum_{i=0}^{k-1} \frac{1}{n-i} \right) + o(\lambda^2)$

Next, by symmetry, we have

$$\underset{\text{edge}}{IP[B_n \in M_k^*]} = \frac{1}{n} IP[\underset{\text{row/node}}{B_i \in M_k^*}] = \frac{\lambda}{n} \sum_{i=0}^{k-1} \frac{1}{m-i} + o(\lambda^2)$$

Finally, we want to express $IP[B_j \in M_k^*]$ in terms of X, Y

Claim 3 - $IP[B_n \in M_k^*] = IP[B_n < X - Y] - o(\lambda^2)$

Warning - the next bit is somewhat sketchy (though true ...)

To see why we need the additional $O(\lambda^2)$ term, note that if $B_n < X - Y$, then the only way $B_n \notin M_k^*$ is if some other B_j is part of an even smaller k matching. For this to happen, we must have $\min_{j \neq n} B_j < \underline{X}$ min matches without B

However, given n iid $\text{Exp}(\lambda)$ n 's, the probability that two of them are less than X is at most (by union bound)

$$E\left[\binom{n}{2} (1 - e^{-\lambda X})^2\right] \leq E[n^2 \lambda^2 X^2] \leq n^2 \lambda^2 E[X^2] = O(\lambda^2)$$

(note - here n is fixed, so $E[X^2] \leq n^2$ which is the sum of all variances.)

Putting things together

$$IP[B_n < X - Y] = \frac{\lambda}{n} \sum_{i=0}^{k-1} \frac{1}{m-i} + o(\lambda^2)$$

$$\text{Also } IP[B_n < X - Y] = P[B_n < \Delta_{k,m,n}] = 1 - E[e^{-\lambda \Delta_{k,m,n}}]$$

$$\text{Thus } E[\Delta_{k,m,n}] = \lim_{\lambda \rightarrow 0} \left[\frac{d}{d\lambda} IP[B_n < X - Y] \right] = \frac{1}{n} \sum_{i=0}^{k-1} \frac{1}{(m-i)}$$