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NASIT, UCLA, August 2022

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The two most basic algorithms?

Given  $n$  elements

Use pairwise comparisons to:

Find maximum  $\rightarrow$  Maxing

Sort  $\rightarrow$  Ranking

# comparisons needed

$N_{\max}$

$N_{\text{sort}}$

Comparison models

Deterministic

Randomized

Close a few gaps

Pairwise comparisons always correct

$$N_{\max} = n - 1$$

$\leq$  sequential elimination: pick candidate, compare, eliminate loser, ...

knockout: essentially same, on tree

$\geq$  graph, edges are comparisons performed

$n - 2$  comparisons  $\longrightarrow$  graph disconnected  $\longrightarrow$  ambiguity

$$N_{\text{sort}} \sim n \log_2 n$$

$\leq$  merge sort, quick sort, heap sort,  $n \log_2 n$

$\geq \log_2 n! \sim \log_2 \left( \sqrt{2\pi n} \cdot \left(\frac{n}{e}\right)^n \right) \sim n \log_2 n$ , correct to constant factor

Comparisons answered randomly – error, noise

Suppose:  $i$  is larger than  $j$

Ask: which larger

Deterministic

- $i$  with probability 1
- $j$  with probability 0

Random

- $i$  with probability  $p_{i,j}$
- $j$  with probability  $p_{j,i} = 1 - p_{i,j}$
- $p_{i,j}$  unknown, except  $\geq \frac{1}{2}$

Repeated comparisons independent

$$p_{i,j} = 0.6 \longrightarrow i, i, j, i, j$$

# Three Comparison Models

Increasing generality

| Model          | $p_{i,j}$  |
|----------------|--|
| Fixed noise    | $\frac{1}{2} \pm \epsilon$ , all same $\epsilon$   |
| Parametric     | Determined by $n$ parameters, one for each element |
| Non-parametric | Essentially arbitrary                              |

$0 \leq \delta \leq 1$  – allowed error

Want answer correct with probability  $1 - \delta$

Max: Largest element w.p.  $1 - \delta$

Sort: Correct order (all pairs) w.p.  $1 - \delta$

|               | $N_{\max}$   | $N_{\text{sort}}$   |
|---------------|--------------|---------------------|
| Deterministic | $n - 1$      | $\sim n \log_2 n$   |
| Randomized    | $\Theta(n)?$ | $\Theta(n \log n)?$ |

Faulty circuits

$$p_{i,j} = \frac{1}{2} \pm \epsilon, \text{ same bias } \epsilon$$

Want: correct max, ranking with probability  $\geq 1 - \delta$

$$N_{\max} = \Theta\left(\frac{n}{\epsilon^2} \log \frac{1}{\delta}\right)$$

$$\text{For } \delta > \frac{1}{n}, \quad N_{\text{sort}} = \Theta\left(\frac{n \log n}{\epsilon^2}\right)$$



## Personal choice

Comparing easier than rating — drinks, vision, drivers, ICML papers

## Advertising

Best ad to show on webpage — room for few (two?) ads, select best

## Design

A/B testing

## Information retrieval

Order of web-search links

## Even classical... Sports

Best or sorted teams

NBA conference, league playoffs

## Modern Times

Facts — alternative

Truth — relative

All opinion — where you come from

## 1990's — Circuits

$p_{i,j}$  — fixed value (0.9 or 0.1), known in advance, typically  $\approx 1$  or 0

## 2010's — Preferences

$p_{i,j}$  — arbitrary, not known in advance, can be  $\approx \frac{1}{2}$

Sports —  $P_{\text{Spain,Portugal}} \approx \frac{1}{2}$  ,  $P_{\text{England,Panama}} \approx 1$

WSJ advertising —  $P_{\text{BMW,Audi}} \approx \frac{1}{2}$  ,  $P_{\text{BMW,Rickshaw}} \approx 1$

Need: Generalize fixed-noise model

## Recall

Elements  $1, \dots, n$

$p_{i,j}$  - probability that  $i$  preferable to  $j$

$\binom{n}{2}$  probabilities

## Parametric models

Probabilities determined by just  $n$  parameters

One associated with each of the  $n$  elements

Parameters  $x_1, \dots, x_n \in \mathbb{R}$

$f : \mathbb{R} \times \mathbb{R} \rightarrow [0, 1]$

$p_{i,j} = f(x_i, x_j)$

$n$  parameters viewed as “weights”

$$w_i \geq 0$$

$$p_{i,j} = f(w_i, w_j) \stackrel{\text{def}}{=} \frac{w_i}{w_i + w_j}$$

$$w_1 = 1, w_2 = 3, w_3 = 5 \longrightarrow p_{3,2} = \frac{5}{5+3} = \frac{5}{8}$$

Chess, Xbox ratings

Fit BTL model, normalize

Also called Plackett Luce [59]

Pairwise comparisons  $\rightarrow$  ordering of  $\geq 2$  elements

Parameters are real values

Corrupted by additive Gaussian noise with 0-mean, known-variance

$P(12 \text{ y/o taller than } 10 \text{ y/o})$

Height at age  $i$  — Mean  $\mu_i$  plus Gaussian noise  $X \sim N(0, 1)$

$$\begin{aligned} p_{i,j} &= f(\mu_i, \mu_j) \\ &\stackrel{\text{def}}{=} \Pr(\mu_i + X' > \mu_j + X'') \\ &= \Pr(X' - X'' > \mu_j - \mu_i) \\ &= \text{erfc}\left(\frac{\mu_j - \mu_i}{\sqrt{2}}\right) \end{aligned}$$

Can have  $N(0, \sigma_i)$  for age  $i$

$p_{i,j} \approx \frac{1}{2} \rightarrow$  many comparisons

BTL

$n = 2$

$(w_1, w_2) \in \{(1, 1 + 10^{-10}), (1 + 10^{-10}, 1)\}$

$N_{\max} = ?$

$\approx \infty$

# PAC Maxing & Ranking [Val 84, Yue et al. 11, Szö et al. 15]

Probably Approximately Correct

$a$  is  $\epsilon$ -preferable to  $b$ :  $p_{a,b} \geq \frac{1}{2} - \epsilon$

Preferable, or “almost preferable”

$\epsilon = 0.02$ ,  $p_{a,b} = .6$  ✓,  $.49$  ✓

Tiny difference  $\rightarrow$  “almost”

$\epsilon$ -maximum: element  $m$  that is  $\epsilon$ -preferable to all others

$$p_{m,x} \geq \frac{1}{2} - \epsilon \quad \forall x$$

$\epsilon$ -ranking: permutation  $r_1, r_2, \dots, r_n$  s.t. whenever  $r_i > r_j$ ,

$$p_{i,j} \geq \frac{1}{2} - \epsilon$$

As before: Answer (for ranking – all) correct with probability  $\geq 1 - \delta$

For Bradley, Terry, Luce

$$\forall \epsilon, \delta \quad N_{\max} = \mathcal{O}\left(\frac{n}{\epsilon^2} \log \frac{n}{\epsilon \delta}\right)$$

$$N_{\text{sort}} = \mathcal{O}\left(\frac{n \log n}{\epsilon^2} \log \frac{n}{\epsilon \delta}\right)$$

+: Applies to all parameters

− : Factor  $\log n >$  deterministic

BTL still too rigid



Arbitrary  $p_{i,j}, 1 \leq i < j \leq n$

$\binom{n}{2}$  parameters

Can this be even done?

For 0.1 maximum: one element beats all others w.p.  $\geq 0.4$

Define:  $\{p_{i,j}\} \longleftrightarrow \text{story} \longleftrightarrow \text{child's play}$

$\{\text{rock, paper, scissors}\}$

$$p_{\text{rock,paper}} = 0 \quad p_{\text{paper,scissors}} = 0 \quad p_{\text{scissors,rock}} = 0$$

Everything loses to something with probability 1

Nothing beats everything with probability  $\geq 0.4$

No 0.1 maximum

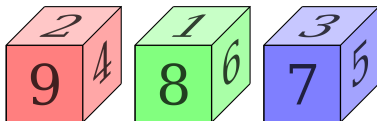
$\epsilon < \frac{1}{2} \longrightarrow \text{no } \epsilon\text{-maximum}$

Similarly, no  $\epsilon$ -ranking

Reason:  $\text{rock} < \text{paper} < \text{scissors} < \text{rock}$

Non-transitive

No maximum or ranking :-)



(a)

(b)

(c)

$$P(a > b) = P\{(9, 8), (9, 6), (9, 1), (4, 1), (2, 1)\} = \frac{5}{9} > \frac{1}{2}$$

$$P(b > c) = P\{(8, 7), (8, 5), (8, 3), (6, 5), (6, 3)\} = \frac{5}{9} > \frac{1}{2}$$

$$P(c > a) = P\{(7, 4), (7, 2), (5, 4), (5, 2), (3, 2)\} = \frac{5}{9} > \frac{1}{2}$$

Intransitive dice

Warren Buffet's favorite ice breaker, till met Bill Gates



To even have maximum & ranking  $\rightarrow$  need “Transitivity”

If “a beats b” and “b beats c” then “a beats c”

Long studied by behavioral psychologists

Satisfied by standard models (BTL, Gaussian)

Three types

Start with strong

Successively relax

See results we can get

How far we can go

$$p_{a,b} \geq \frac{1}{2}, \quad p_{b,c} \geq \frac{1}{2} \quad \implies \quad p_{a,c} \geq \max(p_{a,b}, p_{b,c})$$

$$p_{\text{Spain, Germany}} = 0.7 \quad p_{\text{Germany, Italy}} = 0.6 \quad \implies \quad p_{\text{Spain, Italy}} \geq 0.7$$

$$p_{a,b} \geq \frac{1}{2}, \quad p_{b,c} \geq \frac{1}{2} \quad \text{necessary}$$

$$p_{\text{Germany, Italy}} = 0.6 \quad p_{\text{Italy, England}} = 0.1 \quad \not\Rightarrow \quad p_{\text{Germany, England}} \geq 0.6$$

Transitivity  $\longrightarrow$  order  $\longrightarrow$  maximum and ranking

# BTL and Additive Gaussian are Transitive

$$\forall w_a \geq w_b \geq w_c$$

$$\frac{w_a}{w_a + w_c} \geq \max \left\{ \frac{w_a}{w_a + w_b}, \frac{w_b}{w_b + w_c} \right\}$$

$$\frac{w_a}{w_a + w_c} \geq \frac{w_a}{w_a + w_b} \quad \leftarrow \quad w_c \leq w_b$$

$$\frac{w_a}{w_a + w_c} \geq \frac{w_b}{w_b + w_c} \quad \leftrightarrow \quad \frac{1}{1 + w_c/w_a} \geq \frac{1}{1 + w_c/w_b} \quad \leftarrow \quad w_a \geq w_b$$

BTL transitive

Similarly for additive Gaussian Noise

Stochastic transitivity lower bounds  $p_{a,c}$ :  $p_{a,c} \geq \max(p_{a,b}, p_{b,c})$

Sometimes an upper bound is useful too

Natural upper bound: triangle inequality  $p_{a,c} \leq p_{a,b} + p_{b,c}$

Trivial for  $p_{a,b} \geq \frac{1}{2}$ ,  $p_{b,c} \geq \frac{1}{2}$

Need more relevant formulation

$$\tilde{p}_{i,j} \stackrel{\text{def}}{=} p_{i,j} - \frac{1}{2}$$

## Reformulations

$$\begin{aligned} \text{Anti-symmetry: } p_{b,a} &= 1 - p_{a,b} & 0.6 &\rightarrow 0.4 \\ \tilde{p}_{b,a} &= -\tilde{p}_{a,b} & 0.1 &\rightarrow -0.1 \end{aligned}$$

$$\begin{aligned} \epsilon\text{-preferable: } p_{a,b} &\geq \frac{1}{2} - \epsilon \\ \tilde{p}_{a,b} &\geq -\epsilon \end{aligned}$$

$$\begin{aligned} \text{Transitivity: } p_{a,b}, p_{b,c} &\geq \frac{1}{2} \rightarrow p_{a,c} \geq \max \{p_{a,b}, p_{b,c}\} \\ \tilde{p}_{a,b}, \tilde{p}_{b,c} &\geq 0 \rightarrow \tilde{p}_{a,c} \geq \max \{\tilde{p}_{a,b}, \tilde{p}_{b,c}\} \end{aligned}$$



# Stochastic Triangle Inequality (STI)

$$\tilde{p}_{a,b}, \tilde{p}_{b,c} \geq 0 \implies \tilde{p}_{a,c} \leq \tilde{p}_{a,b} + \tilde{p}_{b,c}$$

$|\tilde{p}_{i,j}|$  – distance between  $i$  and  $j$

Parametric models

Convexity:  $x \geq y \implies \frac{\partial^2 p_{x,y}}{\partial x \partial y} \geq 0$  satisfy STI

BTL, additive Gaussian models satisfy STI

## Previous Results [Yue, Joachims, '11]

With ST and TI  $N_{\max} = O\left(\frac{n}{\epsilon^2} \log \frac{n}{\epsilon \delta}\right)$

+: Broader model than BTL

–: Same  $n \log n$  as BTL

No results for ranking (BTL:  $n \log^2 n$ )

Max:  $n \log n$ ?  $n$ ? in between?

Deterministic: sequential elimination,  $n - 1$  comparisons

How simple for randomized?

1 page? 10 pages? 50 pages?

1 page

But... 10 slides

Bias  $\epsilon$ , allowed error  $\delta$

Compare( $a, b, \epsilon, \delta$ )

Compares  $a$  and  $b$  for  $\frac{1}{2\epsilon^2} \log \frac{2}{\delta}$  times

Element winning more times declared match winner

If  $|\tilde{p}(a, b)| \geq \epsilon$ , better element wins w.p.  $\geq 1 - \delta$

Observations

$\approx \frac{1}{\epsilon^2} \log \frac{1}{\delta}$  comparisons (next slide)

Inversely proportional to  $\epsilon$  and  $\delta$

Algorithm stops earlier if  $\tilde{p}(a, b) \gg \epsilon$

# comparisons  $\frac{1}{\epsilon^2} \log \frac{1}{\delta}$  used frequently

Where from?

$$n = 2$$

$X_1, X_2, \dots, X_t \sim B(\frac{1}{2} - \epsilon)$  independent

$$\overline{X} = \frac{X_1 + X_2 + \dots + X_t}{t}$$

Chernoff bound:  $P(\overline{X} \geq \frac{1}{2}) \leq e^{-\frac{t\epsilon^2}{3}}$

Want  $\delta \leq e^{-\frac{t\epsilon^2}{3}}$

$$\frac{t\epsilon^2}{3} \geq \log \frac{1}{\delta}$$

$$t \geq \frac{3}{\epsilon^2} \log \frac{1}{\delta}$$

Start with algorithm for fixed noise (different from Feige et al.)

First attempt : Sequential Elimination

Pick first element as running element  $r$

Compare  $r$  against any remaining element

Update  $r$  with winner and remove the other element

Maximum element may be compared to  $n - 1$  elements

For union bound, each comparison fails w.p.  $\leq \frac{\delta}{n}$

# comparisons per element:  $\mathcal{O}(\frac{1}{\epsilon^2} \log \frac{n}{\delta})$

Total # comparisons:  $\mathcal{O}(\frac{n}{\epsilon^2} \log \frac{n}{\delta})$

Ensured *all* comparisons correct

$n$  comparisons, each with error  $\delta/n$

Care only about comparisons involving max

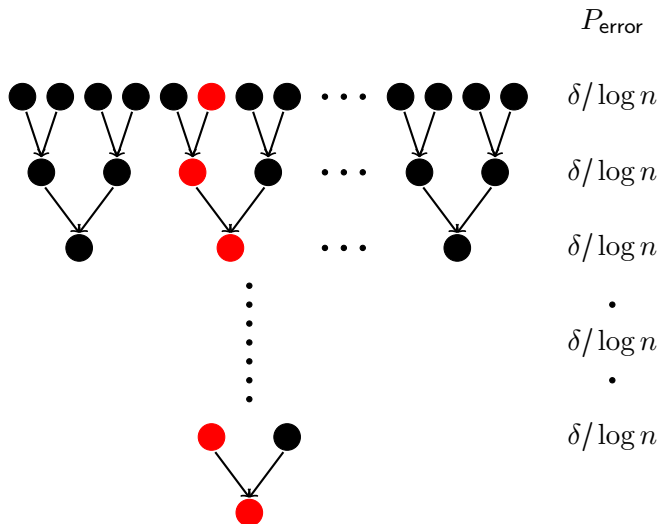
In sequential elimination max may be compared  $n - 1$  times

Can we compare max fewer times?

Knockout!

Every element, including max, compared just  $\log n$  times

$\log n$  levels, compare a pair each time





Error

By union bound  $\frac{\delta}{\log n} \cdot \log n = \delta$

Comparisons

$$n \cdot \frac{1}{\epsilon^2} \log \frac{\log n}{\delta} = \frac{1}{\epsilon^2} \cdot n \cdot \left( \log \log n + \log \frac{1}{\delta} \right)$$

Reduced factor from  $\log n$  to  $\log \log n$

Can we do better?

Error  $\delta$  is sum of error  $\delta_i$  in each stage  $i$

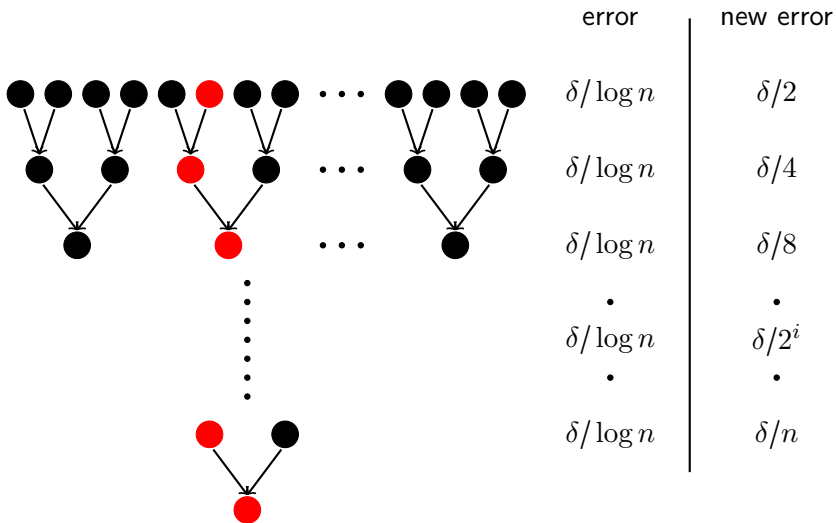
Early stages performed for many pairs

Make  $\delta_i$  large for early stages, small later, keeping sum same

$$\delta_i = \frac{1}{2^i}$$

# Knockout with Decreasing Error

Knockout:  $\log n$  levels, Compare each time



# Error & Comparisons – Another Child's Play

## Error Probability

By union bound 
$$\sum_{i=1}^{\log n} \frac{\delta}{2^i} \leq \delta$$

## Comparisons

$$\sum_{i=1}^{\log n} \frac{n}{2^i} \cdot \frac{1}{\epsilon^2} \log \frac{2^i}{\delta} = \frac{n}{\epsilon^2} \sum_i \frac{1}{2^i} \cdot \left( i + \log \frac{1}{\delta} \right) = \mathcal{O}\left( \frac{n}{\epsilon^2} \log \frac{1}{\delta} \right)$$

As easy as flipping a coin

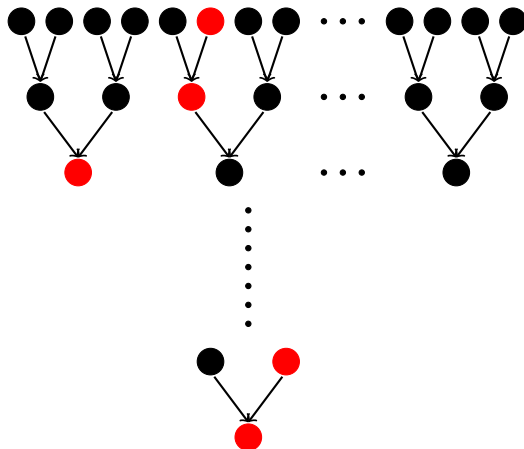
Assumed all  $|\tilde{p}_{i,j}| \geq \epsilon$

Any error has small probability

Some  $\tilde{p}_{i,j}$  may be smaller than  $\epsilon$

Errors may occur, and accumulate

# Maxing: ST +STI



| bias                | confidence   |
|---------------------|--------------|
| $c\epsilon/2^{1/3}$ | $\delta/2$   |
| $c\epsilon/2^{2/3}$ | $\delta/4$   |
| $c\epsilon/2$       | $\delta/8$   |
| $\vdots$            | $\vdots$     |
| $c\epsilon/2^{i/3}$ | $\delta/2^i$ |
| $\vdots$            | $\vdots$     |
| $c\epsilon/n^{1/3}$ | $\delta/n$   |

$p_{i,j}$  can be  $\approx \frac{1}{2}$ , max may lose

Show two properties

ST : Maxim  $m, m'$  in two successive stages satisfy  $\tilde{p}_{m,m'} < \epsilon_i$

If  $m$  wins,  $\tilde{p}_{m,m'} = 0$

If  $m$  loses to  $m'$ , # comparisons ensures  $\tilde{p}_{m,m'} < \epsilon_i$

If  $m$  loses to  $m''$ ,  $\tilde{p}_{m,m'} \leq \tilde{p}_{m,m''} < \epsilon_i$

STI: Maximum at beginning and end satisfy  $|\tilde{p}_{m,m'}| < \sum \epsilon_i$

With ST and TI

$$N_{\max} = \Theta\left(\frac{n}{\epsilon^2} \log \frac{1}{\delta}\right)$$

$$\text{For } \delta > \frac{1}{n}, N_{\text{sort}} = \Theta\left(\frac{n \log n}{\epsilon^2}\right)$$

+ : More general and stronger than BTL

Stronger than previous for ST and TI

Same as non-noisy comparisons

– : Strong transitivity too strong?



$$p_{a,b} \geq \frac{1}{2}, \quad p_{b,c} \geq \frac{1}{2} \quad \implies p_{a,c}$$

Strong Stochastic Transitivity(SST):

$$\geq \max(p_{a,b}, p_{b,c})$$

Medium Stochastic Transitivity(MST):

$$\geq \min(p_{a,b}, p_{b,c})$$

Weak Stochastic Transitivity(WST):

$$\geq \frac{1}{2}$$

$$\text{SST} \implies \text{MST} \implies \text{WST}$$

## Previous Results

| Model          | Maxing  | Ranking  |
|----------------|---|--|
| Constant Noise | $\mathcal{O}\left(\frac{n}{\epsilon^2} \log \frac{1}{\delta}\right)$          | $\mathcal{O}\left(\frac{n \log n}{\epsilon^2}\right)^*$                              |
| BTL            | $\mathcal{O}\left(\frac{n}{\epsilon^2} \log \frac{n}{\epsilon \delta}\right)$ | $\mathcal{O}\left(\frac{n \log n}{\epsilon^2} \log \frac{n}{\epsilon \delta}\right)$ |
| ST with TI     | $\mathcal{O}\left(\frac{n}{\epsilon^2} \log \frac{n}{\epsilon \delta}\right)$ | —  |

## New Results

| Model          | Maxing  | Ranking  |
|----------------|---|--|
| WT with/out TI | $\Omega(n^2)$   | $\Omega(n^2)$                                      |
| MT with/out TI | $\Theta\left(\frac{n}{\epsilon^2} \log \frac{1}{\delta}\right)$ |  |
| ST without TI  |   |  |
| ST with TI     |   | $\Theta\left(\frac{n \log n}{\epsilon^2}\right)^*$ |

\*  $\delta > 1/n$

$\Omega(n^2)$  lower bound

Simple question

Mathematical

A...

$n$  pieces in a line

Check two pieces, in order

Find if they match or not

# comparisons to assemble the puzzle?

$$O(n^2)$$

Even just to find the first piece

$$\Omega(n^2)$$

Connection to maxing

$$p_{i,i+1} = 1, p_{i,i-1} = 0, \text{ other } p_{i,j} = \frac{1}{2}$$

Weakly transitive , not medium or strong

What if no transitivity holds??

Maximum and Ranking can still be defined

Borda scores

$$s(e) \stackrel{\text{def}}{=} \Pr(e \text{ is preferable to random element}) = \frac{1}{n} \sum_i p(e, i)$$

Maximum and sorting based on these scores

$$N_{\max} = \Theta\left(\frac{n}{\epsilon^2} \log \frac{1}{\delta}\right)$$

$$N_{\text{sort}} = \Theta\left(\frac{n}{\epsilon^2} \log \frac{n}{\delta}\right)$$

## Bonus: Approximating each $p_{i,j}$ to $\epsilon$

Approximating all  $\binom{n}{2}$  pairs using  $\mathcal{O}(n \log n)$  comparisons?

If no SST+STI, ranking  $\Omega(n^2)$ , hence all  $p_{i,j}$  also  $\Omega(n^2)$

For SST+STI

Sorting:  $\Theta\left(\frac{n \log n}{\epsilon^2}\right)$  for  $\delta \geq 1/n$

Can we find all  $p_{i,j}$  using slightly more comparisons?

Yes.  $\Theta\left(\frac{n \min(n, 1/\epsilon)}{\epsilon^2} \log n\right)$  comparisons for  $\delta \geq 1/n$

$\mathcal{O}(n \log n)$  for  $n \geq 1/\epsilon$

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