

# Quantifying Uncertainty due to Ties

In Rank Correlation Coefficients

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## 1. Introduction

Rank similarity coefficients can measure agreement between different rankings of the same items. Kendall's  $\tau$  uses concordance:

$R = \begin{bmatrix} a & b & c & [d & e] \end{bmatrix}$  (a,b) is discordant,  
(a,c),(a,d),(a,e),(b,c),(b,d) and (b,e) are concordant.  
 $B = \begin{bmatrix} b & a & [e & d & c] \end{bmatrix}$

$$\tau(R, B) = \frac{\sum_{\{i,j\}} c(i,j)}{n(n-1)/2} = \frac{5}{10} \quad \text{what fraction of item pairs are concordant?}$$

Extensions for  $\tau$ : if some pairs matter more than others, a weight function can be used:

$$w(i,j) = \frac{1}{\max(i,j)} \quad \text{item pairs later in the ranking contribute less}$$

$$\tau_w(R, B) = \frac{\sum c(i,j)w(i,j)}{\{\text{weighted maximum}\}} \approx 0.485$$

Extensions for  $\tau$ : tied items. This can happen when there is uncertainty in the ranking. Variants of  $\tau$ :  $\tau^a$ ,  $\tau^b$  allow for computing  $\tau$  with ties. But they do not reflect uncertainty:

$$\tau^a(R, B) = 0.5 \quad \tau(\langle a, b, c, d, e \rangle, \langle b, a, c, d, e \rangle) = 0.8$$
$$\tau^b(R, B) \approx 0.63 \quad \tau(\langle a, b, c, d, e \rangle, \langle b, a, e, d, c \rangle) = 0.2$$

Each way of breaking ties gives a different  $\tau$

It would be useful to quantify how much variability there is in the correlation.

RBO bounds:  $RBO^{high}$   $RBO^{low}$ ,  
- can show impact of ties on correlation value

## 2. Research Question

Can we efficiently compute  $\tau^{min}/\tau^{max}$ ?  
How do  $\tau^{min}/\tau^{max}$  compare to  $\tau^a/\tau^b$ ?

## 5. Results

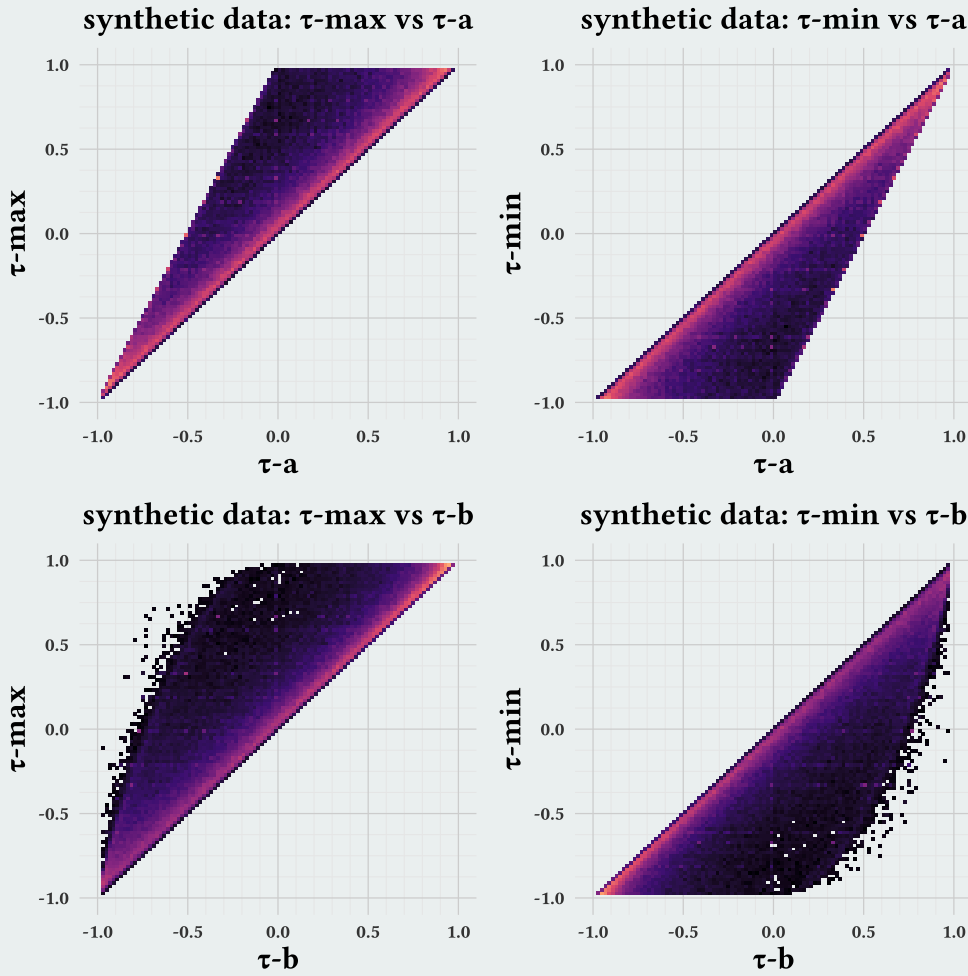
→ Compare  $\tau^a$ ,  $\tau^b$  with  $\tau^{min}$ ,  $\tau^{max}$   
→ Expectation: if variants give accurate estimates in presence of ties,  $\tau^a$  and  $\tau^b$  should be tightly bounded by  $\tau^{min}$ ,  $\tau^{max}$ .

In reality, the interval  $[\tau^{min}, \tau^{max}]$  is wide around  $\tau^a$  and  $\tau^b$ .

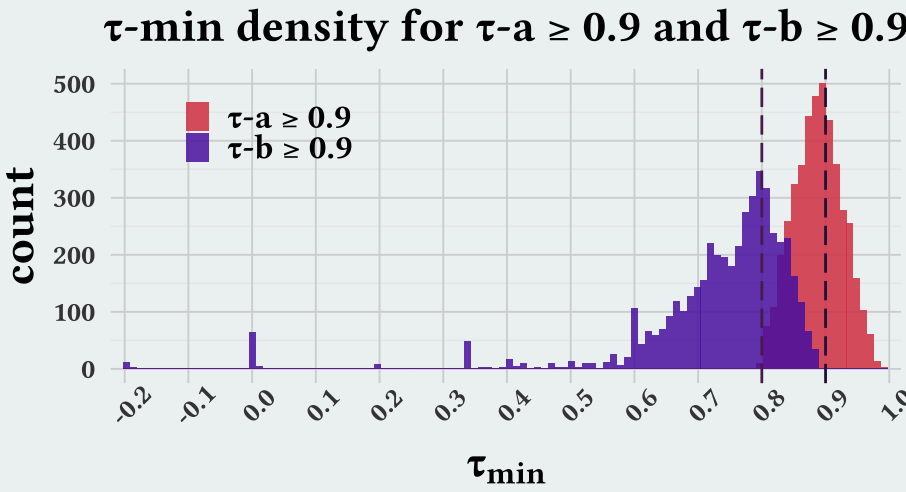
This shows that a single value ( $\tau^a$  or  $\tau^b$ ) cannot encapsulate the uncertainty in the rankings caused by the ties.

These findings were shown both for real (TREC) and synthetic datasets.

(Shown above) Synthetic data: uniform  $\tau$ , uniform tie distribution, lengths 3-150, 250k samples



## 6. Implications



|                   | $\tau^{min} < 0.9$ | $\tau^{min} < 0.85$ | $\tau^{min} < 0.8$ |
|-------------------|--------------------|---------------------|--------------------|
| $\tau^a \geq 0.9$ | 4413               | 2737                | 775                |
| $\tau^b \geq 0.9$ | 9100               | 7424                | 5163               |

When ties represent uncertainty:  
→  $\tau^a$  can misrepresent correlation  
→  $\tau^b$  can misrepresent correlation a lot  
→ Possible consequence: false positive results  
→ Problem:  $\tau^b$  is the most widely used coefficient!

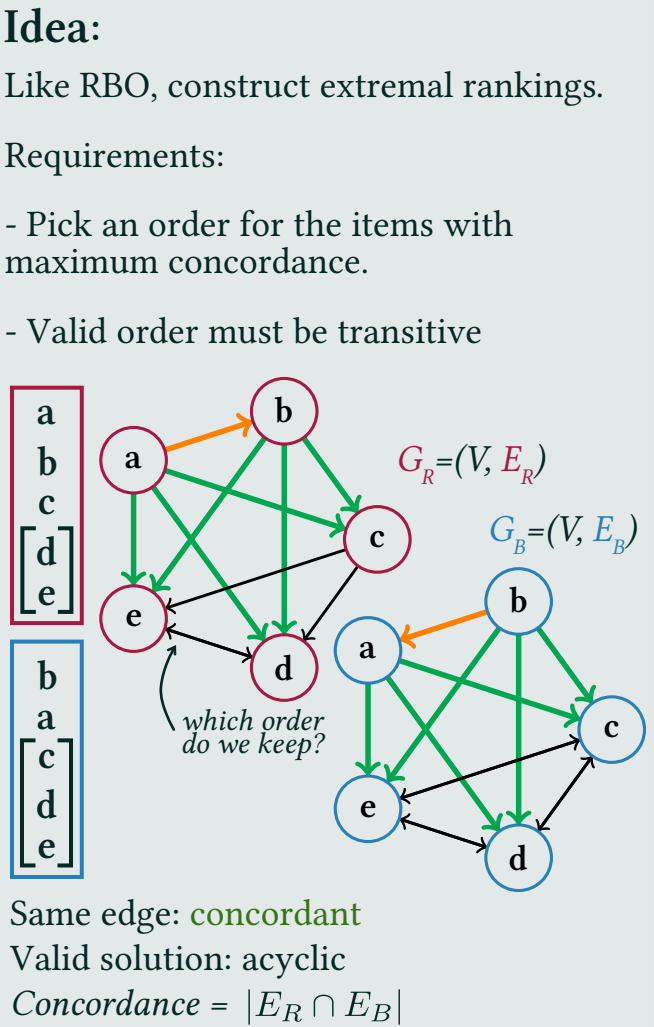
Bounds  $\tau^{min}$ ,  $\tau^{max}$  can inform decisions by supporting or contradicting  $\tau^a$ ,  $\tau^b$  estimates.

## 3. Method

How can we compute  $\tau^{min}/\tau^{max}$ ?  
→ Trying all permutations:  $O(n!)$   
→ Maximising  $\sum_{\{i,j\}} c_{R,B}(i,j)w(i,j)$  is hard (Knapsack)  
→ RBO: construct rankings to maximise overlap  
! Concordance is not the same as overlap

Overlap depends on depth  
Concordance depends only on relative order between item pairs

Concordance: sum of pairs with same relative order in both rankings



## 4. Proposed Algorithm

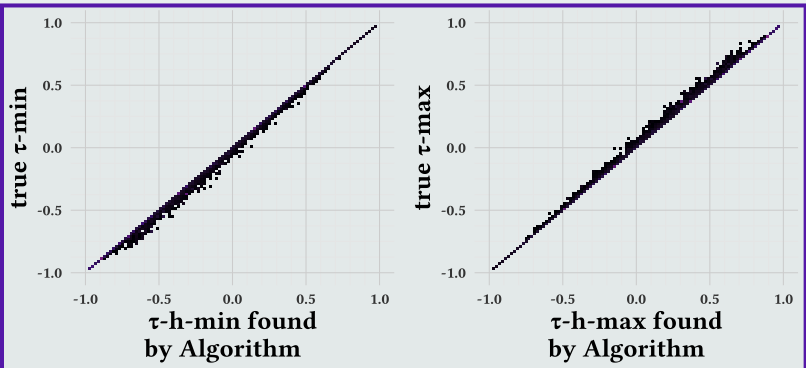
1. Take edges from  $E_R \cap E_B$

2. Check: can they be in the final ranking?  
Condition: no cycles.

3. If allowed, greedily add to solution.

- Optimal for unweighted case (proof: exchange argument)
- Minimisation: invert sorting for  $E_R$
- Weighted case: sort  $E$  by descending weight
  - $\tau_{AP}$ : optimal,
  - $\tau_h$ : approximation

$G_R : (S, E_R) \leftarrow R$   
 $G'_R : (S, E'_R) \leftarrow (S, \emptyset)$   
 $G_B : (S, E_B) \leftarrow B$   
 $G'_B : (S, E'_B) \leftarrow (S, \emptyset)$   
 $\vec{xy} \in G_R \quad G = (S, E_R \cup \{\vec{xy}\})$   
 $\vec{xy} \in G'_R \quad G'_R \leftarrow (S, E'_R \cup \{\vec{xy}\})$   
 $\vec{xy} \in G_B \quad G = (S, E_B \cup \{\vec{xy}\})$   
 $\vec{xy} \in G'_B \quad G'_B \leftarrow (S, E'_B \cup \{\vec{xy}\})$   
 $R', B' \leftarrow R, B$   
 $l = |S|$   
 $tg \quad \lambda xy : x < y \iff \vec{xy} \in E'_R$   
 $tg \quad \lambda xy : x < y \iff \vec{xy} \in E'_B$   
 $R', B'$



## 7. Conclusion

Uncertainty bounds are important for  $\tau$  correlation with ties, just as in RBO.

We proposed an algorithm for computing  $\tau^{min}$ ,  $\tau^{max}$

- Does not generalise to all weight functions.
- Good approximations for reasonable weighting schemes.

Using uncertainty bounds can help inform decisions, when ties in rankings are induced by uncertainty.

## Future Work

- Explore missing constraints on w
- Extend the algorithm to provide a distribution of values, akin to the work shown for RBO.
- Implementation in statistical software packages
- Encourage researchers to question how the tools and metrics they use reflect the properties of the rankings in their field.