Condorcet Fusion: an implementation

Improved Retrieval through Rank Fusion

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Introduction



This project implements some ideas coming from a paper by **Aslam** and **Montague**.

These authors applied Social Choice Theory to Rank Fusion.

They claim that their algorithm beats the ones by **Fox** and **Shaw** most of the time.

bgWave 2 c

Starter: basic fusion strategies



Fox and Shaw devised some simple, score-based fusion formulae:

CombMAX

fused score = max(scores)

CombMIN

fused score = min(scores)

CombSUM

fused score = sum(scores)

Starter: basic fusion strategies



Fox and **Shaw** devised some simple, score-based fusion formulae:

CombMED

fused score = median(scores)

CombANZ

 $fused\ score = \frac{CombSUM}{\#nonZeroScores}$

CombMNZ

 $fused\ score = CombSUM \times \#nonZeroScores$

Are we comparing apples and oranges?



Of course we can't perform any sensible fusion without **normalization**.

Lee came up with two handy formulae:

Max Norm

$$max_norm = \frac{old_score}{max_score}$$

Min Max Norm

$$min_max_norm = \frac{old_score - min_score}{max_score - min_score}$$

What about voting systems?



Metaphor: voters are retrieval systems
candidates are documents

The algorithm inspired by the **Condorcet method** starts from a document comparator known as Simple Majority Runoff.

"Simple Majority Runoff"



Imagine a **contest** between two documents, then the rule below sounds sensible, doesn't it?:

The number of a document's votes is directly related to the number of times it ranks above the other.

A theoretical approach



The first proposed algorithm isn't well-suited for large collections. It's based on the computation of **Hamiltonian paths** on a graph and has a $O(n^2k)$ time complexity.