A new integer programming formulation and refined social choice property for expediting the solution to the consensus ranking problem

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Introduction	Consensus ranking
 Introduction Group decision-making has been studied extensively since the shaping of democratic society Many people devoted their efforts to develop a fair and consistent system that aggregates the opinion of each individual to make better social decisions 	 Limitation of ratings The rating scales of two individuals are, in general, not comparable Consensus ranking problem (i.e. ranking aggregation) The consensus ranking problem is at the center of many group decision-making processes

Example of group decision-making : University rankings, Proposal funding decisions, Online product review	It entails finding an ordinal vector or ranking of a set of competing objects that minimizes disagreement with a profile of preferences (represented as ranking vectors)
Two frameworks of the consensus ranking problem	Computational difficulties
 Given a set of <i>m</i> rankings <i>a</i>¹,, <i>a^m</i> ∈ Ω, the median ranking is the optimal solution to the following problems: The distance-based ranking aggregation problem min ∑^m_{r∈ΩC} d(aⁱ, r) Objective is to minimize the cumulative distance or disagreement The correlation coefficient-based ranking aggregation problem max ∑^m_{r∈ΩC} τ(aⁱ, r) Objective is to maximize the cumulative correlation or agreement 	• Obtaining even just one consensus or median ranking via correlation-based methods (or the equivalent axiomatic distance-based methods) is an NP-hard problem. (Bartholdi et al. 1989) $ \frac{number \text{ of solutions}}{n=2} = \frac{3 \ge \frac{1.442^{\circ}21}{2}}{n=3} = (1,1), (1,2), (2,1)} $ $ \frac{n=10}{n=10} = \frac{1.2442^{\circ}21}{102247563} = \frac{1.0222 \text{ billions}}{1.02247563} = 1.0222 \text{ billions}} $
Approach 1 – Integer programming formulation (IP)	Approach 1 – Computational experiments

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$\begin{array}{ll} \underset{y}{\text{maximize}} & \sum_{i} \sum_{j} c_{ij} (2y_{ij} - 1) \\ \text{subject to} & y_{ij} - y_{kj} - y_{ik} \geq -1 \\ & y_{ij} + y_{ji} \geq 1 \\ & 0 \leq y_{ij} \leq 1 \\ & y_{ij} \in \mathbb{Z} \end{array} \qquad \begin{array}{ll} i, j = 1, \dots, n; & i \neq j \neq k \neq i \\ & i, j = 1, \dots, n; & i \neq j \\ & i, j = 1, \dots, n; \\ & i, j = 1, \dots, n; \end{array}$	 ↑ (V) ⇒ ↑ (Computational time improvement) For all ranges of the number of objects, the computational time improves by 70% Computational time improvement Computational time improvement Output display to the improvement (set)
Approach 2 – Condorcet criterion	Approach 2 – HPC techniques applicability
 Condorcet criterion (Condorcet 1785) Extended Condorcet criterion (Truchon 1998) Generalized Condorcet criterion Definition: If a decisive majority ranks every objects v_i ∈ V ahead of every objects v_j ∈ V' (i.e. a less people believes v_j ∈ V' ahead of v_i ∈ V), then every objects v_i ∈ V must be ahead of v_j ∈ V' in the consensus ranking 	• Each independent object set can be solved by using HPC techniques $\begin{array}{c} \mathbf{Input} \\ \mathbf{rankings} \\ (r judges, n objects) \end{array} \qquad \begin{array}{c} \mathbf{Subset} \\ \mathbf{partitioning} \\ (p airwise comparison) \end{array} \qquad \begin{array}{c} \mathbf{Condorcet} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{decomposition} \\ (p aubsets) \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ \mathbf{Det order } \\ \mathbf{Det order } \\ \mathbf{Partial} \\ \mathbf{Det order } \\ \mathbf{Partial} \end{array} \qquad \begin{array}{c} \mathbf{Partial} \\ Partia$
Approach 2 – Computational experiments	Conclusions and future work
<section-header> ↑ (V) ⇒ ↑ (Computational time) ↑ (φ) ⇒ ↑ (Computational time) Even for φ <0.5, the difference in computational times increases Are computational time informations Are computational time informations Are co</section-header>	 Summary We derived a property that aligns better with Kemeny-Snell distance We introduced an IP formulation to expedite solution process Future work More computational experiments will be conducted on various types of preference data and ranking models We can apply distributed computing to obtain optimal rankings separately and then combine sub-problem solutions