

Multidimensional welfare rankings under weight imprecision

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Multidimensional welfare

- ▶ Many aspects of social well-being are intrinsically multidimensional.
- ▶ E.g., development, poverty, inequality cannot be fully captured by simple, exclusively income-based, measures.
- ▶ Originating in the powerful conceptual writings of Amartya Sen, the idea of multidimensional well-being has had a deep influence on academia as well as policy.
- ▶ Indeed, the primary tools that the UN uses to measure development and poverty, the Human Development and Multidimensional Poverty indices (HDI, MPI) reflect the above concerns.

Multidimensional welfare measurement

- ▶ Need to compare and eventually order possible alternatives (countries, policies, etc) on the basis of multidimensional information.
- ▶ Welfare indices (such as the HDI and MPI) approach this task by integrating the various dimensions of well-being into a scalar measure. This is generally achieved by assigning **weights** to the different dimensions and, in some fashion, **aggregating** over them.
- ▶ Often these choices are not grounded in economic theory or a coherent normative framework, sparking backlash (Ravallion, 2012).
- ▶ For instance, there is disagreement as to whether multidimensional poverty should be communicated through a “dashboard” of indices (Ravallion, 2012), or an aggregate scalar measure such as the MPI (Alkire and Foster, 2011).

The issue of weights

- ▶ Assume that a functional form for the aggregation function is in place (justified by normative desiderata), but weights are undetermined.
- ▶ Their choice can be fraught with complex philosophical and practical dilemmas, despite a multitude of proposed techniques (Foster and Sen, 1997; Decancq and Lugo, 2013).
- ▶ Indeed, there is frequently no single “right” weighting scheme and we are justified, if not compelled to, consider the effect of many different weights at once.
- ▶ Such an analysis would serve two goals:
 - (a) to examine how robust a given ranking of alternatives is to changes in weights, and
 - (b) to determine a compromise ranking that is in some sense “optimal” in the presence of weight imprecision.

Previous work

- ▶ Monte Carlo simulation in the context of broader uncertainty/sensitivity analyses (Saisana et al. 2005).
- ▶ Duclos et al. (2006, 2011) studied multidimensional poverty/inequality comparisons using ideas from stochastic dominance. They established an analytic criterion for determining whether a (pairwise) poverty comparison is robust within a large class of indices.
- ▶ Anderson et al. (2011) imposed monotonicity and quasiconcavity on the aggregation function and derived bounds on welfare levels.
- ▶ Foster et al. (2013) studied linear indices and parameterized weight imprecision with the ϵ -contamination model of Bayesian statistics. Focused on pairwise relations.
- ▶ Pinar et al. (2013) examined the HDI index and used ideas from stochastic dominance to determine the set of weights that results in best-case human development over time.

This paper's contribution and added value

- ▶ I propose a theoretical framework that yields consensus rankings in the presence of weight imprecision, which is formally rooted in the social choice/voting literature.
- ▶ The approach goes beyond existing work in the following ways:
 - (i) It produces a set of **complete** consensus rankings of the alternatives, not welfare bounds or pairwise dominance relations.
 - (ii) It can be justified on **axiomatic** grounds (thus guarding against charges of being ad-hoc).
 - (iii) It can be efficiently implemented in **high-dimensional** settings of multiple alternatives and welfare criteria (unlike techniques based on stochastic dominance).

The paper in a nutshell

- ▶ Consider a vector of weights as a **voter** and a continuum of weights as an **electorate**.
- ▶ With this voting construct in mind, **Kemeny's rule** from social choice theory is introduced as a means of aggregating the preferences of many plausible choices of weights.
- ▶ The axiomatic characterization of Kemeny's rule due to Young and Levenglick (1978) and Young (1988) is shown to extend to the present context.
- ▶ An efficient graph-theoretic algorithm is developed to compute or approximate the set of Kemeny optimal rankings.
- ▶ Further analytic results are derived for a relevant special case of the model.
- ▶ The model is applied to the ARWU index of Shanghai University, a popular and controversial index ranking academic institutions across the world. High problem dimensionality means it is a good "proof of concept".

Model description

- ▶ Set of alternatives \mathcal{A} indexed by $a = 1, 2, \dots, A$ and set of indicators \mathcal{I} indexed by $i = 1, 2, \dots, I$.
- ▶ Let $x_{ai} \in [0, 1]$ denote alternative a 's normalized value of indicator i , $\mathbf{x}_a \in \mathbb{R}^I$ its “achievement” (column) vector, and $\mathbf{X}_{\mathcal{A}} \subset [0, 1]^{I \times A}$ the resulting achievement matrix.
- ▶ Performance across indicators is weighted by a vector \mathbf{w} belonging in the simplex $\Delta^{I-1} = \{\mathbf{w} \in \mathbb{R}^I : \mathbf{w} \geq \mathbf{0}, \sum_{i=1}^I w_i = 1\}$.
- ▶ Welfare corresponding to achievement vector \mathbf{x} and \mathbf{w} is given by a real-valued function $u(\mathbf{x}, \mathbf{w})$.
- ▶ The welfare function is purposely left general in order to accommodate many different multidimensional concepts.

Weight imprecision

- ▶ Now, define an **importance** function f on the simplex Δ^{l-1} , satisfying $f(\mathbf{w}) \geq 0$ for all $\mathbf{w} \in \Delta^{l-1}$ and $0 < \int_{\Delta^{l-1}} f(\mathbf{w}) d\mathbf{w} < +\infty$.
- ▶ f models imprecise beliefs regarding the “correct” set of weights to use.
- ▶ It may be set *a priori* by the decision-maker, or it may be arrived at by aggregating the views of agents to be ranked.
- ▶ In the case of the HDI, f could be set in the following manner: ask each country c to provide its importance function f_c on Δ^2 and then set $f = \sum_c f_c$.
- ▶ Work with continuous f , but model can be straightforwardly extended to account for discrete importance functions on a finite (or countably infinite) subset of weights belonging in Δ^{l-1} .

Weights as voters

- ▶ Define a **profile** L to be a triplet $L = (\mathbf{X}_A, f, u)$, and let \mathcal{L} denote the space of all profiles.
- ▶ Given a profile L , suppose we think of weight vector \mathbf{w} as an imaginary **voter** who (weakly) prefers a_i over a_j if and only if $u(\mathbf{x}_{a_i}, \mathbf{w}) > u(\mathbf{x}_{a_j}, \mathbf{w})$ ($u(\mathbf{x}_{a_i}, \mathbf{w}) \geq u(\mathbf{x}_{a_j}, \mathbf{w})$).
- ▶ Thus, voter \mathbf{w} 's preferences will be expressed as a (possibly partial) **ranking** of the alternatives.
- ▶ Construct an **electorate** of voters by considering each $\mathbf{w} \in \Delta^{I-1}$ and introducing $f(\mathbf{w})$ copies of itself. Thus, the greater $f(\mathbf{w})$ is, the more voters holding \mathbf{w} 's preferences are introduced. This results in a continuum of voters $\mathcal{E}(f)$ of finite measure.

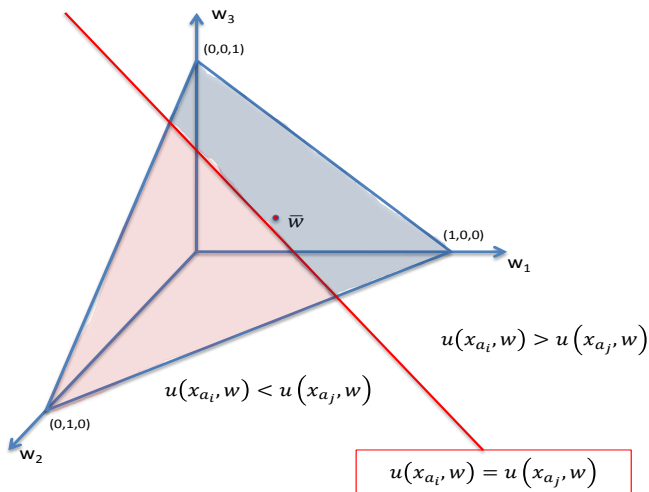
Connections with social choice

- ▶ What would constitute a “good” way of aggregating the preferences of all weight vectors, suitably weighted by the importance a decision maker places on them?
- ▶ More abstractly: Given a set of individual ranked preferences, what voting rule should society use to determine a consensus ranking? What properties should a compromise solution aspire to satisfy? What tradeoffs need to be reconciled?
- ▶ Fundamental questions, whose modern roots lie in the work of Condorcet and Borda.
- ▶ Arrow’s impossibility theorem is a classical result along this vein.

Election matrices

- ▶ Given a profile $L = (\mathbf{X}_{\mathcal{A}}, f, u)$, define the **election (proportion) matrix** \mathbf{Y}^L (\mathbf{V}^L).
- ▶ Y_{ij}^L (V_{ij}^L) defines the **net majority (proportion)** of voters within $\mathcal{E}(f)$ preferring a_i to a_j . Matrix \mathbf{Y}^L (\mathbf{V}^L) summarizes this information for all pairs of alternatives.
- ▶ Generally, \mathbf{Y}^L and \mathbf{V}^L need to be computed numerically.
- ▶ However, analytic solutions are possible for some compelling special cases (see Section 5 in paper).

An example: $f(\mathbf{w}) \equiv 1$ and u linear



- ▶ $\mathcal{E}(f)$ equals the entire simplex with uniform importance.
- ▶ Y_{ij}^L is the difference between the volumes of the **BLUE** and **RED** regions.
- ▶ V_{ij}^L is the ratio of the volumes of the **BLUE** region and the entire simplex.

Kemeny's rule

- ▶ If R_1 and R_2 are rankings, their **pairwise disagreement** (or Kendall- τ distance) is given by the number of pairs (a_i, a_j) such that $R_1(a_i) > R_1(a_j)$ and $R_2(a_i) < R_2(a_j)$.
- ▶ Given a set of voters who each submit ordered preferences on a set of alternatives, **Kemeny's rule** (Kemeny, 1959) produces a ranking that **minimizes the sum of its pairwise disagreements** with respect to voter preferences.
- ▶ Applying this concept to infinite electorate $\mathcal{E}(f)$, the Kemeny-optimal set of rankings K^L can be simplified to ($\mathcal{R}_{\mathcal{A}}$ denotes the set of rankings of alternatives in \mathcal{A})

$$K(L) \equiv K^L = \arg \min_{R \in \mathcal{R}_{\mathcal{A}}} \sum_{(a_i, a_j) \in \mathcal{A} \times \mathcal{A}} \mathbf{1}\{R(a_i) < R(a_j)\} Y_{ji}^L.$$

Normative analysis

- ▶ A **rule** is a function from the set of profiles to the set of nonempty subsets of rankings.
- ▶ Can we justify axiomatically the adoption of rule K as a means of ranking alternatives? In what sense would it be “better” than other methods we could employ?
- ▶ Yes, b/c it turns out that K is the only rule satisfying a set of desirable axioms.

Anonymity, Neutrality, Unanimity, Condorcet

[For rigorous definitions of the following Axioms please see the paper.]

- Axiom 1.** A rule ϕ is *anonymous* if it depends only on the number of voters submitting ranking R as their preference, for all rankings R .
- Axiom 2.** A rule ϕ is *neutral* if the identity of an alternative does not affect the rank it receives.
- Axiom 3.** A rule ϕ is *unanimous* if, when all weights submit the same ranking of the alternatives, then the rule picks this ranking.
- Axiom 4.** A rule ϕ is *extended-Condorcet* if it respects the majority wishes of the electorate, whenever these do not involve intransitivities (i.e., situations where a majority of voters prefer A to B , B to C and C to A).

Reinforcement

Axiom 5. A rule ϕ satisfies *reinforcement* if it acknowledges and reinforces pre-existing consensus, thus imposing a degree of consistency to the aggregation process.

- ▶ Consider the HDI, and suppose Africa and Europe have completely differing opinions regarding the weights of the three dimensions of the HDI.
 - ▶ African countries only want to consider weights \mathbf{w} s.t.
 $w_H > w_I > w_E$.
 - ▶ European countries only want to consider weights \mathbf{w} s.t.
 $w_E > w_H > w_I$.
- ▶ Suppose the UN chooses a method of ranking countries that, when considering the opinions of A and E *separately* leads to the same consensus ranking. In that case, reinforcement requires that the UN's method, when considering the preferences of A and E *jointly*, not disturb their pre-existing consensus.

Local independence of irrelevant alternatives

Axiom 6. *A rule ϕ satisfies **local independence of irrelevant alternatives (LIIA)** if the relative order of alternatives that are ranked “together” in a consensus ranking does not change, when we apply the rule to the restricted problem that focuses just on these alternatives and ignores all others.*

- ▶ Usually such contiguous intervals correspond to meaningful categories of alternatives.
- ▶ Suppose we rank the 100 best universities in the world. We would prefer the relative ordering of the top 20 (representing, say, Tier 1 institutions), to remain unchanged if we re-apply the rule ignoring those universities ranked 91-100, 51-100, or even the entire 21-100 for that matter.

The axiomatic characterization

Theorem 1

- (i) On the domain of profiles \mathcal{L} , K satisfies **anonymity, neutrality, reinforcement, extended-Condorcet, unanimity, and LIIA**.
 - (ii) Let \mathcal{Y}^Q denote the set of rational skew-symmetric matrices whose rows and columns are indexed by the elements of \mathcal{A} . On the restricted domain $\mathcal{L}^Q = \{L \in \mathcal{L} : \mathbf{Y}^L \in \mathcal{Y}^Q\}$, K **uniquely satisfies anonymity, neutrality, reinforcement, unanimity, and LIIA**.
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- ▶ Largely a restatement of results by Young (1974, 1988), Young and Levenglick (1978).
 - ▶ But care must be taken to ensure that their proofs extend to the current, non-standard setting.

(Important) computational issues

- ▶ Unfortunately, computing K is **NP-hard** (Bartholdi et al., 1989), even when the number of indicators is just four (Dwork et al., 2001).
- ▶ The main difficulty arises from **Condorcet cycles**, which imply intransitive majority pairwise preferences. Thus, it is important to identify and, in some fashion, resolve these cycles.
- ▶ Using classical results from discrete algorithms (Tarjan, 1972) and recent approximation algorithms (Van Zuylen and Williamson, 2009), I propose a graph theoretic algorithm that computes or provides a provably-good approximation of K (see Section 4 of the paper).
- ▶ If the size of Condorcet cycles is **“small enough”**, then one gets an exact solution.

A special case of the model I: generalized weighted means

- ▶ A family of welfare functions that is particularly popular in many policy contexts are known as **generalized weighted means** (Decancq and Lugo, 2013).
- ▶ Parameterized by $\gamma \in \mathfrak{R}$, they are denoted by u^γ and satisfy

$$u^\gamma(\mathbf{x}, \mathbf{w}) = \begin{cases} \left(\sum_{i=1}^I w_i x_i^\gamma \right)^{\frac{1}{\gamma}} & \gamma \neq 0, \\ \prod_{i=1}^I x_i^{w_i} & \gamma = 0. \end{cases}$$

- ▶ $\frac{1}{1-\gamma}$ = elasticity of substitution between achievements.
- ▶ When $\gamma = 1(0)$ we recover the weighted arithmetic (geometric) mean. As $\gamma \rightarrow +\infty(-\infty)$, $u^\gamma(\mathbf{x}, \mathbf{w})$ converges to the maximum (minimum) coordinate of \mathbf{x} .

A special case of the model II: ϵ -contamination

- ▶ We are given an initial vector of weights $\bar{\mathbf{w}}$.
- ▶ Suppose that we are willing to grant equal consideration to weights deviating from $\bar{\mathbf{w}}$ that belong to the set W^ϵ , where

$$W^\epsilon = (1 - \epsilon)\bar{\mathbf{w}} + \epsilon\Delta^{I-1} = \left\{ \mathbf{w} \in \mathbb{R}^I : \mathbf{w} \geq (1 - \epsilon)\bar{\mathbf{w}}, \sum_{i=1}^I w_i = 1 \right\}.$$

- ▶ Parameter $\epsilon \in [0, 1]$ measures the **imprecision** associated with $\bar{\mathbf{w}}$. Can be modeled with an importance function f^ϵ assigning weight 1 to all $\mathbf{w} \in W^\epsilon$ and 0 everywhere else.
- ▶ Originally developed in Bayesian analysis (Berger and Berliner, 1986), this way of parameterizing imprecision is referred to as **ϵ -contamination**. Studied also in micro theory (Nishimura and Ozaki, 2006; Kopylov, 2009).
- ▶ First introduced by Foster et al. (2013) in the context of composite indices of welfare.

How could ϵ be set?

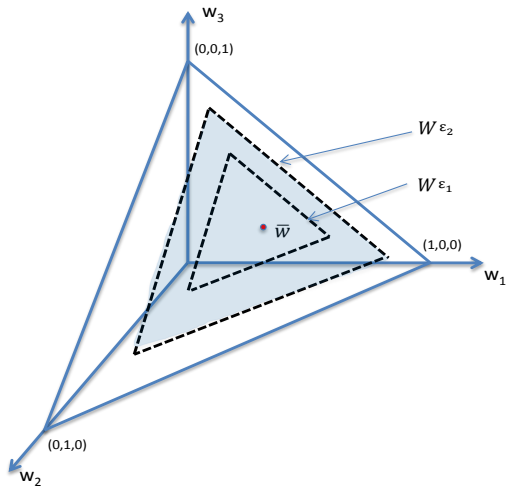
- ▶ Statistically, the parameter ϵ may be interpreted as the amount of error attached to the prior $\bar{\mathbf{w}}$.
- ▶ In our context, the choice of ϵ is largely subjective and should be decided in close consultation with the policy makers.
- ▶ Nevertheless, the simple structure of ϵ -contamination may inform this process by shedding light on the implications of different choices.
 - (i) Places a **uniform bound** on allowable percentage decrease of an indicator's weight with respect to $\bar{\mathbf{w}}$, i.e.

$$\left\{ \frac{w_i}{\bar{w}_i} \geq 1 - \epsilon, \quad \forall i \in \mathcal{I} \right\} \Leftrightarrow \left\{ w_i \in [\bar{w}_i - \epsilon \bar{w}_i, \bar{w}_i + \epsilon(1 - \bar{w}_i)], \quad \forall i \in \mathcal{I} \right\}.$$

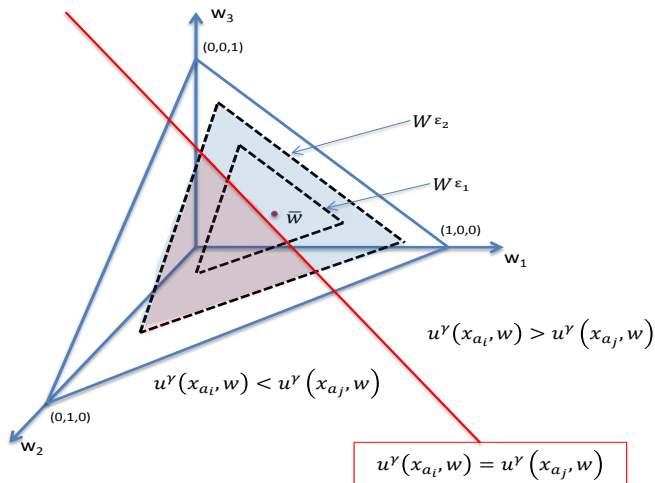
- (ii) Serves as a guide for policy makers who wish to “cover” a **target percentage** of all possible vectors of weights.

$$\frac{\text{Vol}(W^\epsilon)}{\text{Vol}(\Delta^{I-1})} = \epsilon^{I-1}.$$

A graphical illustration of ϵ -contamination



Pairwise comparisons when $u = u^\gamma$ and $f = f^\epsilon$



- ▶ Given \bar{w} , $\epsilon > 0$ and $\gamma \in \mathfrak{R}$, let $V_{ij}^{\epsilon, \gamma}$ denote the proportion of weights favoring a_i over a_j .
- ▶ $V_{ij}^{\epsilon_1, \gamma}$ ($V_{ji}^{\epsilon_1, \gamma}$) is the ratio of the volume of the smaller **BLUE** (**RED**) region to the volume of the inner triangle. Analogously for ϵ_2 .

Theorem 2

When a_i and a_j do not yield identical welfare under $\bar{\mathbf{w}}$, $V_{ij}^{\epsilon, \gamma}$ varies **monotonically** in the imprecision ϵ attached to $\bar{\mathbf{w}}$. It is decreasing if a_i initially dominates a_j and increasing if it is dominated by it. Conversely, when $\bar{\mathbf{w}}$ yields welfare for a_i and a_j , then $V_{ij}^{\epsilon, \gamma}$ remains constant as we vary ϵ .

Theorem 3

Simple geometric structure allows us to exploit the results of Lawrence (1991) and provide an explicit formula for $V_{ij}^{\epsilon, \gamma}$.

Proof of concept: the ARWU index

- ▶ Shanghai University's Academic Ranking of World Universities (ARWU), a popular composite index measuring research excellence in academic institutions.
- ▶ 6 criteria: (1) No. alumni winning Nobel prizes/Fields medals, (2) No. faculty winning Nobel prizes/Fields medals; (3) highly-cited researchers; (4) papers in Nature/Science; (5) papers indexed in leading citation indices; (6) per capita academic performance.
- ▶ ARWU score $u(\mathbf{x}, \mathbf{w}) = \sum_{i=1}^6 w_i x_i$, and $\mathbf{w}_{ARWU} = (.1, .2, .2, .2, .2, .1)$.
- ▶ Despite its increasing influence and popularity, the ARWU index has been criticized on many grounds, including its non-robustness to changes in weights (Saisana et al., 2011).
- ▶ The controversy surrounding this index, in combination with its high dimensionality (100 universities, 6 criteria) make it a good application area for the model.

Applying the model

- ▶ Focus on the top-100 universities reported in the 2013 ARWU rankings, denoted by \mathcal{A}_{100} .
- ▶ I consider imprecision over the ARWU index weights via ϵ -contamination with $\bar{\mathbf{w}} = \mathbf{w}_{ARWU}$ and $\epsilon \in \{1/6, 1/3, 1/2\}$.
- ▶ For convenience, denote by K^ϵ the Kemeny-optimal ranking of universities in \mathcal{A}_{100} when applying the method for different values of ϵ .
- ▶ Differences $K^0 - K^\epsilon$ grow as we increase ϵ , and are much more pronounced for universities in the 51-100 range.
- ▶ There are moreover a handful of really substantial swings in rankings. For instance, the ENS-Paris was ranked 71st in the official 2013 ARWU ranking, whereas its Kemeny-optimal ranks for $\epsilon = 1/6, 1/3, 1/2$ are 62, 54, and 49, respectively.

Numerical application: ARWU index

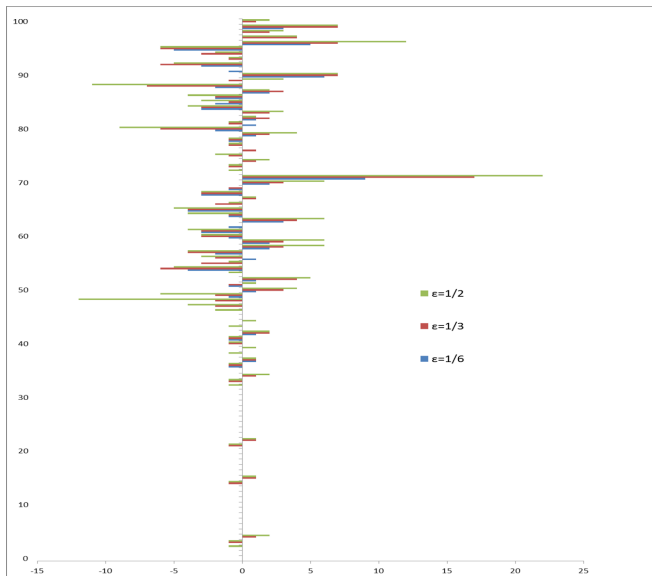


Figure : 2013 ARWU Top-100: $K^0 - K^\epsilon$.

Recap

- ▶ Judgments based on composite indices of welfare depend, sometimes critically, on how different dimensions of performance are weighted.
- ▶ As there is frequently no single “right” way to assign such weights, it is important to take this imprecision into account in a systematic and transparent manner.
- ▶ In this paper I have drawn from the theory of social choice to present a procedure for determining a ranking of the relevant alternatives that is normatively compelling and statistically interpretable.
- ▶ Developed graph-theoretic algorithm to implement rule and the applicability of the proposed framework was illustrated through a numerical example based on Shanghai University's ARWU index.
- ▶ Broader connections with decision-theoretic models of Knightian uncertainty can be explored.