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In multiple-criteria evaluation schemes, rank disequilibrium occurs when an evaluatee is rated higher than other evaluatees on some criteria and lower than other evaluatees on other criteria. In this article, we investigate rank disequilibrium as it relates to the problem of aggregating scores on individual criteria into an overall evaluation. We adopt an axiomatic approach, defining the notion of a *rank aggregation function* and proposing a set of desirable properties — namely, independence, monotonicity, inclusivity, consistency, and equity — that rank aggregation functions may or may not satisfy. We prove that when there are more than three possible scores on each criterion, it is impossible to define a rank aggregation function that satisfies all of these properties. We then investigate potential resolutions to the problems posed by rank disequilibrium.

1. Introduction

According to Pruitt and Kim [2004, p. 24], *rank disequilibrium*, or *status inconsistency*, “exists when there are multiple criteria for assessing people’s merit or contributions, and some people are higher on one criterion and lower on another criterion than others”. Status inconsistency has been studied at length within the sociology and conflict resolution literature, particularly in regards to social stratification, intergroup conflict, and aggression (for example, [Engel 1988; Evan and Simmons 1969; Galtung 1964; Hernes 1969; Kriesberg 1998; Muller and Jukam 1983; Segal et al. 1970]). This article focuses on the phenomenon of rank disequilibrium within the specific context of multiple-criteria evaluation.

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To illustrate, consider the common practice of evaluating employees by first assigning scores on a variety of criteria and then determining an overall performance rating by aggregating these scores in some way. For example, it is common within academia for faculty to be evaluated on teaching, research, and service. But what happens when Professor Smith is rated satisfactory in teaching, outstanding in research, and outstanding in service, while Professor Jones is rated outstanding in teaching, satisfactory in research, and satisfactory in service? Who should receive a higher overall rating? On the one hand, Professor Smith could argue that she is entitled to the higher rating, since she received higher marks on two of the three criteria. On the other hand, Professor Jones could argue that teaching is the most important criterion, and so she should receive the higher rating. The inconsistency between each professor's scores on the various criteria presents challenges to the evaluator who must aggregate the scores and determine an overall evaluation. Indeed, it is conceivable that, regardless of the final evaluations, one of the two professors will perceive that she has been treated inequitably.

Much of the prior research on rank disequilibrium has been empirical or philosophical in nature. In the present work, we take an axiomatic approach. In [Section 2](#), we introduce the definitions that form the basis of our model. We define the notion of a *rank aggregation function*, which is our primary mechanism for aggregating marks on individual criteria into a single, overall evaluation. We then propose several desirable properties of rank aggregation functions — including *independence*, *monotonicity*, *inclusivity*, *consistency*, and *equity* — giving examples to illustrate each. Rank aggregation functions that satisfy all of these properties are said to be *ideal*.

In [Sections 3](#) and [4](#), we consider conditions for the existence and uniqueness of ideal rank aggregation functions. We demonstrate the existence of a unique ideal rank aggregation function in the case where only three scores are possible for each criterion. We then prove that when more than three scores are allowed, no such function exists.

In [Sections 5](#) and [6](#), we consider potential resolutions to the nonexistence (in most cases) of ideal rank aggregation functions. We show that suitable alternatives can be found if we are willing to sacrifice independence or accept a weaker form of equity.

In [Section 7](#), we summarize our work and discuss questions for further research.

2. Definitions and examples

Our model assumes that a finite number of *valuees* will receive one of a finite number of ratings, or *scores*, on each of a finite number of *criteria*.

We let C denote the set of evaluation criteria, where $|C| = n$. We let $Z = \{z_1, z_2, \dots, z_m\}$ denote the set of possible scores for each criterion, where $m \geq 2$.

We assume that Z is totally ordered, with the ordering relation denoted by \succeq , where $z_m \succ z_{m-1} \succ \cdots \succ z_2 \succ z_1$. (Intuitively, z_m corresponds to the best score and z_1 corresponds to the worst score.)

A *score profile*, or *profile* for short, is an n -tuple of elements of Z — that is, an element of the Cartesian product

$$X = Z \times Z \times \cdots \times Z = Z^n.$$

This set X is called the *profile space*. For any profile $x \in X$, we let x_c denote the score from the c -th criterion (coordinate) of x . If, for some $x, y \in X$, we have $x_c \succeq y_c$ for all $c \in C$, then x is said to *dominate* y , denoted $x \gg y$. For any $x \in X$, we define $\min x$ and $\max x$ in a natural way — namely,

$$\min x = \min_{c \in C} \{x_c\} \quad \text{and} \quad \max x = \max_{c \in C} \{x_c\},$$

where the notions of minimum and maximum are defined with respect to the total order \succeq on Z .

We use the notation \vec{z} to denote the profile x for which $x_c = z$ for all $c \in C$. Such a profile is called a *uniform profile*. For every nonempty proper subset R of C , let $X_R = Z^{|R|}$ and $X_{-R} = Z^{n-|R|}$. For all $x \in X_R$ and $y \in X_{-R}$, we denote by (x, y) the profile that coincides with x for the criteria in R and coincides with y for the criteria not in R . In other words, $(x, y)_c = x_c$ for all $c \in R$ and $(x, y)_c = y_c$ for all $c \notin R$. To denote the restriction of a uniform profile \vec{z} to the criteria in R , we write \vec{z}_R .

An *evaluee* e is an ordered pair (x^e, \succeq_e) , where $x^e \in X$ and \succeq_e is a *monotone weak order* on X — that is, a weak order in which $x \succeq_e y$ whenever $x \gg y$. The first coordinate, x^e , represents the vector of scores assigned to e by the evaluation process (one for each evaluation criterion). The second coordinate, \succeq_e , represents e 's perceived ordering of the possible profiles according to their relative value or desirability. (We will call such an ordering a *profile ordering*.) Monotonicity imposes a minimal assumption of rationality on each evaluatee's profile ordering — for example, by prohibiting an evaluatee from viewing a rating that is unsatisfactory in every category as more desirable than one that is satisfactory in every category. However if x and y are two profiles with the property that $x \not\gg y$ and $y \not\gg x$ (in which case $x_i \succ y_i$ and $y_j \succ x_j$ for some $i \neq j$), then the monotonicity condition imposes no restrictions on an evaluatee's relative ordering of x and y . In this case, one can posit the existence of evaluatees e_1 and e_2 for which $x \succ_{e_1} y$ and $y \succ_{e_2} x$.

The set of all possible evaluatees is called the *evaluee space* and denoted \mathcal{E} . The set of all submultisets of \mathcal{E} is denoted $\mathcal{P}(\mathcal{E})$.¹

¹Formally, our definitions must involve multisets to allow for the possibility of multiple evaluatees having the same profile and profile ordering. However, from this point forward, we will, for convenience, use the simpler language of sets and subsets, keeping in mind that our results apply to (and sometimes require) multisets as well.

Definition 2.1. A *rank aggregation function* is a function that maps each pair (x, E) , where x is a profile and E is a subset of \mathcal{E} , to an element of Z . In other words, a rank aggregation function is a mapping $f : X \times \mathcal{P}(\mathcal{E}) \rightarrow Z$.

For any rank aggregation function f , we will use the notation $f_E(x)$ to denote $f(x, E)$. This notation captures the idea that, for a *fixed* set of evaluatees, we can view f as nothing more than a function on the profile space X . By definition, the function f is *anonymous*: it assigns the same overall score to any two evaluatees who have the same profile. However, f may take into account information about the set of evaluatees as a whole — including evaluatees' profile orderings — so that if this information changes in any way, the scores assigned by f may also change. A rank aggregation function that assigns the same scores regardless of the set of evaluatees is said to be *independent*. The next definition formalizes this idea and defines several additional desirable properties that a rank aggregation function might satisfy.

Definition 2.2. Let $f : X \times \mathcal{P}(\mathcal{E}) \rightarrow Z$ be a rank aggregation function. Then:

- (i) f is *independent*, provided that $f_{E_1}(x) = f_{E_2}(x)$ for all $E_1, E_2 \in \mathcal{P}(\mathcal{E})$ and all $x \in X$.
- (ii) f is *monotone*, provided that $x \gg y$ implies $f_E(x) \geq f_E(y)$ for all $x, y \in X$ and all $E \in \mathcal{P}(\mathcal{E})$.
- (iii) f is *inclusive*, provided that for each nonempty, proper subset R of C , there exists $E \in \mathcal{P}(\mathcal{E})$ and profiles $(x, u), (y, u)$, where $x, y \in X_R$ and $u \in X_{-R}$, such that $f_E(x, u) \neq f_E(y, u)$. (In this case, C is said to be *minimal* with respect to f .)
- (iv) f is *consistent* with respect to $E \in \mathcal{P}(\mathcal{E})$, provided that $f_E(\vec{z}) = z$ for all $z \in Z$. If f is consistent with respect to *all* $E \in \mathcal{P}(\mathcal{E})$, then f is said to be *universally consistent*, or simply *consistent*.
- (v) f is *equitable* with respect to $E \in \mathcal{P}(\mathcal{E})$, provided that $x \succ_e y$ implies $f_E(x) \geq f_E(y)$ for all $e \in E$ and all $x, y \in X$. (If, for some evaluatee $e \in E$, $x \succ_e y$ and $f_E(y) \succ f_E(x)$, then e is said to *perceive inequity*.) If f is equitable with respect to *all* $E \in \mathcal{P}(\mathcal{E})$, then f is said to be *universally equitable*, or simply *equitable*.

A rank aggregation function that is independent, monotone, inclusive, consistent, and equitable is said to be *ideal*. If f is an independent rank aggregation function, then we will denote by $f(x)$ the common value of $f_E(x)$ for all $E \in \mathcal{P}(\mathcal{E})$.

Example 2.3. Let f be the rank aggregation function defined by $f(x) = x_1$. (Note that f assigns to each profile the score received on the first criterion.)

By definition, f is independent and consistent. Furthermore, f is monotone since $x \gg y$ implies $x_1 \geq y_1$.

Since f assigns scores based solely on the first criterion of each profile, it is intuitively obvious that f is not inclusive. To prove this formally, let R be any

nonempty, proper subset of C that does not include the first criterion. Then

$$f(x, u) = (x, u)_1 = u_1 = (y, u)_1 = f(y, u)$$

for all $x, y \in X_R$ and all $u \in X_{-R}$. So C is not minimal, and f is not inclusive.

Finally, we will show that f is not universally equitable. Let E be any set of evaluatees containing an evaluatee e for which

$$(z_1, z_m, \dots, z_m) \succ_e (z_m, z_1, \dots, z_1).$$

Such an evaluatee exists since

$$(z_1, z_m, \dots, z_m) \succ\!\!\succ (z_m, z_1, \dots, z_1) \quad \text{and} \quad (z_m, z_1, \dots, z_1) \succ\!\!\succ (z_1, z_m, \dots, z_m).$$

Note, however, that since

$$f(z_m, z_1, \dots, z_1) = z_m \succ z_1 = f(z_1, z_m, \dots, z_m),$$

it follows that e perceives inequity, and f is not equitable with respect to E .

In conclusion, we have shown that f is independent, monotone, and consistent, but neither inclusive nor equitable. \square

Example 2.4. Let f be the rank aggregation function defined by $f(x) = \max x$.

By definition, f is independent and consistent. We will show that f is inclusive and monotone, and f is equitable if and only if $m = 2$.

For inclusivity, let R be any proper, nonempty subset of C . Then

$$f(\vec{z}_1) = f((\vec{z}_1)_R, (\vec{z}_1)_{-R}) = z_1,$$

but

$$f((\vec{z}_m)_R, (\vec{z}_1)_{-R}) = z_m.$$

Since f assigns different scores to two profiles that differ only on R , and R was chosen arbitrarily, it follows that C is minimal with respect to f . Thus, f is inclusive.

For monotonicity, let $x, y \in X$ such that $x \gg y$. Then $x_c \geq y_c$ for all $c \in C$, which implies

$$f(x) = \max x \geq \max y = f(y).$$

Thus, f is monotone.

For equity, note that if $m \geq 3$, any evaluatee e for which

$$\vec{z}_2 = (z_2, z_2, \dots, z_2) \succ_e (z_1, z_1, \dots, z_1, z_m)$$

will perceive inequity, since

$$f(z_1, z_1, \dots, z_1, z_m) = z_m \succ z_2 = f(\vec{z}_2).$$

(Intuitively, such an evaluatee views a profile that receives the second worst score on each criterion as favorable to one that receives the worst score on all but one

of the criteria and the best score on the remaining criterion.) This example clearly breaks down if $m = 2$. Note that for an evaluatee e to perceive inequity, it must be that, for some profiles x and y , we have $x \succ_e y$ and $f(y) \succ f(x)$. But when $m = 2$, $f(y) \succ f(x)$ only if $f(y) = z_2$ and $f(x) = z_1$. This, however, implies $x = \bar{z}_1$, a contradiction to the monotonicity of \succ_e in light of the assumption that $x \succ_e y$.

To summarize, we have shown that f is independent, monotone, consistent, and inclusive. In the case that $m = 2$ (and only in this case), f is also equitable, and thus ideal. \square

Example 2.5. For a more concrete example, consider an evaluation process with three evaluation criteria ($n = 3$) and three possible scores ($m = 3$) for each criterion:

outstanding \succ satisfactory \succ unsatisfactory.

Define a rank aggregation function f as follows (abbreviating each score by its first letter):

$$f_E(x) = \begin{cases} O & \text{if } x \succ_e x^e \text{ for all } e \in E \text{ with } x^e \neq x, \\ U & \text{if } x^e \succ_e x \text{ for all } e \in E \text{ with } x^e \neq x, \\ S & \text{otherwise.} \end{cases}$$

Suppose Sally is being evaluated using f . Then the only way for Sally to be rated as outstanding is for every other evaluatee whose profile is different from hers to view Sally's profile as the more favorable one. In essence, every evaluatee must envy Sally's profile. Likewise, for Sally to be rated as unsatisfactory, every evaluatee with a different profile than Sally's must view their own profile as favorable to hers.

Now consider the set E consisting of three evaluatees— a , b , and c —for whom the following conditions hold:

- $x^a = (S, S, O)$ and $(S, S, O) \succ_a (S, O, S) \succ_a (O, S, S)$;
- $x^b = (S, O, S)$ and $(S, O, S) \succ_b (O, S, S) \succ_b (S, S, O)$;
- $x^c = (O, S, S)$ and $(O, S, S) \succ_c (S, S, O) \succ_c (S, O, S)$.

Note that, in this case, $f_E(S, S, O) = f_E(S, O, S) = f_E(O, S, S) = U$, since every evaluatee favors their own profile over those of each of the other evaluatees. (This is not a terribly surprising outcome, since evaluatees tend to exhibit self-serving biases.)

Consider, however, another set E' of evaluatees who, one might argue, are less tainted by their individual biases:

- $x^u = (S, S, O)$ and $(O, S, S) \succ_u (S, O, S) \succ_u (S, S, O)$;
- $x^v = (S, O, S)$ and $(O, S, S) \succ_v (S, O, S) \succ_v (S, S, O)$;
- $x^w = (O, S, S)$ and $(O, S, S) \succ_w (S, S, O) \succ_w (S, O, S)$.

It is easy to verify that $f_{E'}(S, S, O) = U$, $f_{E'}(S, O, S) = S$, and $f_{E'}(O, S, S) = O$. In other words, changing the set of evaluatees changes the overall scores assigned by f to the profiles (S, O, S) and (O, S, S) . Thus, f is *not* independent. This means that an evaluator who wished to use f would need to have some mechanism for ascertaining the evaluatees' profile orderings. Moreover, as the example illustrates, misrepresentation of profile orderings — and even the influence of unintentional biases — could have a significant impact on the outcome of the evaluation. This phenomenon is analogous to the problem of insincere or strategic voting in social choice theory.

It is also worth noting that, while f may be equitable with respect to E (in this example, the profile orderings of a , b , and c are not specified completely enough to make a definitive conclusion), f is certainly *not* equitable with respect to E' ; in particular, w perceives inequity since $(S, S, O) \succ_w (S, O, S)$ and

$$f_{E'}(S, O, S) = S \succ U = f_{E'}(S, S, O).$$

We leave it as an exercise to the reader to verify that f is monotone and inclusive, but not consistent. \square

In the above examples, we considered rank aggregation functions that were monotone, but not necessarily equitable. In fact, because of the monotonicity requirement for profile orderings, the properties of monotonicity and equity (for rank aggregation functions) are closely related. This relationship can be made clear by examining the contrapositive of each property:

- *Monotonicity*: For all $x, y \in X$ and all $E \in \mathcal{P}(\mathcal{E})$, $f_E(y) \succ f_E(x)$ implies $x \not\gg y$.
- *Equity*: For all $x, y \in X$, all $E \in \mathcal{P}(\mathcal{E})$, and all $e \in E$, $f_E(y) \succ f_E(x)$ implies $y \succeq_e x$.

The difference between monotonicity and equity therefore amounts to the difference between the conditions of $x \not\gg y$ and $y \succeq_e x$. If $y \gg x$, then it is certainly the case that $y \succeq_e x$ for all e . But, as we have seen, $x \not\gg y$ does not necessarily imply $y \gg x$. Likewise, it is also possible for $x \gg y$ and $y \succeq_e x$, which happens if and only if e is indifferent between x and y .

The arguments for monotonicity and equity are often similar, and so we will sometimes invoke the following lemma to prove both properties simultaneously:

Lemma 2.6. *Let f be a rank aggregation function, and suppose that for all $x, y \in X$, $f_E(y) \succ f_E(x)$ implies both (1) $x \not\gg y$, and (2) $y \succeq_e x$ for every evaluatee e . Then f is both monotone and (universally) equitable.*

Finally, it is important to recognize the difference between universality — particularly, as it applies to the properties of consistency and equity — and independence. Independence requires the scores assigned to profiles to depend only on the profiles themselves, and not on any properties of the underlying set of evaluatees. However,

a rank aggregation function need not be independent to be universally consistent or equitable. The latter properties do not require the assigned scores to be invariant with respect to changes in the set of evaluatees. Rather, these properties require only that the function be consistent/equitable with respect to *each* set of evaluatees. Likewise, an independent rank aggregation function may be equitable with respect to some sets of evaluatees, but not others. On the other hand, if an independent rank aggregation function is consistent with respect to any individual set of evaluatees, then it will be universally consistent.

3. Existence of ideal rank aggregation functions

In the previous section, we saw an example of a rank aggregation function that was independent, monotone, consistent, and inclusive, but equitable (and thus ideal) only when $m = 2$ (Example 2.4). The rank aggregation function in Example 2.5 fared even worse: it was monotone and inclusive, but not independent, consistent, or equitable. This, of course, begs the question: What would an ideal rank aggregation function look like for $m \geq 3$? Does such a function even exist? The next theorem provides a partial answer to this question.

Theorem 3.1. *Let $m = 3$, and define the rank aggregation function f as*

$$f(x) = \begin{cases} z_1 & \text{if } x = \vec{z}_1, \\ z_3 & \text{if } x = \vec{z}_3, \\ z_2 & \text{otherwise.} \end{cases}$$

Then f is an ideal rank aggregation function.

Proof. By definition, f is independent and consistent. The argument for inclusivity is similar to that in Example 2.4.

For monotonicity and equity, we will use Lemma 2.6. So suppose $f(y) \succ f(x)$ for some $x, y \in X$. Then either $f(y) = z_3$ or $f(x) = z_1$. If the former, then $y = \vec{z}_3$ and $x \neq \vec{z}_3$. If the latter, then $x = \vec{z}_1$ and $y \neq \vec{z}_1$. In either case, $y \gg x$, which implies, by the monotonicity of profile orderings, that $y \succeq_e x$ for each evaluatee e . Furthermore, $x \not\gg y$, and so f is both monotone and equitable. \square

Unfortunately, the function from Theorem 3.1 cannot be extended to cases where $m \geq 4$. In fact, for $m \geq 4$, the properties of consistency and equity are incompatible.

Theorem 3.2. *For $m \geq 4$, there does not exist a rank aggregation function that is consistent and equitable.*

Proof. Let $m \geq 4$, and assume to the contrary that f is a consistent and equitable rank aggregation function. Let $u = (z_1, z_m, z_m, \dots, z_m)$, and let E contain evaluatees e_1 and e_2 for which

$$\vec{z}_2 \succ_{e_1} u \succ_{e_2} \vec{z}_3.$$

Since f is equitable and consistent,

$$z_2 = f_E(\vec{z}_2) \succeq f_E(u) \succeq f_E(\vec{z}_3) = z_3.$$

This, however, is a contradiction to the fact that $z_3 \succ z_2$ by definition. \square

Theorem 3.2 shows that, for $m \geq 4$, it is impossible for a consistent rank aggregation function to be equitable with respect to *every* possible set of evaluatees. However, it does not rule out the possibility of finding a consistent rank aggregation function that is equitable with respect to *some* sets of evaluatees. If one also desires independence, then this distinction is moot. (If f is independent, then since $f_E(\vec{z}_2) \succeq f_E(\vec{z}_3)$ for some set E of evaluatees, the same ordering would hold for *every* set of evaluatees, leading again to a contradiction.) If we are willing to sacrifice independence, then we may have more options. We will explore this possibility further in [Section 5](#).

4. Cycles and uniqueness of ideal rank aggregation functions

In the proof of [Theorem 3.2](#), two evaluatees' profile orderings were combined in a way that forced a contradiction under the assumptions of both consistency and equity. This idea can be generalized as follows.

Definition 4.1. Let $E \in \mathcal{P}(\mathcal{E})$. Suppose there exist $e_1, \dots, e_k \in E$ and $x^1, \dots, x^k \in X$ such that

$$x^1 \succ_{e_1} x^2 \succ_{e_2} \cdots \succ_{e_{k-1}} x^k \succ_{e_k} x^1.$$

Then the sequence x^1, \dots, x^k, x^1 is said to be a *strong k -cycle* with respect to E .

The following theorem is immediate:

Theorem 4.2. Let $E \in \mathcal{P}(\mathcal{E})$, and let

$$T_E = \{x \in X : x \text{ belongs to some strong cycle with respect to } E\}.$$

Let the relation \sim_E on T_E be defined by $x \sim_E y$ if and only if x and y belong to a common strong cycle. Then \sim_E is an equivalence relation on T_E .

The next theorem can be viewed as a generalization of [Theorem 3.2](#).

Theorem 4.3. Let $E \in \mathcal{P}(\mathcal{E})$, and let f be a rank aggregation function that is equitable with respect to E . If $x \sim_E y$ for some $x, y \in X$, then $f_E(x) = f_E(y)$.

Proof. Suppose f is equitable, and let $x \sim_E y$ for some $x, y \in X$. Then x and y belong to a common strong k -cycle with respect to E — that is,

$$x \succ_{e_1} \cdots \succ_{e_{j-1}} y \succ_{e_j} \cdots \succ_{e_k} x$$

for some $e_1, \dots, e_k \in E$. But then equity requires that $f_E(x) \succeq f_E(y) \succeq f_E(x)$, which implies $f_E(x) = f_E(y)$. \square

Two important corollaries follow.

Corollary 4.4. *Let $E \in \mathcal{P}(\mathcal{E})$. If $\vec{z}_i \sim_E \vec{z}_j$ for some $z_i, z_j \in Z$ with $i \neq j$, then there does not exist a rank aggregation function that is both consistent and equitable with respect to E .*

Proof. Suppose f is equitable with respect to E , and suppose also that $\vec{z}_i \sim_E \vec{z}_j$ for some $z_i, z_j \in Z$ with $i \neq j$. By [Theorem 4.3](#), $f_E(\vec{z}_i) = f_E(\vec{z}_j)$. Since $i \neq j$, it is therefore impossible for f to be consistent with respect to E . \square

Corollary 4.5. *The function defined in [Theorem 3.1](#) is the unique ideal rank aggregation function for $m = 3$.*

Proof. Let $m = 3$, and let f be an ideal rank aggregation function. Since f is consistent, $f(\vec{z}_1) = z_1$, $f(\vec{z}_2) = z_2$, and $f(\vec{z}_3) = z_3$. Choose $x \in X$ such that $x \neq \vec{z}_1, \vec{z}_2, \vec{z}_3$. We will show that there exists $E \in \mathcal{P}(\mathcal{E})$ such that $x \sim_E \vec{z}_2$, which will imply (by [Theorem 4.3](#)) that $f(x) = f(\vec{z}_2) = z_2$. Consider three cases.

Case 1: $x \not\gg \vec{z}_2$ and $\vec{z}_2 \not\gg x$. In this case, there exist evaluatees e_1 and e_2 for which $\vec{z}_2 \succ_{e_1} x \succ_{e_2} \vec{z}_2$. Let E contain both e_1 and e_2 . Then $x \sim_E \vec{z}_2$, as desired.

Case 2: $x \gg \vec{z}_2$. In this case, $x = ((\vec{z}_3)_R, (\vec{z}_2)_{-R})$ for some nonempty $R \subset C$. Let $y = ((\vec{z}_1)_R, (\vec{z}_3)_{-R})$. Note that (1) $\vec{z}_2 \not\gg y$ and $y \not\gg \vec{z}_2$, and (2) $x \not\gg y$ and $y \not\gg x$. By assumption, $x \gg \vec{z}_2$. Thus, there exist evaluatees e_1, e_2 , and e_3 such that

$$\vec{z}_2 \succ_{e_1} y \succ_{e_2} x \succ_{e_3} \vec{z}_2.$$

It follows that $x \sim_E \vec{z}_2$ for any $E \in \mathcal{P}(\mathcal{E})$ containing e_1, e_2 , and e_3 .

Case 3: $\vec{z}_2 \gg x$. In this case, $x = ((\vec{z}_1)_R, (\vec{z}_2)_{-R})$ for some nonempty $R \subset C$. Let $y = ((\vec{z}_3)_R, (\vec{z}_1)_{-R})$. By a similar argument as in Case 2, there exist evaluatees e_1, e_2 , and e_3 such that

$$\vec{z}_2 \succ_{e_1} x \succ_{e_2} y \succ_{e_3} \vec{z}_2.$$

Thus, $x \sim_E \vec{z}_2$ for any $E \in \mathcal{P}(\mathcal{E})$ containing e_1, e_2 , and e_3 .

In each case, there exists $E \in \mathcal{P}(\mathcal{E})$ such that $x \sim_E \vec{z}_2$; hence $f_E(x) = f_E(\vec{z}_2) = z_2$. Since f is independent, it follows that $f(x) = z_2$ for all $x \neq \vec{z}_1, \vec{z}_2, \vec{z}_3$. Thus, we have shown that f is identical to the function from [Theorem 3.1](#). \square

5. Forfeiting independence

As we saw in [Section 3](#), it is impossible to define a consistent rank aggregation function for $m \geq 4$ that is equitable with respect to *all* possible sets of evaluatees. In fact, [Corollary 4.4](#) tells us that consistency and equity are compatible only when the set E of evaluatees is such that no two distinct uniform profiles belong to a common strong cycle. This necessary condition turns out to be sufficient as well, provided that we are willing to sacrifice independence.

In order to proceed, we first need the following definition:

Definition 5.1. Let $E \in \mathcal{P}(\mathcal{E})$, and let $x, y \in X$. Suppose there exist $e_1, \dots, e_k \in E$ and $x^1, \dots, x^{k-1} \in X$ such that

$$x \succ_{e_1} x^1 \succ_{e_2} \dots \succ_{e_{k-1}} x^{k-1} \succ_{e_k} y.$$

Then x is said to *chain-dominate* y with respect to E , denoted $x \rightarrow_E y$.

Note that, for any $x, y \in X$, we have $x \sim_E y$ if and only if $x \rightarrow_E y$ and $y \rightarrow_E x$.

Theorem 5.2. Let f be the rank aggregation function defined as follows:

$$f_E(x) = \begin{cases} z_1 & \text{if } x = \vec{z}_1, \\ \vdots & \\ z_i & \text{if } f_E(x) \text{ is not defined above and either } \max x = z_i \\ & \text{or there exists } w \in X \text{ such that } \max w = z_i \text{ and } w \rightarrow_E x, \\ \vdots & \\ z_m & \text{otherwise.} \end{cases}$$

Then f is monotone, inclusive, and equitable. Moreover, f is consistent with respect to any $E \in \mathcal{P}(\mathcal{E})$ for which $\vec{z}_i \sim_E \vec{z}_j$ only if $i = j$.

Proof. For monotonicity, suppose $x \gg y$ for some $x, y \in X$. We must show that $f_E(x) \geq f_E(y)$ for all $E \in \mathcal{P}(\mathcal{E})$. Suppose $f_E(x) = z_i$. If $i = 1$, then $x = y = \vec{z}_1$, and so $f_E(x) = f_E(y) = z_1$. If $i = m$, then $f_E(x) = z_m \geq f_E(y)$, as desired.

Now suppose $1 < i < m$. Then either (1) $\max x = z_i$, or (2) there exists $w \in X$ with $\max w = z_i$ such that $w \rightarrow_E x$. If $\max x = z_i$, then $\max y \leq z_i$, and so $f_E(y) \leq z_i = f_E(x)$, as desired. Suppose, on the other hand, that there exists $w \in X$ with $\max w = z_i$ such that $w \rightarrow_E x$. Since $w \rightarrow_E x$, there exist $e_1, \dots, e_k \in E$ and $x^1, \dots, x^{k-1} \in X$ such that

$$w \succ_{e_1} x^1 \succ_{e_2} \dots \succ_{e_{k-1}} x^{k-1} \succ_{e_k} x.$$

But $x \gg y$, and so $x \succeq_{e_k} y$, which implies $x^{k-1} \succ_{e_k} y$. Thus $w \rightarrow_E y$, and so $f_E(y) \leq z_i = f_E(x)$. In each case, $f_E(x) \geq f_E(y)$. It follows that f is monotone.

For inclusivity, let R be any proper, nonempty subset of C . Then

$$f_E(\vec{z}_1) = f_E((\vec{z}_1)_R, (\vec{z}_1)_{-R}) = z_1,$$

but

$$f_E((\vec{z}_m)_R, (\vec{z}_1)_{-R}) \neq z_1.$$

Since f assigns different scores to two profiles that differ only on R , and R was chosen arbitrarily, it follows that C is minimal with respect to f . Thus, f is inclusive.

For equity, let $E \in \mathcal{P}(\mathcal{E})$, and suppose $x \succ_e y$ for some $x, y \in X$ and some $e \in E$. We must show that $f_E(x) \geq f_E(y)$. Let $f_E(x) = z_i$. If $i = 1$, then $x = \vec{z}_1$, a contradiction to the fact that $x \succ_e y$. Therefore, $i > 1$. If $i = m$, then $f_E(x) = z_m \geq f_E(y)$,

as desired. Now suppose $1 < i < m$. By the definition of f , either $\max x = z_i$ or there exists $w \in X$ with $\max w = z_i$ such that $w \rightarrow_E x$. If $\max x = z_i$, then the fact that $x \rightarrow_E y$ (since $x \succ_e y$) implies, by the definition of f , that $f_E(y) \preceq z_i$. If, on the other hand, there exists $w \in X$ with $\max w = z_i$ such that $w \rightarrow_E x$, then it must be that $w \rightarrow_E y$ as well (since $x \succ_e y$), and so $f_E(y) \preceq z_i$. In either case, $f_E(x) \succeq f_E(y)$. It follows that e cannot perceive inequity. Thus, f is equitable with respect to E .

For consistency, let $E \in \mathcal{P}(\mathcal{E})$ such that $\vec{z}_i \sim_E \vec{z}_j$ only if $i = j$. It suffices to show that, for each $1 < i \leq m$, there does not exist $w \in X$ such that $w \rightarrow_E \vec{z}_i$ and $\max w = z_j$ for some $j < i$. Suppose, to the contrary, that such a w exists. Since $w \rightarrow_E \vec{z}_i$, there exist $e_1, \dots, e_k \in E$ and $x^1, \dots, x^{k-1} \in X$ such that

$$w \succ_{e_1} x^1 \succ_{e_2} \dots \succ_{e_{k-1}} x^{k-1} \succ_{e_k} \vec{z}_i.$$

Moreover, since $\max w = z_j$ and $j < i$, it follows that $\vec{z}_i \gg \vec{z}_j \gg w$. By the monotonicity of \succeq_{e_1} , we have $\vec{z}_i \succeq_{e_1} \vec{z}_j \succeq_{e_1} w$. Since $w \succ_{e_1} x^1$, this implies that $\vec{z}_i \succ_{e_1} x^1$ and $\vec{z}_j \succ_{e_1} x^1$. By the monotonicity of \succeq_{e_k} , we have $\vec{z}_i \succeq_{e_k} \vec{z}_j$, and so $x^{k-1} \succ_{e_k} \vec{z}_j$. It follows that

$$\vec{z}_i \succ_{e_1} x^1 \succ_{e_2} \dots \succ_{e_{k-1}} x^{k-1} \succ_{e_k} \vec{z}_j \succ_{e_1} x^1 \succ_{e_2} \dots \succ_{e_{k-1}} x^{k-1} \succ_{e_k} \vec{z}_i.$$

But then $\vec{z}_i \sim_E \vec{z}_j$ with $i \neq j$, a contradiction. Therefore, it must be the case that $f_E(\vec{z}_i) = z_i$ for all i , and f is consistent. \square

6. Manifest inequity

In our investigations up to this point, we have not made a distinction between the *potential* for inequity and the actual *manifestation* of inequity in the scores assigned by a given rank aggregation function. The former involves a systemic or structural concern — namely, that a rank aggregation function *may* lead to ratings that are perceived by some to be inequitable, regardless of whether any specific evaluatee receives one of the profiles involved in these potential inequities. However, perceived inequity involving profiles that are actually assigned to evaluatees — what we will call *manifest* inequity — is especially problematic. In this section, we will consider the more modest goal of avoiding manifest inequity, defined formally below.

Definition 6.1. Let f be a rank aggregation function, and let $E \in \mathcal{P}(\mathcal{E})$. Suppose there exist $e_1, e_2 \in E$ such that $x^{e_1} \succ_{e_1} x^{e_2}$ and $f_E(x^{e_2}) \succ f_E(x^{e_1})$. Then

- x^{e_1} and x^{e_2} are called *manifest profiles* with respect to E ; and
- e_1 is said to perceive *manifest inequity*.

A rank aggregation function for which no $e \in E$ perceives manifest inequity is said to be *weakly equitable* with respect to E .

Clearly, one way to avoid manifest inequity is to limit the profiles that are actually assigned to evaluatees. In situations in which a single evaluator both assigns profiles to evaluatees and chooses the rank aggregation function, this solution is both simple and practical. In fact, we will show that it is possible to define a rank aggregation function that is independent, monotone, consistent, inclusive, and weakly equitable with respect to any set of evaluatees that satisfies the pairwise dominance condition defined below.

Definition 6.2. A set E of evaluatees is said to be *pairwise dominant* if for all $e_1, e_2 \in E$, either $x^{e_1} \gg x^{e_2}$ or $x^{e_2} \gg x^{e_1}$. Likewise, a set S of profiles is said to be *pairwise dominant* if for all $x, y \in S$, either $x \gg y$ or $y \gg x$.

Theorem 6.3. Let f be the rank aggregation function defined by $f(x) = \max x$. Then f is independent, consistent, inclusive, monotone, and weakly equitable with respect to any pairwise dominant set E of evaluatees.

Proof. By definition, f is independent and consistent. The proof of inclusivity and monotonicity is given in [Example 2.4](#).

Now let E be any pairwise dominant set of evaluatees. We must show that f is weakly equitable with respect to E . Suppose $x^{e_1} \succ_{e_1} x^{e_2}$ for some $e_1, e_2 \in E$. Since E is assumed to be pairwise dominant, either $x^{e_1} \gg x^{e_2}$ or $x^{e_2} \gg x^{e_1}$. If $x^{e_2} \gg x^{e_1}$, then the monotonicity of \succeq_{e_1} implies $x^{e_2} \succeq_{e_1} x^{e_1}$, a contradiction. Therefore, $x^{e_1} \gg x^{e_2}$, and so

$$f(x^{e_1}) = \max x^{e_1} \succeq \max x^{e_2} = f(x^{e_2}),$$

as desired. □

[Theorem 6.3](#) raises two interesting combinatorial questions. First, what is the cardinality of a maximal pairwise dominant set of profiles? Second, how many such sets exist? The next theorem provides answers to these questions.

Theorem 6.4. Let S be a maximal pairwise dominant set of profiles — i.e., a set of profiles S that is pairwise dominant and is not a proper subset of any other pairwise dominant set. Then

$$|S| = n(m - 1) + 1.$$

Moreover, the number of distinct pairwise dominant sets of profiles is equal to

$$\frac{(n(m - 1))!}{(m - 1)!^n}.$$

Proof. First, note that (X, \gg) is a poset, and x covers y in (X, \gg) if and only if there exists $c \in C$ such that (1) $x_i = y_i$ for all $i \neq c$, and (2) $x_c = z_k$ and $y_c = z_{k-1}$ for some $1 < k \leq m$. The cardinality of any maximal pairwise dominant set is the same as the length of a chain of maximum length in (X, \gg) . Such a chain must have maximum element \vec{z}_m , with each subsequent element formed by changing the

n	m	m^n	$n(m-1)+1$	$\frac{(n(m-1))!}{(m-1)!^n}$
3	3	27	7	90
3	4	64	10	1680
3	5	125	13	34650
4	3	81	9	2520
4	4	256	13	369600
4	5	625	17	63063000

Table 1. Combinatorial results for small m, n .

preceding element's score from z_k to z_{k-1} on exactly one criterion. Since there are n criteria, and the score on each criterion may be decreased (from z_k to z_{k-1}) exactly $m - 1$ times before reaching z_1 , it follows that a chain in (X, \gg) can have length at most $n(m - 1) + 1$.

Each maximal chain (and, hence, each maximal pairwise dominant set) is uniquely determined by the order in which the criteria are decreased. Thus, each maximal pairwise dominant set corresponds to a sequence of $n(m - 1)$ elements of C , where each of the n elements of C appears $m - 1$ times. The standard formula for counting permutations with repetition thus implies that there are

$$\frac{(n(m - 1))!}{(m - 1)!^n}$$

such sequences, as desired. \square

Table 1 illustrates the results of Theorem 6.4 for some small values of m and n . While it is clear that pairwise dominant sets of profiles are small in comparison to the total number of profiles (m^n), it is also the case that there are many such sets to choose from. Therefore, an evaluator who wishes to take advantage of Theorem 6.3 must accept some fairly severe restrictions, but can satisfy these restrictions in a number of different ways.

7. Summary and conclusions

Rank disequilibrium can be a significant factor in social and intergroup conflict. In this article, we investigated rank disequilibrium in the context of multiple-criteria evaluation, using an axiomatic approach to show that, in general, it is impossible to define a function that aggregates scores on individual criteria while satisfying a relatively small set of desirable properties. In particular, we showed the notions of consistency and equity are generally incompatible. It seems perfectly reasonable to expect evaluatees who receive the same score on every criterion to also receive that

score as an overall rating. However, this apparently innocuous requirement opens the door for evaluatees to perceive inequity.

The problems illuminated by our analysis are not insoluble. When evaluators are limited to only three possible scores (say, outstanding, satisfactory, and unsatisfactory), an ideal rank aggregation function can be found. However, the only such function assigns a overall rating of satisfactory to almost all profiles, rating evaluatees as outstanding or unsatisfactory only when they receive these respective scores on *every* criterion.

Another potential solution is to allow the evaluator to use information about evaluatees' individual profile orderings in order to assign scores that minimize the potential for perceived inequity. An evaluator who implements such a system is likely to be motivated more by political considerations than an actual concern for equity. Indeed, doing so requires a great deal of effort (to ascertain reliable information about evaluatees' relative orderings of the various profiles) without an absolute guarantee that the ideals of consistency and equity will be achieved.

A more practical solution involves simply limiting the profiles assigned to evaluatees to ensure that, for any pair of profiles assigned, one profile is viewed as more desirable by *all* rational evaluatees (that is, all evaluatees whose profile orderings meet the modest condition of monotonicity.) An evaluator need not have foreknowledge of the evaluatees' views in order to adopt this strategy, but must be willing to possibly rate evaluatees insincerely in order to avoid assigning profiles that fall outside the allowed set. She must also settle for the weaker goal of avoiding *manifest* inequity, rather than *all* perceived inequity. Broader systemic or structural concerns may still persist, but the most glaring perceptions of inequity will be eliminated.

Our investigations rest on several assumptions that could be relaxed in future work. For example, we have required the same rating scale (i.e., the same set of scores) to be used for each criterion, as well as for the overall evaluation. We have also assumed the set of scores to be discrete. In other multiple-criteria decision contexts (such as voting on referenda; see [Bradley et al. 2005], for example), the distinction between discrete and continuous alternative sets has been shown to be significant.

Our model does not incorporate any assumptions about the relative importance of the evaluation criteria, nor does it address potential interdependencies among criteria. (In early critiques of research on rank disequilibrium, Doreian and Stockman [1969] and Hartman [1974] raise related concerns.) Given the importance of separability in economics and social choice theory [Bradley et al. 2005; Gorman 1968; Hodge and Schwallier 2006], these areas would seem to warrant further investigation.

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