Self versus Assessor Ratings and their Classification in Assessment Centres: Profiling the Self-Rater

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Self versus assessor behavioural ratings from 214 participants were investigated along with psychometric measures in a development centre. Using cluster analysis, it was shown that sub-groups of self-raters could be established without the need to use commonly criticized difference scores. Four clusters of self-rater were identified in this analysis, a result that builds on popular theory (Atwater & Yammarino, 1992). Self-ratings by cluster were plotted with psychometric measures and assessor ratings. Those who self-rated higher tended to score relatively high on personality constructs related to social dominance. However, sub-clusters of self-rater were neither differentiated by cognitive ability scores nor by assessor ratings. Strength of difference between self and assessor ratings was contingent on cluster membership. Cluster analysis is suggested as a useful technique for understanding self-rater behaviour and as a guide to identifying how feedback should be tailored to potential recipients.

A body of literature to date has focused on self versus assessor discrepancies on performance dimensions. Schmitt, Ford, and Stults (1986) studied eight performance dimensions prior and subsequent to the assessment event. They found small correlations between performance on each dimension and assessor judgments (< .20). Similar results were reported by Clapham (1998), who also investigated the moderating effects of cognitive abilities and gender. Again, and even with external variables controlled, small correlations were observed (< .33). Note, however, that an earlier study by Byham (1971) found higher correlations between assessor and post AC participant performance (around .60).

Participants who overestimate (or, at least, self-rate higher than assessors) their performance may, potentially, feel discouraged by assessor ratings. It appears to be the case that overestimation is often the norm with regard to self-perceptions. Comparisons of means and standard deviations from several studies generally appear to indicate higher self-ratings (e.g., Clapham, 1998; Halman & Fletcher, 2000). Maciejczyk (1992). Randall, et al. (2000, p. 445) comment that in cognitive psychology, a “general tendency towards overconfidence” has been observed. This may generalize to ACs to the extent that a corridor effect may occur. Corridor effects constitute a form of range restriction, whereby participants have little idea about their performance and score self-perceptions...
within a narrow corridor around some mean value. Typically, it would seem, these self-rated mean values are higher than those of assessors.

From an alternative perspective, it may be the studies of correlations and mean comparisons between overall self-ratings, assessor ratings, and psychological variables hide important patterns due to corridor effects. As such, classifying sub-groups of self-raters in a meaningful way could assist in revealing complex patterns that might otherwise remain latent (Halman & Fletcher, 2000). In acknowledgement of these range restriction issues, Atwater and Yammarino (1992) suggested categorization as a means of circumventing reliability problems associated with self minus other difference scores (see Johns, 1981). Atwater and Yammarino described over-estimators as those with an inflated view of their performance, those who are in agreement as holding accurate schema with regard to their performance, and under-estimators as those who attenuate their self-ratings. Difference scores were, however, used in the categorization of individuals into these three sub-groups. While difference scores were not used in statistical analyses, it seems less than satisfactory to use an unreliable strategy to categorize individuals. However, this approach has been employed in studies of self versus other ratings in ACs (e.g., Halman & Fletcher, 2000; Randall, et al., 2000). Cluster analysis is a technique, not known to have been previously used in this domain, which allows sub-clusters of individuals to be identified in a reliable manner (see Hair, Anderson, Tatham, & Black, 1998).

Two points are of note here. Firstly, studies of self versus assessor ratings on task-specific assessment centres would add to the literature by presenting an alternative view of the data. One such study was identified and is discussed below. Secondly, given the exercise effects often observed in ACs, it seems implicit that assessors often perceive participant performance to be specific to an exercise or task-specific (see Lowry, 1997). Likewise, the question arises as to whether participants also perceive their performance as being task-specific. This issue is of importance, particularly with regard to DC performance on which feedback is a fundamental component of the process. Providing an appropriate and cognitively intuitive context for this feedback may facilitate the delivery of performance information to participants. Clapham (1998) found that in an AC containing 16 dimensions, both participants and assessors rated dimensions according to an underlying structure that suggested three principal components. It was unclear in Clapham whether these patterns were indicative of exercise effects. However, one could conclude that a small number of components were found relative to desired dimensions. Further insight into the factor patterns observed in self-ratings would inform on how participants perceive their AC experience.

A single study was found comparing self versus assessor ratings in a task-specific AC (Franks, Ferguson, Rolls, & Henderson, 1998). Franks, et al. used difference scores to analyze their data. Thus, while circumventing some of the construct validity issues associated with ACs, other analytical issues were evident. Reliability issues aside, Franks, et al. (1998) found that difference scores were statistically significant in terms of their capacity to distinguish successful versus unsuccessful AC candidates. However, raw self-assessment scores were non-significant predictors. As with other studies incorporating dimension-specific ACs, Franks, et al. found a general over-estimation in performance across exercises in nine out of ten indexes reported for successful and non-successful candidates across five exercises.

The research presented above suggests that social dominance plays a role in the way behaviour is expressed in group situations (Islam & Zyphur, 2005). Social dominance may also play a role in terms of an individual’s self-perceptions of performance on group exercises. Moreover, while cognitive abilities have been researched in terms of their general moderating effects (Clapham, 1998), those who are more cognitively able may also hold a heightened awareness of their own abilities relative to the views of assessors. Investigation into these potential patterns presents a secondary aim for this paper, leading to the following hypotheses:

**Hypothesis 1**: Those who rate themselves incongruently high, relative to assessors, on their assessment centre performance will also score higher on dominance-oriented dimensions. 

**Hypothesis 2**: Participants who score higher on cognitive abilities are more likely to self-rate in a manner that is congruent with assessor ratings.

**Method**

**Participants**

A total of 229 managers from a major postal organisation participated in a DC aimed at developing a number of managerial competencies. Of the total number of participants, 214 were retained as usable for data analysis, the remainder being too incomplete for inclusion. The mean age of the participants was 45.53 (SD = 10.33) and a relatively even split between genders was observed (males, around 54% and females, around 46%). Around 78% of the sample were Caucasian and just over half of the participants had completed high school (around 53%) or had a trade certificate or degree (around 25%).

**Assessors**

The assessor panel consisted of a mixture of senior managers and human resource...
consultants. The panel comprised 23 assessors, who independently completed a 10-item behavioural checklist (see below) for each exercise. Assessors were trained using a frame-of-reference procedure (see Schleicher, Day, Mayes, & Riggio, 2002 for details on this approach), which involved mock administrations of exercises and reporting and discussing ratings that were given. This approach is often presented as a form of standard setting and is recommended for use with ACs (Lievens, 1998).

**Measures**

**Competency/Job Analysis.** The DC under scrutiny was developed using the course of action set out by Lowry (1997). This process began by using a competency model that had been specified for the company under study. This model was based on the Lominger competency framework (see Lombardo & Eichinger, 2002; 2003). Using this framework as an initial guide, subject matter expert interviews were carried out incorporating the views of human resource managers, area managers, and line managers. The aim was to identify tasks and psychological variables that were relevant to the job.

**Assessment Exercises.** Four exercises were developed in total and are described below. Each exercise contained a 10-item behavioural checklist, each of which held acceptable internal consistency (see Nunnally & Bernstein, 1994) as indicated by Cronbach’s alpha (alpha for exercise 1 = .92; 2 = .92; 3 = .93; and 4 = .94).

**Exercise 1: Managing new staff.** The first component of this exercise involved a discussion in small groups around the issue of managing new front-line staff members. The second component of the exercise involved an individual presentation back to the assessors on the important factors to consider when managing new staff in the focal organisation.

**Exercise 2: Selecting new staff.** The format of this exercise was similar to that above and included a discussion and individual presentation. The focus here, however, was on the factors that the participants would consider important for the selection of new staff specializing in insurance and lending.

**Exercise 3: Photo exercise.** The photo exercise was a group discussion in which a series of photos were shown to participants displaying the interior and exterior of postal outlet stores. Access problems and issues around aesthetics were purposely staged in the photos to provide material for debate.

**Exercise 4: Coaching exercise.** The coaching exercise was a role play in which candidates were requested to plan a performance coaching meeting with an employee. Participants were then asked to role play a coaching session in which performance plans for the next six months were agreed.

**Self-Ratings.** After each exercise, participants were given a three-item self-rating measure including the items “how satisfied were you with your performance on this exercise”, “how satisfied do you think other members of the group were with your performance on this exercise”, and “how satisfied do you think the observers were with your performance on this exercise”. These items were scored on a scale ranging from 1 (certainly dissatisfied) to 6 (certainly satisfied). Ratings were purposefully given only after a given exercise to allow the participant’s perspective on their performance specific to an exercise, rather than general self-perceptions of performance. The self-rating scales in this study held acceptable reliability within exercises (Cronbach’s alpha for self-ratings on exercise 1 = .91; 2 = .87; 3 = .84; and 4 = .94). Note that the intention in this study was to cluster participants in terms of their perceptions of behavioural performance on the exercises described above.

**Cognitive Abilities.** The Graduate Reasoning Test battery (GRT1) published by Psytech International Limited was used to assess verbal, numerical, and abstract reasoning in the DC. Reliability for these tests was deemed acceptable from studies that appear in the technical manual for the test battery (Psytech International, 1991) (test re-test reliability coefficients were at .79, .78, and .74 respectively). The information presented in Psytech (1991) also provided evidence of the construct and predictive validity of these measures. Note that, due to practical considerations, it was not possible to determine sample-specific test-retest reliability.

**Personality Measures.** Subscales from the 15FQ+ personality questionnaire (see the technical manual from Psytech International, 2002 for information on validity) were used to assess aspects theoretically related to dominance in group situations. Three scales were used in total, namely socially bold, intellectance, and dominant, all of which are described in the manual as holding acceptable reliability (Cronbach’s alpha > .70). Alphas specific to this study were as follows: Socially bold = .72; intellectance = .70; and dominant = .68. The latter coefficient possibly reflects the broad nature of the construct in its coverage of assertive, competitive, aggressive, and forceful tendencies. Socially bold individuals are described as being confident in social settings. Intellectance relates to the confidence one has in their intellectual abilities. The technical manual states that such individuals may be “quick to take advantage of situations in which they can display their knowledge and intellectual prowess” (p. 13). Dominant individuals are described as being determined to get their own way, as being inclined to take control of situations, and being forceful in dealings with others.

**Results**

**Overall Data Structure**

At the outset, analyses focused on the overall structure of the DC ratings. This involved factor analyses to ascertain the best way to summarize raw ratings, followed by a holistic view of mean differences and correlations between self and other ratings. An exploratory principal axis factor analysis (EFA) suggested that participants rated their performance as being specific to each exercise, as is intended in a task-specific DC. Direct oblimin was chosen as the method of rotation because exercise factors are often found to be correlated in the assessment centre literature (see Lance, Noble, & Scullen, 2002, p. 231). In terms of selecting the number of factors in this analysis, a latent root criterion was used to encourage a data-driven approach. Table I shows the factor loading structure, which is
clearly interpretable as representing task-specific judgments. The overall fit for the model was reasonable, with four factors explaining nearly 75% of the variation in self-rating items. Assessor ratings were also factor analyzed and displayed, as expected, ratings that loaded clearly within exercises. This finding has been reported elsewhere on different samples (e.g., Jackson, et al., 2005) and is, therefore, not focused on here.

To back up the findings of the EFA on self-ratings, a CFA was run incorporating the model suggested above. Self-rating items were specified as observed variables and latent variables were conceptualized as indicating overall self-rated behavioural performance within exercises. Latent variables were treated as correlated effects. Figure 1 shows the CFA model, factor loadings, and accompanying fit indices. The overall fit for this model was found to be acceptable ($\chi^2 = 134.54, df = 48, p < .001; AGFI = .84, TLI = .93, and CFI = .95$).

Fundamental to this study is the idea that an overall perspective on self-raters may serve to hide important information. However, for comparative purposes, mean differences and correlations across all participants were computed. Across the entire sample, participants’ self-ratings ($M = 4.29, SD = 0.65$) tended to be around one scale point higher than their associated assessor ratings ($M = 3.26, SD = 0.74$). This difference was statistically significant ($t = 15.38, df = 418, p < .001$). The overall correlation between self and assessor ratings was weak ($r = .18, p < .01$).

**Typology by Cluster Analysis**

A cluster analysis was performed using Ward’s method of clustering to maximize homogeneity within clusters, along with squared Euclidian distance as a measure of cluster difference (see Hair, et al., 1998). Self-ratings only were used to define the cluster solution, as a key aim was to identify subgroups of self-raters. Agglomeration schedules and dendrogram were used to guide the selection of identifiable clusters (see Everitt, Landau, & Leese, 2001) and tended to suggest four relatively distinct clusters within the total sample. The agglomeration schedule and dendrogram were too large for reproduction in this paper, given the sample size of 214, but are available from the first author upon request. Standardized mean scores across self-raters within each cluster were plotted along with assessor ratings and scores on the personality and cognitive ability dimensions (see Figure 2). These constituted Z scores with a constant of 2 added to avoid negative values and thus ease interpretation ($M = 2, SD = 1$). Note that the data points in Figure 2 should be interpreted in terms of relative standing on a particular variable, as opposed to absolute values. By way of example, in the pecking order, cluster 3 self-rated lowest in comparison to the other clusters. The general research objective of this study was, thus, supported, as the cluster profiles suggested an extension of Atwater and Yammarino’s (1992) classification, using an empirical technique for developing self-rater profiles. In terms of cluster interpretation, cluster 1 could be thought of as bold in terms of their behavioural performance on exercises, cluster 2 confident, cluster 3 retiring, and cluster 4 self-effacing. Note that these categories are intended only as interpretational guides.

**Table 1. Exploratory Factor Analysis of Self-Ratings**

<table>
<thead>
<tr>
<th>Item</th>
<th>F1</th>
<th>F2</th>
<th>F3</th>
<th>F4</th>
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<tbody>
<tr>
<td>Ex1sr1</td>
<td>.90</td>
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<td></td>
<td></td>
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<tr>
<td>Ex1sr2</td>
<td>.84</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Ex1sr3</td>
<td>.88</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Ex2sr1</td>
<td>.80</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Ex2sr2</td>
<td>.89</td>
<td></td>
<td></td>
<td></td>
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<tr>
<td>Ex2sr3</td>
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<tr>
<td>Ex4sr3</td>
<td>.89</td>
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</tbody>
</table>

Principal axis factoring was used with a latent root criterion and direct oblimin rotation. F = factor, Ex = exercise, sr = self-rating item, SS = sums of squared loadings at extraction, % = cumulative percent of variance explained. Loadings < .3 were suppressed to ease interpretation.

Figure 1. CFA for self-ratings of task-specific development centre

Srex1 – srex4 = self-ratings on exercises 1 through 4.
Sr1i1 – sr4i3 = self rating items for exercises 1 through 3.

Goodness-of-fit indices suggested an acceptable model (GFI = .90, AGFI = .84, PGFI = .56, NFI = .93, TLI = .93, CFI = .95, PRATIO = .79).
Table 2. Mean Differences between Self and Assessor Ratings by Cluster

<table>
<thead>
<tr>
<th>Cluster #</th>
<th>n</th>
<th>Self Mean ± SD</th>
<th>Assessor Mean ± SD</th>
<th>Diff Mean ± SD</th>
<th>t</th>
<th>λ</th>
</tr>
</thead>
<tbody>
<tr>
<td>Cluster 1</td>
<td>60</td>
<td>4.93 ± 0.34</td>
<td>3.34 ± 0.17</td>
<td>1.59</td>
<td>14.64**</td>
<td>.36</td>
</tr>
<tr>
<td>Cluster 2</td>
<td>69</td>
<td>4.39 ± 0.26</td>
<td>3.35 ± 0.68</td>
<td>1.04</td>
<td>11.89**</td>
<td>.49</td>
</tr>
<tr>
<td>Cluster 3</td>
<td>26</td>
<td>3.47 ± 0.74</td>
<td>3.02 ± 0.82</td>
<td>0.45</td>
<td>2.05</td>
<td>.92</td>
</tr>
<tr>
<td>Cluster 4</td>
<td>59</td>
<td>3.90 ± 0.45</td>
<td>3.16 ± 0.74</td>
<td>0.74</td>
<td>6.49**</td>
<td>.73</td>
</tr>
</tbody>
</table>

* = p < .05, ** = p < .01, Diff = mean difference (self – assessor ratings). Difference scores are presented to highlight magnitude of difference only. λ = Wilks’ λ

Statistical differences among clusters were indicated using the Kruskal-Wallis H non-parametric test of difference (see Siegel & Castellan, 1988). Significant differences were observed across self-raters ($\chi^2 = 143.27$, df = 3, p < .001) which was expected, given that the cluster solution was defined by this variable. Significant differences were also observed on social boldness ($\chi^2 = 8.68$, df = 3, p < .05), intellectance ($\chi^2 = 20.59$, df = 3, p < .001), and dominance ($\chi^2 = 15.85$, df = 3, p < .01). Differences among clusters on assessor ratings and measures of cognitive ability were non-significant. A post-hoc comparison of the social boldness variable indicated that significant differences were only evident between clusters 1 and 4 (using both Dunnette’s T3 and Tukey’s HSD post hoc tests). Note, a standard ANOVA of the data suggested the same conclusions as the non-parametric tests. These results were generally in support of Hypothesis 1, showing that dominance-related personality dimensions distinguished among types of self-rater. However, Hypothesis 2 was unsupported, as cognitive abilities yielded no significant differentiation.

To provide a perspective of the data in absolute terms, mean scores by cluster were provided across self and assessor ratings (see Table 2). Statistical significance was computed, this time, to indicate differences between self and assessor ratings. All mean comparisons were statistically significant, except for cluster 3. Note that a power analysis using G power (version 2.0) indicated that, given relative sample size by cluster, it was appropriate to use $t$ tests but not correlation coefficients in this case. As such, correlations were not computed by cluster. As a further indication of effect size, Wilks’ $\lambda$ was calculated based on sums of squares from an ANOVA table (see Spicer, 2005). In this case, a larger Wilks’ $\lambda$ indicates greater congruence between self and assessor ratings.

Discussion

There are several observations that may be gleaned from the results of this study. Notably, cluster analysis is a useful profiling tool that can aid an understanding of self-rater behaviour. Atwater and Yammarino (1992) suggested three main types of self-rater, including those who, in comparison to assessor ratings, overestimate, agree, or underestimate. This system of classification has been used by several researchers, but could be criticized on the basis that it lacks an empirical grounding and, in practice, uses unreliable difference scores to allocate within the typology (see Cronbach & Furby, 1970; Johns, 1981). The use of cluster analysis, on the other hand, provides a data-driven mechanism for identifying a self-rater typology, and the results of this study build on Atwater and Yammarino by suggesting varying degrees of self-rating congruence.

Table 2 shows mean comparisons between self and assessor ratings by the four sub-clusters of self-rater identified in this study. Cluster 1 was clearly the least concordant, and the participants within this cluster rated themselves over 1.5 scale points higher,
on average, than their assessors. Cluster 3 was evidently the most concordant, and yielded a non-significant difference. Between these points were varying degrees of agreement with regard to perceptions of self-performance, and it is these empirically derived sub-clusters that should, according to this study, be scrutinized further. Note that Table 2 suggests the overall correlation between assessor and self-ratings ($r = .18, p < .01$) may hide important patterns in the data through aggregation.

Following on from the interpretation of the results presented earlier, an understanding of divisions in the perceptions of raters could help in terms of informing those who provide feedback. It is suggested that the framework identified in this organisation might assist assessors in terms of how they tailor the format, tone, expression, and style of such feedback. Further research would be needed to identify the effects of different styles on those who are bold, confident, retiring, or self-effacing, and to identify the extent to which this framework generalizes across organisations.

The cluster solution also allows typologies to be meaningfully profiled along with other variables of interest. In this case, and given the suggestions of previous research (e.g., Clapham, 1998; Islam & Zyhur, 2005), DC participants, divided into self-rater clusters, were profiled with scores on personality attributes related to social dominance and cognitive abilities. The personality attributes mentioned appeared to discriminate across those who were in one cluster versus another, however, cognitive abilities showed no such discrimination. Turning to Figure 2, a clear division occurred across self-rating clusters; however, assessor ratings did not significantly discriminate in this regard. While, in absolute terms, participants tended to rate themselves higher than assessors (as shown in Table 2), Figure 1 shows that assessors tended not to react to this, and remained relatively homogenous with regard to their ratings across clusters.

With regard to psychological measurements external to the DC behavioural judgments, the highest self-raters in cluster 1 tended to be the most socially bold, the most willing to display their intellectual prowess (intellectance) and the most dominant. The lowest self-raters tended to score lower on intellectance and dominance. Cognitive abilities did not significantly discriminate across clusters, despite the intellectance scores shown, particularly in cluster 1. One could speculate on the degree to which impression management may play a role in the manifestation of the personality attributes mentioned above (Fahr & Werbel, 1986; Halman & Fletcher, 2000), despite the notion that these ratings were taken from a developmental context. In any case, it would appear that the attempts of the participants in cluster 1 to present themselves in a dominant manner did not result in a statistically significant differentiation on the part of the assessor panel.

Another finding in this study, although less central to the profiling of self-raters, is the suggestion that participants in a task-specific DC rated their performance as being specific to exercises. Lance et al. (2004) suggested that empirical study of task-specific, as opposed to the traditional dimension-specific, ACs is sorely needed as most of the research specifically in this area is anecdotal, with a few exceptions (Franks, et al., 1998; Jackson, et al., 2005). While there is evidence that assessors regard task-specific AC performance as being situational (see Jackson et al.), this is the first known research to have looked at the perspective of the participants. It would be informative, given the construct validity issues highlighted by Lance et al., to ascertain whether or not participants in a traditional dimension-specific AC view their performance as being related to monotrait-heteromethod dimensions.

As suggested by Schmitt et al. (1986), an indication of pre-assessment self-ratings would be informative. Note, however, that the overall correlations between self and assessor ratings in this study were in line with the findings of other studies that did use pre-assessment. Further, the focus in this study was to provide an uncluttered view of performance specific to exercises, in line with a task-specific approach. As previously stated, it would also be interesting to repeat this study on the traditional dimension-specific approach to ACs to ascertain participant perceptions of performance on such measures, and to investigate the effects of using dimensions on sub-clusters of self-raters.

In this study, those who rated their AC performance highly also tended to score higher on variables related to social dominance, but their assessor ratings and cognitive ability scores were not significantly different from other subgroups of self-rater. Providing feedback to such individuals might present a challenging scenario for practitioners. Such a scenario could be compounded by making participants wait for long periods before feedback is provided (Fletcher & Kerslake, 1992). Provision of feedback for those who hold lower self-perceptions of performance would also need to be handled in a particular manner, as such individuals tended to report less dominance in social settings. Future studies could investigate the extent to which different feedback approaches could work with more or less efficacy, depending on the recipient’s self-rating cluster.

In conclusion, the results of this study suggest that cluster analysis provides an informative technique for profiling self-rater behaviour. The profiles generated in this sample build on the discrete theoretical profile of overestimation, accuracy, and underestimation suggested by Atwater and Yammarino (1992). It appears that varying degrees of concordance may be reflected within self-ratings when compared to those of external assessors. Moreover, this study suggests that personality dimensions related to social dominance may define certain sub-groups of self-raters, particularly those who rate themselves high relative to assessor ratings.

References


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