A triadic approach to assessment centre’s construct validity; The effect of categorising dimensions into a feeling, thinking, power taxonomy

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Abstract

This study examined the influence on construct validity of implementing the triad Feeling, Thinking and Power as a taxonomy for behavioural dimensions in assessment centre (AC) exercises. A sample of 1,567 job applicants participated in an AC specifically developed according to this taxonomy. Each exercise tapped three dimensions, one dimension from each cluster of the taxonomy. Confirmatory Factor Analysis of the multitrait-multimethod matrix showed evidence for construct validity. Thus the ratings matched the a priori triadic grouping to a good extent. Practical implications are discussed.

Keywords:
ASSESSMENT CENTERS, DIMENSIONS, CONSTRUCT VALIDITY,
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The effect of categorising dimensions into a feeling, thinking, power taxonomy

Since AT&T’s initial corporate application in the 1950s, the assessment center (AC) has thrived as one of the most popular methods for evaluating the performance of individuals for selection and development (e.g., Spychalski, Quinoñes, Gaugler, & Pohley, 1997). Due to the labour intensive nature of the AC architecture, practitioners are faced with numerous developmental and implementation problems. One of the problems highlighted by practitioners is how to select and define dimensions in exercises (e.g., Lievens & Goemaere, 1999). The fact that practitioners perceive problems with dimension selection and definition is consistent with the finding in the scientific AC literature that different dimensions within exercises correlate higher than similar dimensions across exercises, and that hence construct-related validity of the AC dimensions is low (e.g., Fleenor, 1996; Neidig & Neidig, 1984). Thus, from both practical and research perspectives, there is a need for a systematic procedure that enables AC developers to select independent and easily measurable dimensions, and that enables AC users to effectively distinguish between these dimensions.

The present study aims to contribute to AC practice and research by proposing a functional taxonomy of three broad dimension clusters, from which three operational dimensions are each time selected for every exercise. These clusters are Feeling (e.g. sensitivity, empathy, client orientation), Thinking (e.g. problem analysis, creativity, judgement) and Power (e.g. persuasiveness, risk taking, tenacity). The concepts feeling, thinking and power should be regarded as
category labels for clusters of behavioural dimensions. This study investigates
whether applying this taxonomy in a working AC as a means of selecting
dimensions for each exercise increases construct validity.

**A Taxonomy for Dimension Selection.**

Most AC dimensions are not completely orthogonal, which makes it
difficult for assessors to decide which behaviours go with which dimensions. AC
dimensions have been hypothesised to have a(n unknown) magnitude of true inter-
correlation which lowers discriminant validity (Sackett & Dreher, 1982). In
similar vein, in order to obtain construct validity, care should be taken to define
dimensions in a non-ambiguous and uni-dimensional manner (Joyce, Thayer, &
Pond, 1994; Kleinmann, Exler, Kuptsch, & Köller, 1995). The use of a large
number of dimensions to be rated in an exercise has been viewed to cause
cognitive overload and thus to lower construct validity (Bycio, Alvares, & Hahn,
1987; Gaugler & Thornton, 1989; Klimoski & Brickner, 1987; Silverman,
Dalessio, Woods, & Johnson, 1986). In order to compensate for this cognitive
overload, it seems that assessors themselves reduce a large number of dimensions
into a smaller manageable number of categories during the rating process (Sackett
& Hakel, 1979; Sagie & Magnezy, 1997; Shore, Thornton, & MacFarlane Shore,
1990).

For these reasons, it has been suggested to cut down the number of
dimensions to three to five per exercise (Arthur Jr., Anthony Day; Gaugler &
Thornton, 1989; McNelly, & Edens, 2001). While some studies have scrutinised
the effects of conceptual distinctiveness and the number of AC dimensions on
construct validity (see Lievens & Conway, 2001, for an overview of these studies), it has not yet brought about a functional taxonomy for AC development, and more specifically, for dimension selection.

Previous work in the fields of ACs, leadership and personality jointly provides a small set of higher order constructs that meets the requirements of conceptual distinctiveness and non-ambiguity and that could be used as an easily applicable taxonomy for selecting dimensions in exercises. Taking a step back in history, Plato, in The Republic, distinguished three classes of individual differences: cognitive, the faculty of knowing, affective, the faculty of feeling, and conative, the faculty of volition (in: Ackerman & Humphreys, 1993). Ackerman and Humphreys (1993) translated these faculties into modern terminology as mental abilities, such as intelligence or spatial ability, temperament, such as emotionality, objectivity, or masculinity-femininely, and motivation, such as effort. More contemporary studies in the relevant area’s of ACs, leadership and personality, pointed largely in the same direction. In the leadership literature, Yukl (1998) noted that managers use a threefold combination of leadership behaviour: soft, rational and power tactics. Also, Zand (1997) suggested that the three forces ‘trust’, ‘knowledge’, and ‘power’ coherently serve as the basis of effective leader performance. Within the personality literature, we are most interested in the factors that describe the domain of interpersonal behaviour, because this may be of particular importance for the AC. Within this domain, Wiggins (1973, p. 479) noted that: “The circular arrangement of variables....was postulated to be
generated by two orthogonal axes representing Power (dominance versus submission) and Affiliation (love versus hate”).

These publications led to the notion that managerial behaviour that is important in AC exercises may be effectively summarised in the three categories Feeling (trust, affiliation, soft tactics, etc.), Thinking (knowledge, rational tactics, etc.), and Power (also previously labelled as such).

Within the AC construct validity literature, some researchers have focused on dimension categorisation. It has been acknowledged that applicants are evaluated on the basis of their cognitive and interpersonal behaviours (Sagie & Magnezie, 1997; Shore et al., 1990). Shore et al. (1990) made this dual category distinction within 11 operational dimensions, between interpersonal style and performance style dimensions. This resulted in improved construct validity of the two dimension categories, compared to the construct validity of the operational dimensions. The present study follows up on Shore et al. (1990) by directly evaluating whether assessors are able to distinguish between more than two categories. The rationale for this is that assessors might be able to discern two elements within Shore’s performance style dimensions, namely Thinking (e.g. planning & organising, judgement, analytical skills, or Shore’s “recognizing priorities”) and Power (e.g. persuasiveness, assertiveness, decisiveness, or Shore’s “work drive”). This corresponds to the AC dimension distinction proposed by Jansen (1991) between ‘thinking power’ and ‘social power’. The above mentioned dyadic approaches (Wiggins, Shore, Jansen) seem to complement each other: concepts that are central in one approach are ignored in another approach and vice
versa. The triadic approach can therefore be regarded as an integration of previous
dyadic approaches.

Concluding, it seems reasonable to expect that behavioural dimensions
within AC exercises can be categorised into a threefold taxonomy: feeling,
thinking and power, and that this leads to enhanced construct validity. This paper
reports the data that result from implementing this triad in an operational AC. In
this AC, each exercise consists of three operational dimensions, one dimension
from the cluster Feeling, one from Thinking and one from Power.

Method

Participants. Participants were 1,567 Dutch job applicants (1079 male),
tested in 1999 at a psychological consulting firm. The participants applied for a
variety of jobs, mostly in management. Their mean age was 36 with a standard
deviation of 8.

Assessors. The applicants were assessed by 26 assessors and 22 role-
players. Both groups received an extensive and recurring assessor training and rate
applicants on a day-to-day basis. The rater-ratee ratio was 2:1. Assessor and role-
players were confronted with each applicant only once in order to make sure that
the exercises are independent. This procedure minimises rater bias between
exercises (Kolk, Born, & van der Flier, 2002). After completion of each exercise,
the role player and the assessor independently rated the applicants. Interrater
reliability of the ratings in the present study was obtained by calculating the mean
PPM correlation coefficient across dimensions and exercises, reaching a value of \( r = .63 \). Although this value indicates a moderate interrater reliability, it does not
deviate from previously reported reliabilities (Thornton, 1992; Thornton & Byham, 1982).

Exercises. Each applicant participated in two exercises. These exercises were not the same for each applicant, as they applied for different jobs. Several types of commonly administered exercises were used (Thornton, 1992): interview simulations with subordinates, clients and colleagues (68%), case-analyses (29%), and in basket exercises (3%)1.

Dimensions. The exercises were designed to tap three dimensions, one dimension from each of the three clusters (Feeling, Thinking and Power). In other words, the dimensions were a priori grouped within the three clusters. This a priori categorisation was based on two pre-studies. First, we asked a group of 25 psychologists to sort approximately 350 behavioural examples into the categories Feeling, Thinking, Power or neither of those. In order to ensure maximum conceptual dissimilarity, we used only those behavioural examples that fell into just one of the three categories (interrater reliability > .80) to create descriptions of the dimensions. Subsequently, we asked another team of four expert raters (psychologists with a mean rating experience of 14.5 years) to independently classify these dimensions, including their matching behavioural descriptions, into Feeling, Thinking, Power or neither of those. This categorisation procedure was done on rational grounds, following the Shore et al. (1990) and the Sagie and Magnezie (1997) studies.
Assessors rated the dimension in the AC on a 1 (low) to 5 (high) point scale, where ratings on intermediate scores (1.5, 4.8 etc.) were allowed. Table 1 shows the dimensions that we used per exercise, clustered into the proposed triad.

Analyses. Thus, the analyses were conducted on a conglomerate sample of exercises and dimensions to see if the intended dimensional triadic structure emerges. For a formal test of the hypothesis, confirmatory factor analysis (CFA) was performed. The multitrait-multimethod (MTMM) covariance matrix was analysed with LISREL 8.30 (Jöreskog & Sörbom, 1989). A commonly noted problem in analysing trait by method MTMM matrices using CFA, is the occurrence of ill-defined solutions, such as convergence problems, negative (error) variances or out-of-range factor intercorrelations (Kenny & Kashy, 1992).

As in the majority of studies examining MTMM data (e.g., Lance et al., 2000), we ran into estimation problems using the traditional CFA approach when testing some of the competing models. Kenny and Kashy (1992) suggested an alternative to testing the traditional CFA model that is not subject to the aforementioned problems: the so-called correlated uniqueness (CU) model. This approach specifies trait factors and does not create method factors, but allows its unique factors to be correlated across measures within the same method. Variances of the two methods (i.e. the exercises) were equalised throughout the models, for these can be assumed to be roughly similar.

Criteria for evaluating the competing CU models are firstly measures of overall fit, namely the $\chi^2$/degrees of freedom ratio (should approach 1), the $\chi^2$ p-
value (should not be significant), the Adjusted Goodness-of-Fit Index, which adjusts the degrees of freedom relative to the number of variables in the model (AGFI, should approach 1), and the Root Mean Square Error of Approximation which evaluates the closeness of fit given the number of degrees of freedom (RMSEA, should be lower than .05). Secondly, since we were interested in comparing several competing models, we used Akaike’s Information Criterion, which penalises for leniency (the model with the smallest AIC should be selected).

The competing CU models were interpreted following the Widaman (1985) procedure of comparing the fit of hierarchically nested MTMM models (more specifically: model E, see also Marsh, 1989). This means that a more parsimonious and therefore more restrictive model is tested against a less restrictive model using a likelihood ratio test. Generally, the more parameters there are to be estimated in a model, the better the model fits. Therefore, a less restrictive model can only be accepted when it provides a statistically significant improvement in the description of the data. CFA models based on MTMM data using this procedure always have a fixed method/trait factor intercorrelation null matrix, which is to say that trait and method factors are orthogonal (Donahue, Truxillo, Cornwell, & Gerrity, 1997; Kenny & Kashy, 1992; Widaman, 1985). In the present study, the most restrictive yet meaningful model is a method-factor only model (i.e. correlated uniquenesses), for previous research has consistently shown the appearance of these method-factors (without trait-factors). Less restrictive models add parameters, until all meaningful parameters are estimated in the complete trait by method CU model.
A method-factor only model represents a primarily halo or exercise effect, indicating that assessors do not distinguish between any of the dimensions (model I). This is the typical AC model that is usually found in previous research.

Secondly, a model is tested which adds only one general dimension factor (model II). A model with two method- (correlated uniquenesses) and two trait-factors (Feeling and Power/Thinking) represents the hypothesised structure by Shore et al. (1990) (model III). The complete trait by method CU model represents the intended triadic structure (model IV). A $\chi^2$ likelihood ratio test was used to determine whether the complete model (model IV) fits the data significantly better than models nested within the complete model (model I, II and III).

Results

Insert Table 2 about here

The MTMM correlation matrix of the two exercises is reported in Table 2. The observed correlation pattern does not meet the Campbell and Fiske (1959) MTMM criterion for establishing construct validity, for the heterotrait-monomethod correlations exceed monotrait-hetromethod correlations, instead of vice versa. This result was found in all previous AC construct validity studies, except one (i.e. Reilly, Henry, & Smither, 1990).

Insert Table 3 about here

Table 3 shows the fit indices of the CFA of the four competing CU models. The method-factor only model (model I: halo or exercise effect) did not show an adequate fit in the present study. In addition, Model II adds a general dimension factor. The $\chi^2$ value of this model is also highly significant and the
model does not fit well. Next, we tested a two traits by two methods model conforming to the Shore et al. (1990) assumption, where Feeling falls into the interpersonal style category, and Power and Thinking in the performance style category (Model III). This model also yields a significant $\chi^2$, and the fit indices are unsatisfactory. Differences in $\chi^2$/d.f., RMSEA, AGFI and AIC magnitude all indicate a superiority of the complete three traits by two methods CU model (Model IV).

A $\chi^2$ likelihood ratio test between the complete trait by method model IV and most restrictive model I (halo), indicates a significant improvement in fit ($\Delta \chi^2 = 192.54, 6; p < .001$). Similarly, the complete model also provided a significantly better description of the data than model II ($\Delta \chi^2 = 48.19, 3; p < .001$), and model III, representing the Shore et al. (1990) distinction ($\Delta \chi^2 = 33.99, 2; p < .001$). In sum, the hypothesised complete trait by method model provided the best description of our data.

As stated before, we analysed a data set consisting of a conglomerate sample of dimensions and exercises, in order to investigate whether the intended triadic structure was indeed established. In addition, we performed CFA on a sub-sample of candidates participating in two similar exercises, tapping the same dimensions (i.e. two interview simulations, $n = 560$). Results indicated again a significantly better fit for the complete two by three factor model ($\chi^2 = 0.23, 3; p = .97; \text{RMSEA} = .00$), compared to model I ($\Delta \chi^2 = 95.74, 6; p < .001$), model II ($\Delta \chi^2 = 38.17, 3; p < .001$) and model III ($\Delta \chi^2 = 22.77, 2; p < .001$).
Table 4 presents the completely standardised solution of the actual factor loadings, factor variances and intercorrelations, and the error uniquenesses of the CU model (model IV). Within this model, evidence for discriminant validity is established by the inter-factor correlations, which can be found in the lower half of Table 4. All inter-factor correlations were significant, implying that discriminant validity coefficients are still poor. Evidence for convergent validity is established if the values in the standardised matrix of the factor loadings are significant (upper left corner of Table 4), which is the case.

Kenny & Kashy (1992, p. 170) noted that the correlated uniqueness approach assumes zero method-method correlations. When this assumption is not met, it can have a biasing effect on construct validity, through artificially enhancing convergent validity and worsening discriminant validity (see also . This biasing effect may also be present in our model, for a zero method-method correlation is quite untenable in the case of ACs. This is not to say that the results from Table 3 will be negatively affected, as the fit would probably increase by adding a method-method intercorrelation². The results in Table 4, on the other hand, should be regarded with some caution, as both the factor loadings and the factor intercorrelations may be overestimated, at the expense of discriminant validity.

To sum up, on a matrix level, the present data show evidence for both discriminant and convergent validity--the complete model provides the best description of our data. On a parameter level, on the other hand, evidence for convergent validity is established, whereas evidence for discriminant validity seems weak.
Discussion

The results of the present study confirm previous findings showing that heterotrait-monomethod correlations (discriminant validity coefficients) are predominantly higher than monotrait-heteromethod correlations (convergent validity coefficients) (e.g. Sackett and Dreher, 1982; Neidig and Neidig, 1984; Bycio, et al., 1987; Klimoski and Brickner, 1987). Notwithstanding, the results extend previous research, since they also support our a priori dimension grouping, into the categories Feeling, Thinking, and Power. Confirmatory factor analysis yields a complete two methods by three traits model, providing convincing evidence for construct validity. The complete trait by method model fitted the data better than any alternative model. The method-factor only model (halo or exercise effect), which is predominantly found in several previous studies (e.g. Chan, 1996; Fleenor, 1996; Joyce et al., 1994; Schneider & Schmitt, 1992), received no support in the present study. Sagie and Magnezy (1997) and Kudish, Ladd and Dobbins (1997) found comparable CFA results. Additionally, a single dimension factor did not provide an adequate description of our data. In this light, our attempt to increase construct validity by applying the triadic taxonomy Feeling, Thinking, Power to AC dimensions seems viable.

Moreover, our taxonomy seems to be a tenable extension of the Shore et al. (1990) distinction between interpersonal and performance style dimensions. Although this dual category proposition also received support in previous research (Sagie & Magnezy 1997), this study shows that assessors are able to distinguish between three orthogonal categories. Building an AC upon the dual taxonomy
could perhaps even imply a loss of information, thereby impairing criterion-related validity (Lievens, 1998, p. 146). Therefore, in designing an AC, a triadic approach might be more legitimate.

**Implications for AC practitioners.**

This study provides an easily applicable tool for AC practice, through facilitating the selection of maximally dissimilar dimensions within exercises and by improving construct validity of these dimensions. In our own practical experience, the taxonomy proved a helpful tool in formulating conceptually dissimilar dimensions, and thereby facilitating the rating task. In addition, providing feedback in terms of Feeling, Thinking and Power resulted in greater comprehension and acceptance by assessors, as well as applicants. Receiving feedback on these three domains makes it easier to direct future development, than receiving such feedback on all dimensions individually.

The taxonomy presented in this study may be used by AC developers to select and define dimensions for exercises. Pre-existing dimensions or behavioural indicators that follow from job analysis may be categorised in the triad by (expert) raters during AC development workshops. After determining which dimensions are to be measured, each dimension can be attributed to one of the three clusters, feeling, thinking, and power. Subsequently, exercises can be developed that tap these dimensions. Ahmed, Payne, and Whiddett (1997) provided a model for exercise design that may be of help.

If a target dimension cannot be categorised in one of the three clusters, it may indicate that the dimension is in fact a multi-faceted concept (e.g., the
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dimension ‘leadership’). If so, the dimension may also be difficult to evaluate during an exercise, because the assessors may not agree on the meaning of the dimension, or they may not know which behaviours to look for. A solution might be to divide this dimension into two or three parts, that can be attributed to Feeling, Thinking or Power. For instance, ‘leadership’ may be divided into the dimensions delegation (thinking), decisiveness (power), and sensitivity (feeling), depending on the definition of leadership that derives from the job analysis.

Another solution to overcome the commonly noted confusion over the meaning of the target dimensions is a so-called frame-of-reference training, which has been recently advocated (Lievens, 2001; Schleicher & Day, 1998). This type of training aims to provide a shared performance theory for raters, such that each rater evaluates applicants on the basis of the same conceptualisation of effective performance. This is generally established during pre-assessment workshops. The Feeling, Thinking, Power triad may provide a helpful contribution to this type of workshop, in that assessors are not only trained on a shared frame-of-reference, but also on means to distinguish the dimensions during AC exercises. Practical implications may also involve adjustments in AC training programs, in which assessors are trained to classify observed behaviours into Feeling, Thinking and Power.

One possible pitfall of using a smaller number of dimensions, is defining them at a much high level of abstraction in order to cover a broad scope of behaviour (the information ‘loss’ caused by the reduction of the number of dimensions is compensated by giving the remaining dimensions a broad
definition). Lievens and Conway (2001) warned that when broadly defined dimensions are used, behaviours may overlap between those dimensions making it difficult for assessors to distinguish among them. The feeling, thinking, power taxonomy should not be misused to define broader dimensions. The taxonomy has only one purpose, and that is to carefully select dimensions - that have already been specifically defined - from each of the clusters feeling, thinking and power.

**Limitations and directions for future research.**

A first limitation of our study is that we used no more than two exercises to represent a full AC, while in practice an AC usually includes a broader range of different exercises (say five).

Another potential shortcoming in the present study is that the dimensions measured in the exercises were often similar within the three domains (Table 2). For instance, the dimensions Judgement, Tenacity and Sensitivity were used in all but three exercises. It is conceivable that using more diversified dimensions might influence the results. In addition, the clustering of dimensions into the triad was done on an a priori basis, using expert raters for classifying the dimensions. Results indicated that implementing a triadic taxonomy indeed yields a meaningful triadic latent within and across exercise structure. Thus the ratings match the a priori triadic grouping to a good extent. Yet, we did not unfold whether the Feeling, Thinking and Power domains can be adequately and fully measured within an AC. As such, the current results should be regarded as an incentive for testing this taxonomy. Future research could make an attempt to re-analyse previous research through meta-analysis, and cluster multiple dimensions
within the presently proposed taxonomy. Subsequently, structural equation models could be tested in order to confirm the clustering into the triadic taxonomy, where 1-factor models should be fitted on the clustered dimensions. However, the often lacking description of the complete MTMM matrix on a dimension level, may present a difficulty for such a re-analysis.

Another route to test whether the triadic approach is at all feasible and perhaps even superior to a dyadic approach, is to experimentally vary the dimension composition per exercise, while holding all boundary conditions constant (e.g., in one condition the dimensions are selected according to the feeling/thinking/power taxonomy, while in another condition three dimensions are randomly selected).
References


108.


Author Note

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Footnotes

1 By adding up several types of exercises, there could be more common variance shared by the dimensions than by the exercises, due to for instance a difference in exercise ‘difficulty’. This might result in stronger dimension factors. A multivariate analysis of variance on the dimension scores, using type of exercise as the independent variable, revealed a significant yet small effect (Pillai’s Trace for exercise 1 and 2, respectively .09/.10; F: 10.81/12.05, p < .05). So, only 3% of the variance was accountable for a difference in exercise ‘difficulty’.

2 Because of these difficulties, we also performed a traditional correlated trait-correlated method CFA in addition to the correlated uniqueness model. (This analysis was possible within the complete sample (N=1567), but not in the extracted subsample (n=560), because the solution did not converge.) CFA of the correlated trait-correlated method revealed similar results as the correlated uniqueness model (model IV: $\chi^2$: 3.06, 3, RMSEA: .00; AGFI: 1.00; AIC: 39.06). Further details of this analysis are available from the first author.

3 We thank an anonymous reviewer for this suggestion.
Table 1

Division of currently used AC dimensions into triad Feeling-Thinking-Power

<table>
<thead>
<tr>
<th>Exercises</th>
<th>FEELING</th>
<th>THINKING</th>
<th>POWER</th>
</tr>
</thead>
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<tr>
<td>Case analysis</td>
<td>Sociability</td>
<td>Analytical Skills</td>
<td>Tenacity</td>
</tr>
<tr>
<td>Coaching interview</td>
<td>Sensitivity</td>
<td>Judgement</td>
<td>To give direction</td>
</tr>
<tr>
<td>Client Interview</td>
<td>Client Orientation</td>
<td>Judgement</td>
<td>Tenacity</td>
</tr>
<tr>
<td>Subordinate interview</td>
<td>Sensitivity</td>
<td>Judgement</td>
<td>Tenacity</td>
</tr>
<tr>
<td>Staff Meeting</td>
<td>Co-operation</td>
<td>Judgement</td>
<td>Tenacity</td>
</tr>
<tr>
<td>In-Basket exercise</td>
<td>Sensitivity</td>
<td>Analytical Skills</td>
<td>Delegating and Control</td>
</tr>
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Table 2

<table>
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<td>3. Power</td>
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<td>4. Feeling</td>
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<td>5. Thinking</td>
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<td>.20</td>
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<td>6. Power</td>
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<td>.16</td>
<td>.27</td>
<td>.40</td>
<td>.62</td>
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Note: all correlations are significant at the 0.01 level. Cases are excluded listwise. N = 1567. Mean heterotrait-monomethod correlation: $r = .52$; mean monotrait-heteromethod correlation: $r = .23$
Table 3

Fit indices for the Correlated Uniqueness Models

<table>
<thead>
<tr>
<th>Models</th>
<th>$\chi^2$</th>
<th>d.f.</th>
<th>$\chi^2$ / d.f.</th>
<th>RMSEA</th>
<th>AGFI</th>
<th>AIC</th>
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<td>I. 2 orthogonal method correlated errors</td>
<td>195.02*</td>
<td>9</td>
<td>21.67</td>
<td>.11</td>
<td>.91</td>
<td>219.02</td>
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<tr>
<td>II. 2 orthogonal method correlated errors + 1 general dimension factor</td>
<td>50.67*</td>
<td>6</td>
<td>8.44</td>
<td>.07</td>
<td>.96</td>
<td>80.67</td>
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<tr>
<td>III. 2 orthogonal method correlated errors + 2 oblique trait-factors (Shore et al.)</td>
<td>36.47*</td>
<td>5</td>
<td>7.94</td>
<td>.06</td>
<td>.97</td>
<td>68.47</td>
</tr>
<tr>
<td>IV. 2 orthogonal method correlated errors + 3 oblique trait-factors</td>
<td>2.48</td>
<td>3</td>
<td>.83</td>
<td>.00</td>
<td>1.00</td>
<td>38.48</td>
</tr>
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</table>

Note: N=1567

* p < .001
Table 4

Parameter Estimates for the complete Trait by Method CU model (Model IV)

<table>
<thead>
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<th>Source</th>
<th>7</th>
<th>8</th>
<th>9</th>
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</tr>
<tr>
<td>Exercise 1</td>
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Factor Intercorrelations

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Note: Values of 1 and 0 were fixed a priori.

*p < .05