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## Exploratory Structural Equation Modelling (ESEM): application to the SET-37 questionnaire for students' evaluation of teaching

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### Abstract

This study contains a re-validation of the SET37-questionnaire for student evaluations of teaching (Mortelmans & Spooren 2009), using both confirmatory factor analysis and exploratory structural equation modelling (ESEM, Asparouhov & Muthén 2009), that integrates both exploratory and confirmatory factor analyses. In sum, the results provide strong evidence for the relevance of the questionnaire as CFA- and ESEM-models provide a good fit to the data. Although, as expected, item factor loadings and correlations between the twelve factors in the instrument are substantially lower in the ESEM-results. CFA-models which require zero cross-loadings might overvalue factor correlations when testing SET-instruments. The implications for future validation strategies in SET are discussed.

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### 1. Introduction

Nowadays, students' evaluations of teaching (SET) are used as a measure of teaching performance in almost every institution for higher education throughout the world. Universities and university colleges have developed more or less complex procedures and instruments to collect, analyze and interpret these data as the dominant and sometimes sole indicator of teaching quality. This widespread use has much to do with the (apparent) ease of collecting the data and presenting and interpreting the results. In most institutions, SET are used for both formative (i.e., students' feedback for the improvement of teaching) and summative (i.e., mapping teaching competence for administrative decision-making and institutional audits) reasons. This double use makes SET very delicate. On the one hand, most teachers are convinced of the usefulness of SET as an instrument for feedback on their teaching (Richardson 2005). As a consequence, many instructors gratefully make use of student evaluations for the improvement of their subsequent teaching. On the other hand, it is argued that nowadays the principal purpose of SET lies in its use as a measure for quality monitoring and administrative policy-making and mapping whether or not teachers reach a certain required standard in their teaching practice (Penny & Coe 2004). It is clear that SET-instruments should be tested and re-tested extensively on their psychometric value, especially when they are used

for administrative decision-making. Still, many instruments remain to be *ad hoc* instruments that were not tested at all (Richardson 2005).

## 2. Objectives

In this contribution, we report on a re-validation procedure of the SET37 questionnaire designed at the University of Antwerp (Belgium). This instrument consists of twelve dimensions that are measured by means of 37 Likert Items and is based on both educational theory and empirical testing (exploratory factor analysis –EFA– and confirmatory factor analyses –CFA) (Spooren, Mortelmans & Denekens 2007). A re-validation procedure using confirmatory factor analysis provided strong evidence for the established dimensional structure (Mortelmans & Spooren 2009). However, Marsh et al. (2009) and Asparouhov & Muthén (2009) argued that, especially in psychological instruments that mostly have a well-defined EFA structure, CFA approaches do not represent such a structure appropriately. These measurement instruments often have small cross loadings that are well motivated by the theory (i.e., an item that measures a teacher’s presentation skills might have small loadings on other dimensions concerning effective teaching in a SET-instrument as well), but are fixed to zero in a CFA approach. Model modifications to compensate for these misspecifications when searching for a well-fitting model often lead to distorted factors with overestimated factor correlations (Marsh et al. 2009). The new exploratory structural equation modelling methods (ESEM) therefore integrate the advantages of EFA (using factor loading matrix rotations) and CFA (access to all usual SEM parameters).

This note therefore contains a re-validation of the SET37 questionnaire by using both a CFA-approach and the more recent ESEM-approach (using an oblique rotation) on the same dataset. This allows a comparison of the outcomes of both approaches concerning the dimensional structure in the instrument.

## 3. Results

**Instrument.** For this re-validation study, we used the SET37-questionnaire which represents 12 quasi-balanced scales. All 37 items are measured on a six-point scale and were used in an online SET-procedure (for example items, see Appendix A).

**Participants.** SET were administered during the fall semester of the 2008-2009 academic year. A total of 2837 evaluation forms completed by students enrolled in 75 courses were analysed in the present study. Evaluations were completed in various educational programs by both undergraduate and graduate students.

**Re-validation of the SET37: CFA versus ESEM.** The goodness of fit indices of both analyses (CFA and ESEM) are presented in Table 1. Item loadings and factor correlations of the CFA model are presented in Table 2, the detailed parameters of the ESEM model are provided in Table 3.

Table 1: Summary of Goodness of Fit Statistics for Total Group CFA and ESEM Models

Model	$\chi^2/df$	NFParm	CFI	TLI	RMSEA	CI RMSEA	AIC	BIC	corBIC
CFA	4175.581/563	177	.939	.928	.048	.046 - .049	267914.065	268967.304	268404.913
ESEM	1523.499/288	452	.979	.952	.039	.037 - .041	265811.983	268501.610	267065.448

Note. CFI = comparative fit index; TLI = Tucker-Lewis Index; NFParm = number of free parameters; AIC = Akaike’s Information Criterion; BIC = Bayesian Information Criterion; corBIC = sample-size adjusted BIC; RMSEA = root mean square error of approximation; CI RMSEA = 90 percent confidence interval RMSEA.

Both approaches provide an excellent fit to the data, although the  $\chi^2$  tests of exact fit are significant whereas the objective is to achieve a nonsignificant p-value ( $< .05$ ). However, Hatcher (1994) indicates that a significant  $\chi^2$  does not make a confirmatory factor analysis model inadequate. The fit indices of the ESEM model suggest a better fit than the corresponding CFA model (Table 1).

Table 2: ICM-CFA Solution – Twelve ESEM Factors Based on Responses to 37 Items

Factors	Factor loadings											
Items	1	2	3	4	5	6	7	8	9	10	11	12
Clarity of objectives (F1)												
CO1	<b>.83</b>	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00
CO2	<b>.86</b>	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00
CO3	<b>.80</b>	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00
Value subject-matter (F2)												
VSM1	.00	<b>.99</b>	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00
VSM2	.00	<b>.99</b>	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00
VSM3	.00	<b>.89</b>	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00
Build-up subject-matter (F3)												
BSM1	.00	.00	<b>.75</b>	.00	.00	.00	.00	.00	.00	.00	.00	.00
BSM2	.00	.00	<b>.59</b>	.00	.00	.00	.00	.00	.00	.00	.00	.00
BSM3	.00	.00	<b>.78</b>	.00	.00	.00	.00	.00	.00	.00	.00	.00
Presentation skills (F4)												
PS1	.00	.00	.00	<b>.98</b>	.00	.00	.00	.00	.00	.00	.00	.00
PS2	.00	.00	.00	<b>.92</b>	.00	.00	.00	.00	.00	.00	.00	.00
PS3	.00	.00	.00	<b>.92</b>	.00	.00	.00	.00	.00	.00	.00	.00
Organization course – learning (F5)												
OC1	.00	.00	.00	.00	<b>.96</b>	.00	.00	.00	.00	.00	.00	.00
OC2	.00	.00	.00	.00	<b>.90</b>	.00	.00	.00	.00	.00	.00	.00
OC3	.00	.00	.00	.00	<b>.79</b>	.00	.00	.00	.00	.00	.00	.00
Course materials (F6)												
CM1	.00	.00	.00	.00	.00	<b>.99</b>	.00	.00	.00	.00	.00	.00
CM2	.00	.00	.00	.00	.00	<b>.99</b>	.00	.00	.00	.00	.00	.00
CM3	.00	.00	.00	.00	.00	<b>.86</b>	.00	.00	.00	.00	.00	.00
CM4	.00	.00	.00	.00	.00	<b>.99</b>	.00	.00	.00	.00	.00	.00
Course difficulty (F7)												
CD1	.00	.00	.00	.00	.00	.00	<b>.87</b>	.00	.00	.00	.00	.00
CD2	.00	.00	.00	.00	.00	.00	<b>.84</b>	.00	.00	.00	.00	.00
CD3	.00	.00	.00	.00	.00	.00	<b>.87</b>	.00	.00	.00	.00	.00

Help teacher during learning process (F8)												
HT1	.00	.00	.00	.00	.00	.00	.00	.00	<b>.71</b>	.00	.00	.00
HT2	.00	.00	.00	.00	.00	.00	.00	.00	<b>.71</b>	.00	.00	.00
HT3	.00	.00	.00	.00	.00	.00	.00	.00	<b>.77</b>	.00	.00	.00
Authenticity of the examination(s) (F9)												
AE1	.00	.00	.00	.00	.00	.00	.00	.00	.00	<b>.92</b>	.00	.00
AE2	.00	.00	.00	.00	.00	.00	.00	.00	.00	<b>.99</b>	.00	.00
AE3	.00	.00	.00	.00	.00	.00	.00	.00	.00	<b>.99</b>	.00	.00
Linking-up with fore-knowledge (F10)												
FK1	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	<b>.80</b>	.00
FK2	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	<b>.94</b>	.00
FK3	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	<b>.99</b>	.00
Content validity examination (F11)												
CVE1	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	<b>.83</b>
CVE2	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	<b>.89</b>
CVE3	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	<b>.89</b>
Formative evaluation(s) (F12)												
FE1	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	<b>.96</b>
FE2	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	<b>.99</b>
FE3	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	.00	<b>.97</b>

*Factor Correlations*

	<i>F1</i>	<i>F2</i>	<i>F3</i>	<i>F4</i>	<i>F5</i>	<i>F6</i>	<i>F7</i>	<i>F8</i>	<i>F9</i>	<i>F10</i>	<i>F11</i>	<i>F12</i>
F1	1.00											
F2	.52	1.00										
F3	.65	.68	1.00									
F4	.65	.55	.62	1.00								
F5	.47	.51	.52	.52	1.00							
F6	.66	.46	.63	.71	.40	1.00						
F7	.57	.50	.53	.52	.35	.48	1.00					
F8	.67	.46	.61	.67	.56	.51	.54	1.00				
F9	.31	.33	.38	.30	.58	.24	.11	.34	1.00			
F10	.40	.48	.45	.33	.32	.30	.61	.31	.09	1.00		
F11	.56	.47	.54	.41	.43	.41	.50	.49	.36	.35	1.00	
F12	.55	.32	.38	.37	.59	.39	.29	.55	.42	.22	.34	1.00

*Note.* ICM-CFA = independent clusters model-confirmatory factor analysis. This model assumed an independent cluster structure in which each of the SET37-items was allowed to load on only one latent factor and cross-loadings were not allowed. As all nontarget loadings are constrained to zero, only the target loading relating each item to its priori factor is presented. All parameter estimates are completely standardized. N = 2837 evaluations for 75 courses.

The factor loadings in the CFA-solution are very high and significant (Table 2), whereas the nontarget loadings are necessarily set to zero. The factor correlations are high as well, as they exceed .50 several times. This might be interpreted as a threat to the discriminant validity of the various dimensions in the instrument as it looks as they measure the same construct.

**Table 3: ESEM Solution – Twelve ESEM Factors Based on Responses to 37 Items**

<i>Factors</i>	<i>Factor Loadings</i>											
<i>Items</i>	<i>1</i>	<i>2</i>	<i>3</i>	<i>4</i>	<i>5</i>	<i>6</i>	<i>7</i>	<i>8</i>	<i>9</i>	<i>10</i>	<i>11</i>	<i>12</i>
Clarity of objectives												
CO1	<b>.699</b>	.042	.036	.104	.037	-.052	.037	.033	.039	.036	.009	.056
CO2	<b>.496</b>	.070	.039	.019	.118	.007	.056	.031	.012	.181	.052	.120
CO3	<b>.551</b>	-.002	.111	.003	.044	.098	.069	-.022	.059	.055	.107	.027
Value subject-matter												
VSM1	.079	<b>.784</b>	.053	.081	-.004	.017	.046	.025	.027	-.012	.010	.019
VSM2	-.020	<b>.665</b>	.133	.035	.036	.025	.063	.009	.044	.074	.040	.011
VSM3	.030	<b>.516</b>	.046	.011	.024	.124	.015	.043	.124	.007	.085	-.001
Build-up subject-matter												
BSM1	.135	.082	<b>.584</b>	.089	.028	.066	-.005	.016	.028	.004	.034	-.007
BSM2	-.006	.112	<b>.579</b>	-.029	.022	-.066	.009	.086	.005	.182	.041	.050
BSM3	.052	.081	<b>.600</b>	.060	.041	.110	.098	.037	.099	.002	.075	-.009
Presentation skills												
PS1	.080	.099	.027	<b>.763</b>	.032	.034	.040	.027	.019	.034	.031	.021
PS2	.077	.012	.116	<b>.648</b>	.086	.068	.098	.003	.019	.033	.017	-.010
PS3	.012	.054	.010	<b>.575</b>	.132	.021	.011	.028	.025	.297	.014	.026
Organization course-learning												
OC1	-.003	.025	.008	-.012	.040	<b>.737</b>	.025	.076	.024	.037	.029	.034
OC2	-.024	-.007	.046	.031	-.037	<b>.636</b>	.012	.024	.042	.084	.017	.126
OC3	.067	.186	.025	.174	.046	<b>.383</b>	-.018	.186	.011	-.008	.015	.066
Course materials												
CM1	.103	-.024	.131	.159	<b>.624</b>	.012	.012	-.012	.012	-.010	.008	.054
CM2	.057	.051	.032	.048	<b>.736</b>	.008	.069	.013	.041	.113	.039	.036
CM3	.065	.020	.312	.101	<b>.488</b>	.022	.009	-.005	-.016	.024	.005	.082
CM4	.117	.069	.017	.129	<b>.684</b>	.067	.074	.028	.014	.002	.057	.001
Course difficulty												
CD1	.031	.036	.033	.005	.068	-.029	<b>.621</b>	-.012	.114	.098	.040	.008
CD2	.025	.062	.052	.076	-.011	.040	<b>.695</b>	-.031	.134	.021	.035	.017
CD3	.084	.039	.015	.046	.044	.042	<b>.737</b>	.003	.036	.052	.075	.018

Help teacher during learning process												
HT1	.131	.012	.089	.089	-.080	.153	.037	.008	.021	<b>.431</b>	.083	.057
HT2	.116	-.001	.054	.058	-.016	.008	.065	.035	-.005	<b>.626</b>	.037	.077
HT3	.023	.036	.044	.083	.070	.103	.066	.005	.013	<b>.659</b>	.024	.039
Authenticity of the examination(s)												
AE1	-.048	.004	.012	.006	.003	.047	-.062	<b>.638</b>	-.034	.034	-.027	.110
AE2	.054	.044	.025	.006	.003	.116	-.013	<b>.734</b>	-.011	.006	.048	.032
AE3	.025	.015	.059	.031	.002	.042	.023	<b>.776</b>	.013	.018	.063	.010
Linking-up with fore-knowledge												
FK1	-.007	.161	.046	.003	.083	-.046	.031	-.003	<b>.538</b>	.071	.066	.073
FK2	.031	.002	.055	.030	-.033	.019	.187	.006	<b>.689</b>	.000	-.025	-.020
FK3	.043	.030	.013	.009	.005	.061	.032	-.017	<b>.889</b>	-.009	.036	.010
Content validity examination												
CVE1	-.049	-.015	.072	.018	.034	-.048	.059	.040	.020	.094	<b>.691</b>	.075
CVE2	.037	.038	.012	.014	-.002	.058	.012	-.013	.021	-.009	<b>.769</b>	.010
CVE3	.165	.074	.039	.018	.022	.077	.077	.094	.033	.016	<b>.634</b>	-.030
Formative evaluation(s)												
FE1	.193	.052	.001	.064	.013	.119	.010	.023	.005	-.023	.041	<b>.611</b>
FE2	.013	-.013	.029	-.011	.012	.039	.012	.030	.016	.047	.012	<b>.793</b>
FE3	.018	.019	.008	-.002	.034	.091	.020	.079	.017	.109	.018	<b>.687</b>

Factor Correlations

	F1	F2	F3	F4	F5	F6	F7	F8	F9	F10	F11	F12
F1	1.000											
F2	.221	1.000										
F3	.311	.347	1.000									
F4	.325	.270	.280	1.000								
F5	.311	.162	.315	.383	1.000							
F6	.182	.235	.195	.240	.106	1.000						
F7	.262	.214	.212	.231	.185	.123	1.000					
F8	.106	.162	.167	.116	.052	.307	-.023	1.000				
F9	.166	.265	.199	.115	.091	.144	.374	-.007	1.000			
F10	.305	.163	.286	.339	.177	.232	.260	.123	.105	1.000		
F11	.262	.204	.233	.127	.131	.181	.238	.157	.154	.205	1.000	
F12	.27	.110	.135	.118	.154	.306	.079	.211	.077	.276	.121	1.000

Note. ESEM = exploratory structural equation modeling; All parameter estimates are completely standardized. N = 2837 evaluations for 75 courses. A priori targets are set in bold..

The correlations between the factors for the ESEM solution are much smaller, with the strongest correlation being .38. Moreover, the factor loadings of the items on their posited indicators are much lower –and more realistic– as well. Their loadings on other factors, which are set free in the ESEM model, appear to be not that high and suggest their unidimensionality (i.e., they belong to only one factor).

In sum, both the CFA model and the ESEM model fit the data adequately and provide further evidence concerning the multidimensional structure of the SET37 questionnaire. Still, the ESEM approach takes into account the typical and complex EFA-structure of many measurement models in psychology and educational sciences, and provides results that are more in alignment with the basic theoretical assumptions of these models. These assumptions allow factor indicators in SET-instruments to have (low) loadings on nontarget factors. It is therefore recommended to use ESEM validation procedures next to the more common CFA-models when mapping the validity of SET-instruments.

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