Dimensionality and measurement invariance in the Satisfaction with Life Scale in Norway

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Abstract

Purpose Results from previous studies examining the dimensionality and factorial invariance of the Satisfaction with Life Scale (SWLS) are inconsistent and often based on small samples. This study examines the factorial structure and factorial invariance of the SWLS in a Norwegian sample.

Methods Confirmatory factor analysis (AMOS) was conducted to explore dimensionality and test for measurement invariance in factor structure, factor loadings, intercepts, and residual variance across gender and four age groups in a large (N = 4,984), nationally representative sample of Norwegian men and women (15–79 years).

Results The data supported a modified unidimensional structure. Factor loadings could be constrained to equality between the sexes, indicating metric invariance between genders. Further testing indicated invariance also at the strong and strict levels, thus allowing analyses involving group means. The SWLS was shown to be sensitive to age, however, at the strong and strict levels of invariance testing.

Conclusion In conclusion, the results in this Norwegian study seem to confirm that a unidimensional structure is acceptable, but that a modified single-factor model with correlations between error terms of items 4 and 5 is preferred. Additionally, comparisons may be made between

the genders. Caution must be exerted when comparing age groups.

Keywords Satisfaction with life · Mental health · Measurement invariance · Gender · Age · Dimensionality · Partial measurement invariance

Abbreviations

AMOS Analysis of moment structures
CFA Confirmatory factor analysis
CFI Comparative fit index
CI Confidence interval
EM Expectation-Maximization
ML Maximum likelihood

PGFI Parsimony goodness of fit index

PRATIO Parsimony ratio

PNFI Parsimony normed fit index

RMSEA Root mean square error of approximation

SE Standard error

SPSS Statistical Package for Social Sciences

SWLS Satisfaction with life score F1 Simple one-factor model

F2 Two-factor model

F1cov One-factor model with covariance between

residuals of items 4 and 5

ADF Asymptotically distribution free

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Introduction

Satisfaction with life is one of several aspects of positive mental health. It is not a direct, verifiable experience, nor a known personal fact, but a cognitive product that involves a comparative process between the individual's current life situation and internalized standards, allowing respondents



Item	Question	Strongly disagree	Disagree	Disagree slightly	Neither agree nor disagree	Agree slightly	Agree	Strongly agree
1	In most ways my life is close to ideal	2.1	6.2	6.6	12.6	24.1	37.4	11.0
2	The conditions of my life are excellent	1.4	3.2	4.1	7.8	16.7	46.5	20.2
3	I am satisfied with my life	1.1	2.9	4.8	5.9	14.9	48.1	22.5
4	So far, I have gotten the important things I want in life	1.6	4.2	6.1	9.5	22.6	40.2	15.8
5	If I could live my life over, I would change nothing	4.8	10.2	12.1	12.6	21.7	28.4	10.2

Table 1 Overview of the five items of the Satisfaction with Life Scale—percent response for each item (Norwegian health interview survey 2005: N = 4984)

to use the information they subjectively deem relevant when evaluating their own lives [1].

The Satisfaction with Life Scale (SWLS) [2, 3] is perhaps the most commonly used measure of life satisfaction worldwide. The scale consists of five statements (Table 1) and was originally developed to circumvent problems inherent in previous scales based on single items, or scales based toward domain or culture-specific items. As people derive their life satisfaction from different sources and vary considerably in their ideas about what constitutes a good life, the SWLS measures people's perception of their life as a whole, using items that are supposedly free from the varying criteria people use when evaluating their lives. The scale thus reflects a global evaluative judgment, partly determined by the respondent's current mood and immediate context, and partly by stable personality factors [4, 5] and genetic influences [6].

Although the SWLS is extensively studied and shows good psychometric properties including validity, internal consistency, and test–retest reliability [2, 3, 7, 8], there are still important issues that need to be addressed.

One issue concerns the dimensionality of the scale. Many studies have supported a unidimensional model, attesting a single latent factor accounting for a majority of the variance in life satisfaction scores [2, 9–12]. Some of these studies were based on traditional factor analysis, however, and when there are well founded hypotheses about dimensionality, confirmatory factor analysis (CFA) is a preferred analytical method. Some studies report the fifth item to be more weakly associated with the latent life satisfaction construct than the remaining four items (Table 1). Other studies claim essential, but not strict unidimensionality, as item 5 shows a weaker association with the latent variable than the remaining four items [13– 18]. Yet other studies support a modified unidimensional structure [19]. Some studies even suggest that a two-factor structure consisting of strongly correlated "present" (i.e., items 1-3 measure the status at the moment) and "past" (i.e., items 4 and 5 measure the individual to reflect the status over the life sequence) factors should be considered [14, 20]. Most studies involved small, non-random samples, however.

Another issue concerns the *invariance* of the scale. Measurement invariance indicates that the same underlying construct is measured across the relevant comparison groups. This ensures that group differences can be interpreted in terms of group differences in the underlying construct. Should the assumption of invariance not hold, comparisons across groups may not be valid, the subsequent interpretations may not be meaningful, and the conclusions incorrect.

Findings concerning the invariance of the SWLS are somewhat inconsistent. Some studies have reported the SWLS to be invariant (factor loadings, unique variances, factor variance) across gender [21] and age groups [22–24], whereas other studies have reported sensitivity to either sex [25] or age [26]. These inconsistencies may partly be explained by inadequate sample sizes and/or composition of samples. To explore invariance sufficiently well, respondents should represent the entire adult life span and both genders. Most studies, however, are based on small to moderately sized, e.g., [21, 26, 27] or highly homogenous samples such as Spanish junior high school students [25], Taiwanese [18] and British [21] university students and Swedish student teachers [14] and consequently exhibit both a restricted age range, biased sex ratio, and limited socio-demographic profiles.

This study explores the dimensionality and measurement invariance of the SWLS across gender and age in a large (N=4,984), nationally representative subsample of persons aged 15–79, thus including both male and female participants from emerging to older adulthood. The respondents are Norwegian and along with the other Scandinavian countries, Norwegian SWLS scores generally rank among the highest in the world, perhaps due to the distribution of welfare benefits in these countries. Scandinavian studies may therefore provide insights into differences in SWLS that may relate to benefits associated with the welfare state that attempt to equalize income and social/health benefits over the entire age span.



Method

Sample

The data are from the 2005 wave of a regularly repeated (every 3 years) health investigation in Norway. The crosssectional investigation is based on a nationally representative subsample of 10,000 persons living at home. The data are selected to be representative based on a stratified selection by municipality of residence. Information was collected through a postal questionnaire (one reminder) that each individual completed and returned through the postal services. Of the 9,187 that received the questionnaire, 5,212 responded (57%). Individuals with 3 or more missing values on SWLS, or missing gender or age were removed prior to analysis, leaving altogether 4,984 respondents. The final sample consisted of 2,369 men (mean age 46.2 years) and 2,615 women (mean age 44.1 years). Sample size by age group can be found in Table 3.

The study was approved by the Regional Committees for Medical Health and Research Ethics, and each participant gave informed consent.

Measures

Satisfaction with life was measured using the five-item Satisfaction with Life Scale (SWLS) [2, 3]. Responses were rated on a Likert scale ranging from 1 (strongly disagree) to 7 (strongly agree) (Table 1). This battery includes the following five questions:

"Using the 1–7 scale below, indicate your agreement with each of the items by placing the appropriate number on the line preceding that item. Please be open and honest in your responding.

- 1. In most ways my life is close to ideal
- 2. The conditions of my life are excellent
- 3. I am satisfied with my life
- 4. So far, I have gotten the important things I want in life
- 5. If I could live my life over, I would change almost nothing"

Statistical methods

All preliminary analyses were performed by the Statistical Package for Social Sciences (SPSS) version 17.0. Factor analysis operations were then conducted using Maximum Likelihood (ML) estimation by means of Analysis of Moment Structures (AMOS 17) [28].

There were 26, 33, 33, 67, and 15 cases with missing values for questions 1–5, respectively. For respondents with two or less missing items, the Expectation Maximization

(EM) option in SPSS was used to impute missing values for each SWLS item using the remaining SWLS items. The EM procedure is a process of regression imputation based on the observed relationship between variables. Missing values are replaced iteratively until successful iterations are sufficiently similar, and yield a complete set of data.

The data were handled as continuous data based on observations that 7-point Likert scales are best handled using continuous methodology [29]. To test the validity of handling the data as continuous, the analyses were repeated using Bayesian methodology, which is the preferred method for ordinal data.

To evaluate the dimensional structure, we performed confirmatory factor analyses (CFA) [30] using responses both from the entire sample and from each of the different subgroups. The analyses were run by means of ML estimation. The use of ML estimation can cause problems when using non-normal data, but is considered to be robust when used with moderately non-normal data from large samples [31].

The data were tested for normality and found to be univariate normal (highest kurtosis value was 2.42) [32], but not multivariate normally distributed (multivariate kurtosis was equal to 25.4; Marida's normalized estimate 73.9) [33]. The analyses were therefore repeated using asymptotic free distribution (ADF) estimation. In addition, the data were normalized using Tukey's formula, and ML estimation repeated on normalized data. Finally, results of analyses using ML were tested with Bootstrapping, using 2,000 samples, 95% CI, and significance tested with bias corrected confidence intervals.

Due to inconsistencies in the previous literature regarding the factorial structure (dimensionality) of the scale (Table 2), two alternative baseline models were specified. Altogether four models were tested in this study:

- 1. A simple one-factor model
- 2. A two-factor model including "past" (last two items) and "present" (first three items)
- 3. A modified one-factor model allowing the residual terms of items 4 and 5 to be correlated. This model is nested under model 1 and the modification based on modification indices (Fig. 1)
- 4. A model testing for inter-item correlation when the items were presented consecutively and successively rather than scattered throughout the questionnaire [34]

As the chi² has been shown to be problematic for assessing model fit in large samples [33, 35], model fit was primarily assessed using the root mean square error of approximation (RMSEA) with values of 0.08, 0.05, and 0, and the comparative fit index (CFI), with values 0.90, 0.95, and 1.0 demonstrating reasonable, close, and exact fit, respectively. It is strongly recommended to include



Table 2 Overview of the literature examining dimensionality of the Satisfaction with Life Scale

Study reference	Sample characteristics	Sample size	Gende	er	Age		
Author			Male	Female	Range	Average	
One-factor solution							
Anaby et al. [9]	Israeli adults	487	190	297	27-60		
Arrindel et al. [10]	Dutch young adults	2,800	888	887	18-30		
Atienza et al. [25]	Spanish junior high students	2,080	1,023	1,057			
Balatsky and Diener [11]	Soviet students	116				18.9	
Blais et al. [22]	French-Canadian students	871					
	French-Canadian elderly	313					
Durak et al. [23]	Turkish univ students, correctional officers and elderly adults (3 groups)	547, 166 and 123				20.7, 37.2, 68.2	
Lewis et al. [12]	Czech university students	109	38	71		23.0	
Oishi [15]	Chinese and American students	556 chinese; 442 American					
Pons et al. [26]	Spanish junior high students	266	65	65	11-15		
	Spanish elderly		68	65	60-91		
Shevlin et al. [21]	Undergraduates	258	173	85	18–57	20.6 (m) versus 22.9 (f)	
Swami and Chamorro-Premuzic [41]	Malay community sample	816					
Vaultier et al. [34]	Group 1 Successive item presentation	494	233	261		47.7	
	Group 2 Scattered item presentation	795	334	461		37.1	
Not one-factor solution (Modified	l 1- or 2-factor models)						
Clench-Aas et al. (this study)	Community sample	4,984	2,369	2,615	16–79	46.2 (M); 44.1(F)	
Gouveia et al. [13]	Five groups, high school students, teachers undergraduate students, physicians, general population	2,180 (306–797)				(21–43)	
Hultell and Gustavsson [14]	Swedish student teachers	2,900	453	2,447		28.9	
Sachs [19]	Hong Kong University students	123	43	80		32	
Slocum-Gori et al. [17]	Canadian (BC) adults	410	239	166	18-90	46.9	
Wu and Yao [18]	University students	476	207	269			

measures of parsimony that control for degrees of freedom, especially when testing complex models [33, 35]. Parsimony was evaluated here using the parsimony ratio (PRATIO), the parsimony goodness of fit index (PGFI) and the parsimony normed fit index (PNFI).

Testing of measurement invariance was conducted by multigroup CFAs using ML estimation in AMOS 17. This method employs successive analyses where constraints to the models are added consecutively. The baseline model is an unconstrained model, with one-factor loading constrained to unity. The weak (metric) model, nested under the baseline model, constrains the factor loadings to be equal across groups. Non-significance at this level allows comparing relationships. The strong (scalar) model also constrains factor loadings and intercepts to equality across

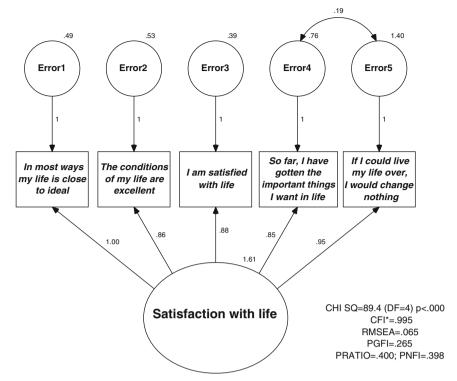
comparison groups, thus allowing comparing means, and the strict model additionally constrains the residuals. Invariance at the strict level is very seldom achieved.

The chi² alone was not deemed useable in this large sample, but the ΔChi^2 was measured and reported when comparing the model fit in different subgroups. The ΔCFI was considered a more appropriate test, however, and a cut-off ≤ 0.01 has been suggested when testing for significant differences between subgroups [33, 35].

Partial measurement invariance was examined with a successive removal of constraints at each level of invariance testing based on examination of modification indices [36]. Level 1 removed constraints on factor loadings one by one. Thereafter constraints were removed for the intercepts keeping the factor loading structure achieved in the partial



Fig. 1 The best fitting model, with unstandardized estimates, based on results of Confirmatory Factor Analysis of the five items in the Satisfaction with Life Scale. This modified one-factor model (correlation between item 4 and 5) was used for all further analyses (Norwegian health interview survey 2005; N = 4,984)



*CFI – Comparative fit index; RMSEA – Root-mean-square error of approximation; PGFI – Parsimony goodness of fit index; PRATIO – Parsimony ratio; PNFI – Parsimony normed fit index

variance testing at level 1. This method was then repeated at the strict level with progressive removal of constraints on the variances [37]. Tables 5 and 6 indicate which parameters are constrained at each level.

Results

Descriptives

The average response category endorsed by respondents for items 1 through 5 of the SWLS were 5.1, 5.6, 5.6, 5.3, and 4.6, respectively, on a scale of 1–7. Cronbach's alpha was estimated to be 0.91. This is consistent with values found elsewhere in the literature [16, 38].

Common factoring with principal axis extraction and varimax rotation resulted in 74% of variance explained by a single factor.

Dimensionality

CFAs were then used to compare a one-factor (F1) to a two-factor solution (F2). The two-factor model including "past" (last two items) and "present" (first three items) factors yielded better fit than the unconstrained one-factor model (CFI = 0.995 vs. 0.986); RMSEA = 0.065 vs. 0.094). However, the correlation between the factors was

close to unity (r = 0.93), indicating that the two factors could not be easily differentiated. In addition, parsimony was slightly better with the unidimensional model (PNFI: F1 = 0.493, F2 = 0.398).

A modified unidimensional model allowing the residual variance for items 4 and 5 to correlate (F1cov) showed improved fit relative to the baseline model (CFI: F1 = 0.986, F1cov = 0.995; RMSEA: F1 = 0.094, F1cov = 0.065) and identical fit to the two-factor solution (Fig. 1), since the two models are equivalent [39]. This latter model reflects, however, the time dependency in items 4 and 5 more specifically.

A fourth variant suggested by Vautier [34] tested for inter-item correlation when the items were presented to the participants consecutively and successively rather than scattered throughout the questionnaire. The fit of this model was very high (CFI = 0.999; RMSEA = 0.043), but consideration of parsimony indicated that this model should be rejected (PNFI = 0.100, P6FI = 0.067).

Since the correlation between the two factors in the two-factor model was very high (0.93) and minor secondary factors are inherent in most psychological measures [17], the modified single-factor model (Fig. 1) was retained for the subsequent analyses.

¹ Detailed results of the analyses of these two models and the model for inter-item correlation is available from the corresponding author.



Table 3 Standardized factor loadings for all five items of the Satisfaction with Life Scale in a one-factor model with correlation between error terms for items 4 and 5, for the entire sample and for

each subgroup by gender and age. Mean, N, and statistical tests are included (Norwegian health interview survey 2005)

	Entire sample	Gender		Age groups (Age groups (years)						
		Males	Females	16–24	25–44	45–64	65+				
N	4,984	2,369	2,615	623	1,838	1,843	680				
Mean*	26.20	26.17	26.23	26.76	26.29	25.93	27.12				
Item 1	0.88	0.88	0.88	0.85	0.89	0.87	0.86				
Item 2	0.83	0.82	0.84	0.72	0.83	0.85	0.90				
Item 3	0.87	0.87	0.87	0.87	0.88	0.88	0.85				
Item 4	0.78	0.78	0.78	0.67	0.79	0.82	0.79				
Item 5	0.71	0.70	0.73	0.66	0.72	0.73	0.68				
$\chi^2(df)$	89.4 (4)	35.3 (4)	61.4 (4)	16.6 (4)	18.5 (4)	72.7 (4)	49.3 (4)				
CFI	0.995	0.996	0.993	0.992	0.998	0.990	0.981				
RMSEA	0.065	0.057	0.074	0.071	0.044	0.097	0.129				

^{*} Means on a range of 5-35

Table 3 and Fig. 1 shows the factor loadings and fit measures for the modified single-factor model—for the total sample and for the different subgroups. The factor loadings showed basically the same pattern across subgroups and were generally high (>0.70). For the youngest age group, however, factor loadings for items 4 and 5 were estimated to be <0.70 and relatively lower for item 2 (0.72) than observed in the remaining age groups (0.82–0.90). In the oldest age group, the factor loading for item 5 was also <0.70.

Since the responses are based on a 7-point Likert scale, we assumed continuous variables. The analyses were repeated using Bayesian techniques, however, which are recommended for ordinal data. The results from the two estimation techniques were identical principally to the third decimal (Table 4).

To further examine the effect of the non-normality of the data on results obtained using ML estimation, we also reran the analysis (main model, Fig. 1) using asymptotically distribution free (ADF) testing on the original data set (Table 4). Comparing results from ML estimation with ADF testing resulted in very similar estimates of factor loadings and variance, and all parameters were significant using both methods. Model fit using ADF testing was slightly worse when measured by CFI and PNFI and slightly better when measured by RMSEA.

To further examine the effect of non-normality of the data, the ML testing were repeated on data normalized using Tukey's formula. The estimated factor loadings, intercepts, and variance differed as expected from those based on ML estimation. The factor loadings remained, however, significant, but the intercepts were no longer found to be significant. Additionally, tests of model fit resulted in worse fit as measured by CFI and parsimony (Table 4).

Finally, bootstrapping, the recommended analysis technique for non-normal data, confirmed the results obtained with standard ML estimation and indicated significance for factor loadings, intercepts, and variance (Table 4).

A total assessment based on Table 4, thus seems to indicate that ML yields satisfactory results even when accounting for the non-normality and ordinal nature of the data.

Measurement invariance

Gender

The results of the tests for multigroup invariance between genders are given in Table 5. No significant differences in Δchi^2 were found, indicating weak (metric) invariance between the sexes. Both the strong and the strict invariance tests indicated significant differences across men and women. Due to the large sample sizes, tests involving chi^2 can be misleading, however. We therefore used the ΔCFI test which is more appropriate for large sample sizes [33, 40]. The ΔCFI results (Table 5) indicate measurement invariance at the weak, strong, and strict levels between genders. Partial invariance techniques indicated invariance at the strict level.

Model fit as measured by RMSEA improved as more constraints were imposed to the model while fit measured by CFI remained consistently high.

Age groups

Results of the tests for multigroup invariance between age groups are shown in Table 6. All three tests (weak, strong,



Table 4 Non-standardized parameters and fit indices with standard error (SE) for main model (F1cov) when using maximum likelihood (ML), asymptotically distribution free (ADF) testing, normalized data

(Tukey's formula), and bootstrapping techniques (Norwegian health interview survey 2005; N = 4,984)

	ML (SE)	Bayesian analysis (SE)	ADF (SE)	Normalized data ML (SE)	Bootstrapping ^c ML (SE)
λ11 ^a	1.000	1.000	1.000	1.000	1.000
λ21	0.859 (0.011)*	0.859 (0.011)	0.843 (0.014)*	0.921 (0.013)*	0.859 (0.015)*
λ31	0.881 (0.011)*	0.881 (0.011)	0.871 (0.014)*	0.957 (0.012)*	0.881 (0.014)*
λ41	0.850 (0.013)*	0.850 (0.013)	0.849 (0.016)*	0.882 (0.013)*	0.850 (0.016)*
λ51	0.948 (0.016)*	0.969 (0.016)	0.951 (0.017)*	0.844 (0.014)*	0.949 (0.016)*
$\tau 1$	2.935 (0.02)*	2.934 (0.020)	_	-0.017 (0.013)	2.936 (0.02)*
τ2	2.443 (0.019)*	2.443 (0.018)	_	-0.025 (0.013)*	2.443 (0.018)*
τ3	2.355 (0.018)*	2.354 (0.018)	_	-0.026 (0.013)*	2.355 (0.018)*
τ4	2.688 (0.02)*	2.687 (0.019)	_	-0.021 (0.013)	2.689 (0.019)*
τ5	3.377 (0.024)*	3.376 (0.024)	_	-0.011 (0.013)	3.378 (0.024)*
θ 1	0.486 (0.014)*	0.486 (0.015)	0.450 (0.019)*	0.204 (0.006)*	0.486 (0.021)*
θ 2	0.530 (0.014)*	0.530 (0.014)	0.533 (0.021)*	0.255 (0.007)*	0.530 (0.022)*
θ 3	0.392 (0.011)*	0.393 (0.012)	0.386 (0.018)*	0.201 (0.006)*	0.391 (0.019)*
θ 4	0.759 (0.018)*	0.760 (0.018)	0.753 (0.028)*	0.333 (0.008)*	0.757 (0.028)*
θ 5	1.396 (0.031)*	1.400 (0.032)	1.362 (0.042)*	0.404 (0.009)*	1.395 (0.043)*
cov45	0.195 (0.018)*	0.195 (0.017)	0.174 (0.023)*	0.071 (0.006)*	0.195 (0.023)*
α	1.608 (0.042)*	1.610 (0.042)	1.610 (0.045)*	0.650 (0.017)*	1.607 (0.045)*
Chi ² (df) ^b	89.4 (4)*	_	44.4 (4)*	64.3 (4)*	$97.5 \ (0.611) \ (df = 4)^{d}$
CFI	0.995	_	0.978	0.996	
RMSEA	0.065	_	0.048	0.055	
PRatio	0.400	_	0.400	0.267	
PNFI	0.398	_	0.390	0.266	

 $\lambda =$ item factor loading between latent variable (1) and items 1 through 5; $\tau =$ item intercepts for items 1 through 5; $\theta =$ item residuals for items 1 through 5; $\cos \theta =$ covariance between item 4 and 5; $\alpha =$ latent mean; afixed at 1.0; bin 1,000 s; Bootstrapping with 2,000 samples, 95% CI, and significance tested with bias corrected CI; mean chi²; * P < 0.01. Model as described in Fig. 1

strict) indicated non-invariance across age as measured by significant differences in Δchi^2 . The ΔCFI test, however, indicated invariance at the weak level, but not at the strong or strict levels of invariance testing. Partial invariance testing, however, did indicate invariance at the weak level when removing constraints on factor loadings for item 1 (data not shown), as well as better model fit. Further testing of partial invariance at the strong and strict level did not support measurement invariance across age. In conclusion, the finding of invariance at the weak level assures that comparisons can be made as to the relationships between the factors (factor coefficients) across age groups. The results indicate, however, that caution should be used in analyses involving comparison of means between groups.

Discussion

The Satisfaction with Life Scale [3] is perhaps the most widely used measure of well-being worldwide. The

dimensionality of SWLS has been widely discussed, but most studies have been based on specialized sample groups limited in size and biased with respect to gender, age, and relevant socio-demographic parameters. This study aimed to examine the dimensionality of the SWLS in a large and representative sample from Norway (including nearly 5,000 respondents), and to study the robustness of the scale in different subpopulations. This was done by exploring the (1) dimensional structure and (2) measurement invariance across gender and age. No other study has studied dimensionality or subgroup invariance across a continuous age distribution in a comparatively large community sample.

This study also examined the comparability of results from different estimation techniques, including standard ML estimation using raw scores, Bayesian estimation, ADF estimation, and ML estimation using normalized data. The results were consistent regardless of estimation technique, indicating that use of standard ML estimation is satisfactory when studying dimensionality of SWLS when scored on a 7-point Likerts scale.



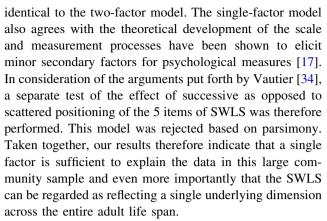
Table 5 Non-standardized parameter estimates and fit indices for measurement invariance models for men and women: baseline (unconstrained), weak (measurement weights), strong (measurement intercept), and strict (measurement residual) (Norwegian health interview survey 2005; $N = 4{,}984$)

Parameter	Basel	ine	Weak	C	Stron	g	Strict		
	M	F	M	F	M	F	M	F	
λ11	1.00	1.00a	1.00	1.00a	1.00	1.00a	1.00	1.00a	
λ21	0.87	0.85	0.86	0.86b	0.86	0.86b	0.86	0.86b	
λ31	0.90	0.87	0.88	0.88b	0.88	0.88b	0.88	0.88b	
λ 41	0.88	0.83	0.85	0.85b	0.85	0.85b	0.85	0.85b	
λ51	0.97	0.93	0.95	0.95b	0.95	0.95b	0.96	0.96b	
$\tau 1$	2.91	2.96	2.91	2.96	2.93	2.93b	2.93	2.93b	
τ2	2.43	2.46	2.43	2.46	2.44	2.44b	2.44	2.44b	
τ3	2.34	2.37	2.34	2.37	2.35	2.35b	2.35	2.35b	
τ4	2.72	2.66	2.72	2.66	2.69	2.69b	2.69	2.69b	
τ5	3.43	3.33	3.43	3.33	3.37	3.37b	3.38	3.38b	
$\theta 1$	0.45	0.51	0.44	0.52	0.44	0.52	0.49	0.49b	
θ 2	0.53	0.53	0.53	0.53	0.53	0.53	0.53	0.53b	
θ 3	0.38	0.40	0.38	0.40	0.38	0.40	0.39	0.39b	
θ 4	0.72	0.79	0.72	0.78	0.73	0.79	0.76	0.76b	
θ 5	1.47	1.32	1.47	1.32	1.47	1.33	1.40	1.40b	
cov45	0.20	0.19	0.20	0.18	0.21	0.18	0.20	0.19	
α	1.48	1.73	1.51	1.69	1.51	1.69	1.51	1.70	
Chi^2 (df)	96.7 (8)		101.1 (12)		125.5 (17)		145.0 (22)		
$\Delta \text{chi}^2 (\Delta df)$	_		4.4 (4)		24.4 (5)*		19.5 (5)*		
CFI	0.995		0.995		0.993		0.992		
RMSEA	0.047		0.039		0.036	5	0.034		

 $\lambda=$ item factor loading between latent variable (1) and items 1 through 5; $\tau=$ item intercepts for items 1 through 5; $\theta=$ item residuals for items 1 through 5; cov covariance between item 4 and 5; $\alpha=$ latent mean. M male; F female; a= fixed at 1.0; b= constrained to equality with first group in the same model. * P<0.01. Model as described in Fig. 1

Dimensions

Our data essentially support a single-factor solution for the SWLS with 74% of variance explained by this single factor. The loadings are on the high side compared to previous studies, and there is a tendency for the last two items to load on a second, less important factor reflecting past accomplishments. This finding is in accordance with several previous reports, but the finding has been interpreted differently across the studies. The correlation between the two factors estimated in this study was very high (r = 0.93), however, and similar to previously reported estimates [14, 18], indicating that the two factors could not be easily differentiated. A post-hoc modification test on the data showed gained fit for the single-factor model when allowing the residual variances for items 4 and 5 to be correlated. This modified single-factor model improved the fit relative to the baseline model and produced fit measures



The two last items obviously share residual variance over and above what is accounted for by the main latent construct. In the present study, for example, the single-factor model fitted the data better for men than for women, and gave better fit for the two youngest age groups than for the two older age groups. Despite the fact that the last items, perhaps due to their reference to past accomplishments rather than current conditions, appear to involve a somewhat different cognitive search, the overall results support a single dimension in all the subgroups investigated.

Invariance

Between genders

No gender differences were observed at the level of factor loadings, indicating metric or "weak" invariance across gender in the total sample. This attests that the latent variable is related to the items in the same way for men and women. Further constraints equating the intercepts (strong invariance) and the residuals (strict invariance) resulted in significantly reduced fit in terms of the chi² test. Analyses based on large samples may result in high chi² values, however, and increased risk for rejecting good models. In the current study, additional fit indices were either improved (RMSEA) or only slightly decreased (CFI) when adding further constraints (equating intercepts and residuals) to the baseline model. This suggests that the intercepts and residuals may be fixed to equality in men and women, thereby supporting the assumption of strict invariance across gender. This implies that group means on the latent variable as well as analyses involving correlations with the latent variable are comparable across gender. This finding corroborates a number of previously described findings [14, 18, 21, 41], although Atienza et al. [25] in Spanish junior high students did not agree.

Between age groups

People differ in what they require for a satisfying life, and different dimensions of well-being seem to be meaningful



Table 6 Non-standardized parameter estimates and fit indices for measurement invariance models for the subgroups of age: baseline (unconstrained), weak (measurement weights), strong (measurement

intercept), and strict (measurement residual) (Norwegian health interview survey 2005; N = 4,984)

Para-meter	Baselin	ie			Weak				Strong				Strict			
	16–24	25–44	45–64	65+	16–24	25–44	45–64	65+	16–24	25–44	45–64	65+	16–24	25–44	45–64	65+
λ11	1.00	1.00a	1.00a	1.00a	1.00a	1.00a	1.00a	1.00a	1.00a	1.00a	1.00a	1.00a	1.00a	1.00a	1.00a	1.00a
λ21	0.65	0.85	0.91	0.95	0.87	0.87b	0.87b	0.87b	0.87	0.87b	0.87b	0.87b	0.87	0.87b	0.87b	0.87b
λ31	0.87	0.88	0.91	0.84	0.88	0.88b	0.88b	0.88b	0.88	0.88b	0.88b	0.88b	0.88	0.88b	0.88b	0.88b
λ41	0.81	0.87	0.85	0.85	0.85	0.85b	0.85b	0.85b	0.85	0.85b	0.85b	0.85b	0.85	0.85b	0.85b	0.85b
λ51	0.88	0.97	0.94	0.94	0.95	0.95b	0.95b	0.95b	0.95	0.95b	0.95b	0.95b	0.95	0.95b	0.95b	0.95b
τ1	3.07	2.90	2.99	2.75	3.07	2.90	2.99	2.75	2.93	2.93b	2.93b	2.93b	2.93	2.93b	2.93b	2.93b
τ2	2.33	2.41	2.54	2.37	2.33	2.41	2.54	2.37	2.44	2.44b	2.44b	2.44b	2.44	2.44b	2.44b	2.44b
τ3	2.26	2.34	2.45	2.24	2.26	2.34	2.45	2.24	2.35	2.35b	2.35b	2.35b	2.35	2.35b	2.35b	2.35b
τ4	3.12	2.68	2.64	2.44	3.12	2.68	2.54	2.44	2.65	2.65b	2.65b	2.65b	2.68	2.68b	2.68b	2.68b
τ5	3.46	3.36	3.45	3.08	3.46	3.38	3.45	3.08	3.38	3.38b	3.38b	3.38b	3.39	3.39b	3.39b	3.39b
$\theta 1$	0.46	0.46	0.52	0.48	0.48	0.48	0.51	0.48	0.48	0.48	0.51	0.48	0.49	0.49b	0.49b	0.49b
θ 2	0.66	0.53	0.53	0.29	0.65	0.53	0.54	0.33	0.67	0.53	0.54	0.33	0.53	0.53b	0.53b	0.53b
θ 3	0.45	0.37	0.39	0.36	0.46	0.36	0.40	0.35	0.49	0.36	0.40	0.35	0.39	0.39b	0.39b	0.39b
θ 4	1.29	0.71	0.60	0.60	1.31	0.72	0.59	0.58	1.53	0.72	0.60	0.59	0.76	0.76b	0.76b	0.76b
θ 5	1.67	1.40	1.31	1.42	1.66	1.41	1.30	1.39	1.67	1.41	1.30	1.42	1.42	1.42b	1.42b	1.42b
cov45	0.16	0.15	0.20	0.31	0.17	0.16	0.19	0.29	0.21	0.16	0.19	0.31	0.04	0.17	0.29	0.38
α	1.64	1.58	1.69	1.39	1.43	1.60	1.74	1.42	1.43	1.57	1.75	1.45	1.55	1.55	1.76	1.41
Chi ² (df)	163.7 (18)				244.0 (30)			453.3 (45)			744.7 (59)					
$\Delta \text{Chi}^2 (\Delta df)$	_				80.3 (12)*			209.3 (15)*			291.3 (14)*					
CFI	0.991				0.987			0.976			0.959					
RMSEA	0.040				0.038			0.043				0.048				

 $\lambda =$ item factor loading between latent variable (1) and items 1 through 5; $\tau =$ item intercepts for items 1 through 5; $\theta =$ item residuals for items

to people of varying age. Different ages and life circumstances may cause systematic shifts in how people evaluate their life situation. Oishi and colleagues [42] have, for example, proposed a "value as moderator model" which predicts that as individual's age, changes in values lead to changes in the determinants of their life satisfaction. Ryff [43] found middle-aged individuals to stress the importance of self-confidence, self-acceptance, job, and career issues, whereas older respondents focus more on health issues. In the present analyses, we find that the SWLS is sensitive to age at the strong and strict levels indicating that life satisfaction as measured by the SWLS does not have the same meaning across the life span.

The results from our current study also indicate that the underlying construct is not fully comparable across the age groups. Our finding is in accordance with previous reports, [14, 26] although others [13, 22–24] found invariance among age groups. These studies were based on far more age homogenous samples (mainly students) and were therefore not able to examine invariance across the entire adult life span. By including respondents from 15 to 79 years, the present study shows that intercepts and

residuals vary across the adult life span. Manifest and latent SWLS scores are therefore only partially comparable across age groups. This important finding may partly be due to different adaptation strategies, cohort effects, socialization practises, age specific circumstances influencing interpretations, and conceptualizations of the items on the SWLS as well as increased individual differences in physical health and mobility [44]. Older individuals have been shown to make more global evaluations, be more present oriented and to stress interpersonal aspects, whereas younger people focus more on intrapersonal and specific evaluations [44]. The temporal framing of the items may also be important. The SWLS scale incorporates items referring to both current conditions and past accomplishments, and the time perspectives are likely to vary across age groups [16].

Strengths and limitations of this study

Our study has two major advantages: (1) the relatively large sample size and (2) respondents representing the



entire country—all levels of society and a large age span. The shortcomings are related to a moderate response rate, perhaps leading to a less representative sample. When compared to population statistics, women and the age group from 45 to 64 years are overrepresented in this study. The eldest population group (>65 years) consists of fewer individuals, and only includes those living at home, and not in institutions. Likewise, immigrants with a non-Western ethnic background are clearly underrepresented in this material. In addition, using the AMOS analytical package did not allow robust ML testing (Satorra-Bentler scaled statistic) that would have strengthened the analysis when using ordinal non-normal data.

Conclusions

The overall results indicate that the one-factor latent structure of the SWLS is valid in the Norwegian data and that comparing men and women is feasible whereas some caution should be exerted when comparing age groups.

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