

# Student's Academic Efficacy or Inefficacy? An Example on How to Evaluate the Psychometric Properties of a Measuring Instrument and Evaluate the Effects of Item Wording

João Maroco<sup>1\*</sup>, Ana Lúcia Maroco<sup>1</sup>, Juliana Alvares Duarte Bonini Campos<sup>2\*</sup>

<sup>1</sup>Unidade de Investigação em Psicologia e Saúde, Instituto Superior de Psicologia Aplicada (ISPA), Lisboa, Portugal

<sup>2</sup>Departamento de Odontologia Social, Faculdade de Odontologia de Araraquara, UNESP, Universidade Estadual Paulista, Araraquara, Brazil

Email: \*[jpmaroco@ispa.pt](mailto:jpmaroco@ispa.pt), \*[jucampos@fcar.unesp.br](mailto:jucampos@fcar.unesp.br)

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## Abstract

This study evaluated the effect of item inversion on the construct validity and reliability of psychometric scales and proposed a theoretical framework for the evaluation of the psychometric properties of data gathered with psychometric instruments. To this propose, we used the Maslach Burnout Inventory, which is the most used psychometric inventory to measure burnout in different professional context (Students, Teachers, Police, Doctors, Nurses, etc...). The version of the MBI used was the MBI-Student Survey (MBI-SS). This inventory is composed of three key dimensions: Exhaustion, Cynicism and Professional Efficacy. The two first dimensions—which have positive formulated items—are moderate to strong positive correlated, and show moderate to strong negative correlations with the 3rd dimension—which has negative formulated items. We tested the hypothesis that, in college students, formulating the 3rd dimension of burnout as Inefficacy (reverting the negatively worded items in the Efficacy dimension) improves the correlation of the 3rd dimension with the other two dimensions, improves its internal consistency, and the overall MBI-SS' construct validity and reliability. Confirmatory factor analysis results, estimated by Maximum Likelihood, revealed adequate factorial fit for both forms of the MBI-SS (with Efficacy) vs. the MBI-SSi (with Inefficacy). Also both forms showed adequate convergent and discriminant related validity. However, reliability and convergent validity were higher for the MBI-SSi. There

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\*Corresponding authors.

were also stronger (positive) correlations between the 3 factors in MBI-SSi than the ones observed in MBI-SS. Results show that positively rewording of the 3rd dimension of the MBI-SS improves its validity and reliability. We therefore propose that the 3rd dimension of the MBI-SS should be named Professional Inefficacy and its items should be positively worded.

## Keywords

**Burnout, Maslach Burnout Inventory—Student Survey, Validity, Reliability, Item Inversion**

## 1. Introduction

Measuring can be defined as a process of encoding characteristics or attributes of objects or people. If it is conducted properly researchers must take into account the objectives, the reliability and validity of the measure which is a function not only of the measuring instrument but also depends on the context, the variables to be assessed and the sample/target population characteristics [1]. Besides the acquisition of information about the characteristics/attributes investigated, the measuring act requires that the researcher adopts a responsible and ethical position, since every measure consists of true value and random error (Classical Theory of Measurement), and no psychometric measuring instrument *per se* can be considered valid or reliable. Therefore information about the quality of the data should be presented every time a psychometric instrument is used to measure some construct.

To assess the quality of information, the validity and reliability of the data must be evaluated. This process becomes especially challenging when the measurement of latent variables that are not directly observable or measured against a “gold-standard” is involved [2]. That is the case of several medical and psychiatric variables (e.g., stress, depression, anxiety, cognitive complains, pain, etc...). This type of variables required the construction of scales and measuring instruments that are solid grounded in theoretical models which relate the construct or latent variable to its observed manifestations in a proper valid and reliable way [3]. Psychometric scales are the most common measuring instruments used to assess latent variables in the Social and Health Sciences. Even so, the reporting of the psychometric properties of the data gathered with these instruments has been neglected in most studies. This fact may compromise the data analysis and conclusions reached [4]-[6]. It also need to be highlighted that, often, in the literature it is stated that the validity and reliability are properties of the measuring instrument which is a misnomer, since these are properties of the data gathered with the instrument when applied to a sample and in specific context and not of the instrument *per se* [7]. So, whenever these contexts are changed the validity and reliability of the data need to be re-estimated and therefore establishing the psychometric properties of an instrument is a time consuming, and always necessary dynamic process. The lack of familiarity of both researchers and practioners in the specific areas of statistics/psychometrics required to estimate validity and reliability of the data can be one of the causes of the problems reported.

Several statistical methodological approaches have been proposed to evaluate the psychometric properties of data. Some are based on simple bivariate correlations, while others are supported by complex Item Response Theory (IRT) or Structural Equation Modelling (SEM). Nowadays, the most usual method to estimate the psychometric properties of data gathered by a measuring instrument resort to SEM. [3] [8]-[12]. Despite the several different technics that can be used, the simple bivariate correlational analysis has been losing interest as compared to more advanced techniques, like SEM, since as pointed out by Kane [13] and Pais-Ribeiro [1], psychometric evaluation requires the development of a strong and explicit theory about the underlying concept/construct to be studied.

This study aims to present the SEM based steps of analysis that should be used to assess the validity and reliability of data gathered with a psychometric scale using the theoretical validity and reliability framework proposed by Anastasi *et al.* [5]. We also evaluated the effects of negative vs. positive item wording in the psychometric properties of data gathered with scales that use both formats. For that end, we used the Maslach Burnout Inventory—Student Survey. Student Burnout, as assessed by the Maslach Burnout Inventory-Student Survey (MBI-SS), was originally proposed as a tri-factorial syndrome associated with high Emotional Exhaustion (EX), high Cynicism/disbelief (CY) in one’s college studies and low Professional Efficacy (EF) [14].

Our results suggest that in college students Inefficacy is better correlated with Exhaustion and Cynicism than

Efficacy when estimation the correlation of Burnout's 3 sub-scales. Our data support the observation that psychometric properties not only dependent on the instrument used, the context where it's used, but also on how the items are formulated.

## 2. Methods

### 2.1. Sample Design and Participants

A cross-sectional study was performed. The minimum required sample size was estimated accordingly to Hair *et al.* [15] which recommends the use of at least 5 subjects per parameter to be estimated in the model, for  $\alpha = 5\%$  and power of 80%. The study included 268 students from Brazilian ( $n = 179, 67\%$ ) and Portuguese ( $n = 89, 33\%$ ) universities. The average age was 21.0 (SD = 5.4) year and 77% were female. Of the students, 82% were enrolled in public universities, 63% was majoring in health sciences, 33% in social sciences and humanities and 4% in biological sciences.

### 2.2. Measuring Instrument

The Portuguese Maslach Burnout Inventory—Student Survey (MBI-SS) were transculturally validated for both Brazilian and Portuguese students by Campos and Maroco [16]. The Portuguese version of the MBI-SS is composed by 15 ordinal items (0-never to 6-always) defining 3 sub-scales (Emotional Exhaustion—EX, Cynicism—CY and Professional Efficacy—EF). In its original formulation [14] the Efficacy dimension has its items positively worded (e.g., item 3—In my opinion, I am a good student) while Emotional Exhaustion and Cynicism are composed by negatively worded items (e.g., item 1—I feel emotionally drained by my studies and item 7—I have become less enthusiastic about my studies, respectively). In our study, two forms of the MBI-SS were used. The one with the Efficacy as the 3rd dimension with original formulated items and a new MBI-SS where the items in the 3rd dimension were negatively worded to give raise to the Inefficacy (INEF) dimension as proposed also by Bresó *et al.* [17]. The Portuguese version was translated independently by the authors of this study and retro translated by a bilingual psychologist who had no knowledge of the original versions of the MBI-SS. The MBI with the Inefficacy dimension was named MBI-SSi as opposed to the common MBI-SS (with Efficacy as the 3rd dimension).

### 2.3. Procedures

Participants completed the instruments in the classroom, on a schedule agreed beforehand. Students were informed that their participation was voluntary, that no personnel information that could be used to identify them was required and that they could drop out of the study at any time without consequences. The MBI-SS and MBI-SSi instruments were completed separately, with one week interval between fills. The order of presentation of the MBI-SS and MBI-SSi was randomly assigned. The present study followed the ethical precepts dictated by Resolution 196/96 of the Brazilian National Health Council and the UIPES-ISPA's Ethics committee.

### 2.4. Data Analysis

While recognizing other frameworks for Psychometric data analysis (e.g. Nunnally and Bernstein [18], American Educational Research Association (AERA), American Psychological Association (APA) and National Council on Measurement in Education (NCME) [19], etc....), the psychometric properties analysis framework in this study follow the proposal presented by Anastasi *et al.* [5], Hair *et al.* [15], Kline [20] and Maroco [11] as described next.

### 2.5. Distributional Properties

Distributional properties of the items from the MBI-SS and MBI-SSi were assessed by Skewness (Sk) and Kurtosis (Ku) measures. Absolute values of Sk and Ku larger than 7 and 3 respectively [11] [20] were indicative of significant deviation from normality recommending against its use in further psychometric analysis since they deviate strongly from the normal distribution, which is expected if the item is a manifestation of a population's construct. Outliers were diagnosed with the Squared Mahalanobis distance. Values greater than 4 were assumed to be multivariate outliers [11].

## 2.6. Construct Related Validity

Construct related validity was assessed by means of factorial validity, convergent and discriminant validity as described next.

## 2.7. Factorial Related Validity

To access the extent in which the items are good manifestations of their theoretical constructs, a confirmatory factor analysis (CFA) was conducted. CFA is equivalent to the measuring model of SEM. It lays out the linear structural relationships between a theoretical latent construct and its observed manifestations (for a review on SEM, see e.g. Bollen, 1989 or Maroco, 2014). The maximum likelihood estimation method was used on the Pearson correlation matrix. To assess the goodness of fit the following indices and reference values [11] [20]-[22] were used:  $\chi^2/df$  (chi-square over degrees of freedom; should be smaller than 3 - 5), CFI (*comparative fit index*), TLI (*Tucker-Lewis index*), GFI (*goodness of fit index*) which should all be greater than 0.95 and RMSEA (*root mean square error of approximation*) which should be smaller than 0.08. At least one index from the 3 main indices families (absolute, relative and population discrepancy was used).

Modification indices, estimated through the Lagrange multipliers (LM) were used to investigate possible correlations between measurement errors. LM greater than 11 ( $p < 0.001$ ) were assumed to indicate significant correlations which could be added to the model if theoretically justifiable [11].

## 2.8. Convergent Related Validity

The factors' convergent related validity assess if the items which are manifestation of the scale converge on that scale or factor. If that is the case, then the scale explains a large percentage of the items' variance. Accordingly to Fornell and Larcker [23] the Average Variance Extracted (AVE) as defined by (1) can be used to evaluate convergent validity. AVE greater than 0.5 is indicative of the factors' convergent validity [15]. Accordingly to Maroco [11] the AVE for a  $j$  factor can be estimated by its  $k$  items' standardized factor weights ( $\lambda$ ) as:

$$AVE_j = \frac{\sum_{i=1}^k \lambda_{ij}^2}{\sum_{i=1}^k \lambda_{ij}^2 + \sum_{i=1}^k \varepsilon_{ij}} \quad (1)$$

Using the same  $\lambda$ , the Construct reliability can be estimated by:

$$CR_j = \frac{\left( \sum_{i=1}^k \lambda_{ij} \right)^2}{\left( \sum_{i=1}^k \lambda_{ij} \right)^2 + \sum_{i=1}^k \varepsilon_{ij}} \quad (2)$$

where can be estimated as  $\varepsilon_{ij} \approx 1 - \lambda_{ij}^2$ .

Accordingly to Hair et al. [15]  $AVE_j \geq 0.5$  and  $CR_j \geq 0.7$  are indicative of construct convergent related validity. For exploratory research, values below 0.7 are generally acceptable [15].

## 2.9. Discriminant Validity

Discriminant related validity assess if different factors evaluate different constructs. Accordingly to Fornell and Larcker [23], if the AVE from every two constructs is larger than the squared Pearson Correlation ( $\rho_{ij}^2$ ) between the constructs, than the variance explained by one of the constructs as predictor of the other is smaller than the variance that the construct explains of its manifest items. Therefore, two constructs  $i$  and  $j$  show discriminant related validity if  $AVE_i$  and  $AVE_j$  are both larger that  $\rho_{ij}^2$ .

## 2.10. Reliability

The reliability of a construct can be assessed by three different properties: Stability, Equivalence and Internal consistency [24]. The Stability evaluates how well the measure retains its properties along time. This is general-

ly observed by strong test-retest correlations, interclass correlation coefficients or Cohen's K. Equivalence refers to strong correlations with similar constructs, or the two split-halves of the same construct, evaluated by means of Pearson correlations. The internal consistency is by far the most used measure of reliability. It evaluates how consistent a set of items are as manifestations of a given construct, that is how much of construct true variance is captured by the items (scale) used to measure it. Cronbach's alpha is the most used estimate for internal consistency [24]. For scales with different type of items the standardized Cronbach  $\alpha$  is given by

$$\alpha = \frac{k\bar{r}}{1 + (k-1)\bar{r}} \quad (3)$$

where  $k$  is the number of items in the scale and  $\bar{r}$  is the average correlation between the  $k$  items. When psychometric inventories have more than one scale it is recommend to calculate the Stratified Cronbach's  $\alpha$  [24]:

$$\alpha_{\text{estr}} = 1 - \frac{\sum_{i=1}^f s_i^2 (1 - \alpha_i)}{s_T^2} \quad (4)$$

where  $s_i^2$  is the variance of items in the  $i$  factor,  $s_T^2$  is the variance of the sum of the items and  $\alpha_i$  is the alpha estimate for factor  $i$ .

## 2.11. Second Order Hierarchical Factors

When the first order factors show moderate to strong correlation and a high hierarchical factor is theoretically justifiable, a second order factor can be tested by CFA. The significance of the standardized structural weights ( $\beta$ ) between the 2nd order factor and the 1st order factors can be used to evaluate the relative contribution of the first order factors as facets of the higher order factor.

## 2.12. Statistical Analysis

Statistical analyses were performed using IBM SPSS Statistics (v.22, SPSS An IMB Company, Chicago, IL) and AMOS 22.0 (SPSS Inc., Chicago, IL) software.

# 3. Results

## 3.1. Distributional Properties

Descriptive statistics and asymmetry and kurtosis statistics of the items from MBI-SS and MBI-SSi are given in **Table 1**. None of the items from either MBI-SS or MBI-SSi showed descriptive and distribution statistics which showed considerable deviation from the normal distribution. Its use in further psychometric data analysis was therefore recommended.

## 3.2. Factorial Related Validity

**Figure 1** give the structural CFA models for both MBI-SS and MBI-SSi fitted to the study sample. Both the MBI-SS ( $\chi^2/df = 2.267$ ; GFI = 0.910; CFI = 0.943; TLI = 0.930; RMSEA = 0.069) and the MBI-SSi ( $\chi^2/df = 3,017$ ; GFI = 0.883; CFI = 0.936; TLI = 0.922; RMSEA = 0.087) show acceptable goodness-of-fit to the items' variance-covariance matrixes. On general, items from the INEF dimension have larger standardized factor loadings that their similar items from EF.

## 3.3. Convergent and Discriminant Related Validity and Reliability

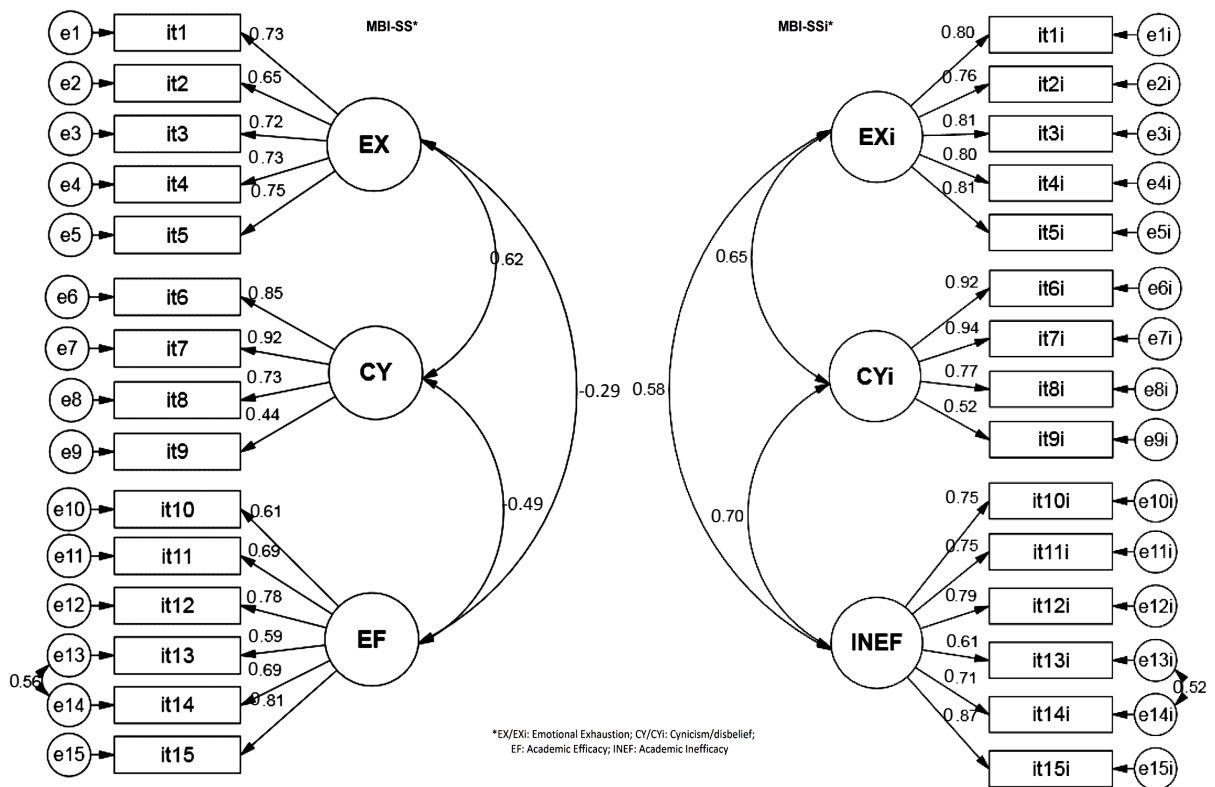
**Table 2** gives the estimated AVE, CR squared Pearson correlations and Cronbach's  $\alpha$  for both MBI-SS and MBI-SSi.

Accordingly to Fornell and Larcker [23] all factors show both adequate convergent and discriminant validity. Accordingly to Cronbach  $\alpha$  all factors show high reliability.

**Figure 2** shows the structural analysis of MBI-SS and MBI-SSi with Burnout as a 2nd order factor. Acceptable

**Table 1.** Descriptive statistics, Skewness and Kurtosis for the MBI-SS items and MBI-SSi (Brasil/Portugal, 2013) (Note: Short forms of items are given just for content illustration proposes. Full items are copyrighted by Mind Garden Inc.)

Item	MBI-SS						MBI-SSi					
	Mean	Median	Mode	SD	SK	Ku	Mean	Median	Mode	SD	SK	Ku
1. Feel emotionally drained	2.94	3	3	1.43	-0.10	-0.62	3.07	3	3	1.60	0.06	-0.80
2. Feel used up at the end of a day	3.33	3	4	1.49	-0.10	-0.93	3.38	3	4	1.56	-0.08	-0.83
3. Feel tired getting up in the morning	2.91	3	1	1.73	0.20	-1.07	2.90	3	2	1.74	0.18	-0.93
4. Studying or attending a class is strain	1.84	2	1	1.39	0.57	-0.26	2.01	2	1	1.50	0.54	-0.48
5. Burned out from studies	2.34	2	2	1.52	0.40	-0.49	2.66	3	3	1.63	0.29	-0.70
6. Less interested in studies	1.02	0	0	1.45	1.58	1.69	1.34	1	0	1.56	1.32	1.08
7. Less enthusiastic about studies	1.10	1	0	1.42	1.39	1.10	1.41	1	0	1.54	1.15	0.76
8. More cynical about usefulness of studies.	1.51	1	0	1.51	0.97	0.20	1.76	1	0	1.66	0.88	-0.03
9. Doubt the significance of studies	1.94	2	0	1.83	0.75	-0.44	2.07	2	0	1.76	0.55	-0.73
10. Effectively solve problems in studies	4.16	4	5	1.37	-0.66	-0.27	-	-	-	-	-	-
11. Make effective contribution to classes	3.57	4	3	1.56	-0.25	-0.60	-	-	-	-	-	-
12. I am a good student	3.95	4	5	1.41	-0.50	-0.41	-	-	-	-	-	-
13. Feel stimulated achieving study goals	4.99	6	6	1.44	-1.76	2.70	-	-	-	-	-	-
14. Learned many interesting things	4.68	5	6	1.54	-1.34	1.19	-	-	-	-	-	-
15. During class confident of effectiveness	3.74	4	5	1.38	-0.46	-0.31	-	-	-	-	-	-
10i. I cannot effectively solve problems in studies	-	-	-	-	-	-	2.29	2	2	1.38	0.54	-0.31
11i. I don't make effective contribution to classes	-	-	-	-	-	-	2.29	2	2	1.60	0.44	-0.49
12i. I am a poor student	-	-	-	-	-	-	1.78	1	1	1.65	0.87	-0.13
13i. I don't feel stimulated achieving study goals	-	-	-	-	-	-	1.12	0	0	1.68	1.67	1.85
14i. I have not learned many interesting things	-	-	-	-	-	-	1.26	1	0	1.50	1.14	0.46
15i. During class I am not confident or feel I am effective	-	-	-	-	-	-	1.78	1	1	1.53	0.78	-0.04



**Figure 1.** Structural CFA models for MBI-SS and MBI-SSi. Values are the standardized factor weight ( $\lambda$ ) and correlations between factors  $\rho$ .

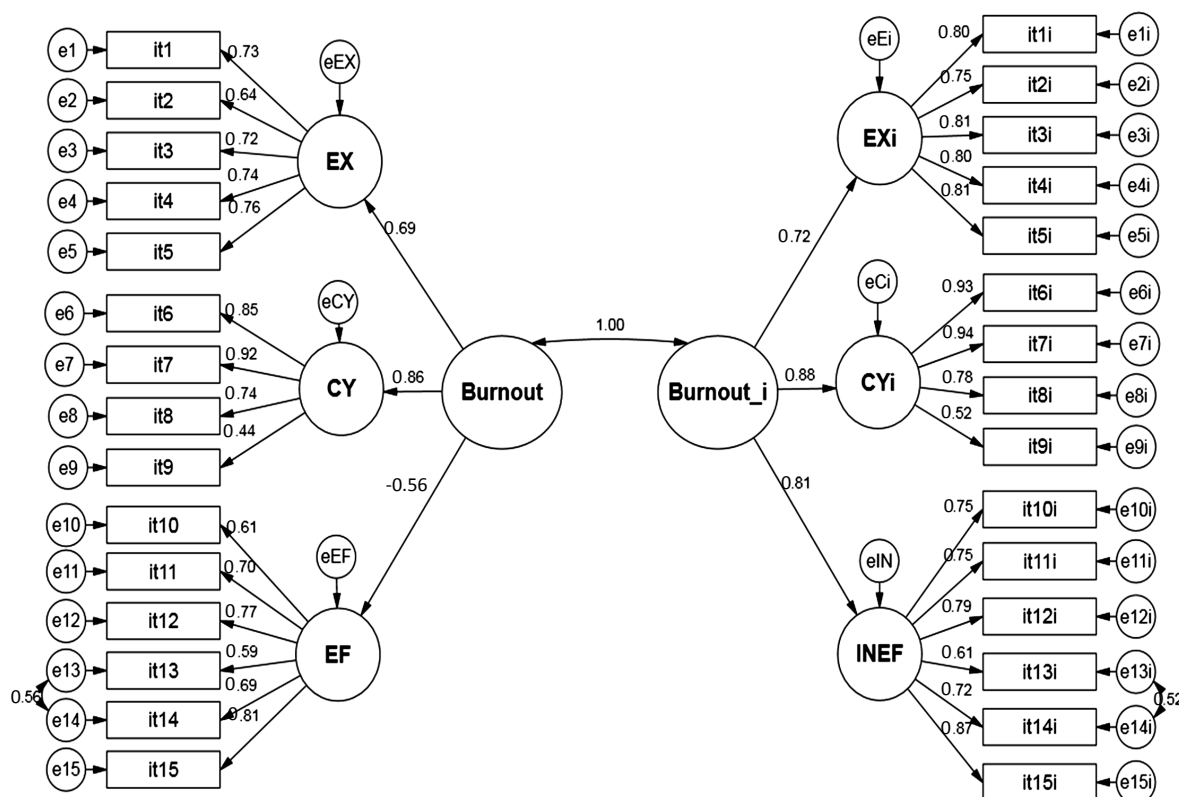
**Table 2.** Average variance extracted (AVE), composite reliability (CR), coefficient of determination ( $\rho^2$ ) and Cronbach's alpha coefficient ( $\alpha$ ) for both MBI-SS and MBI-SSi.

Instrument	AVE	CR	$\rho^2$	$\alpha$
<b>MBI-SS</b>	-	-	-	0.98
Emotional Exhaustion (EX)	0.51	0.84	-	0.84
Cynicism (CY)	0.57	0.83	-	0.81
Efficacy (EF)	0.50	0.85	0.08 - 0.38	0.86
<b>MBI-SSi</b>	-	-	-	0.99
Emotional Exhaustion (EXi)	0.63	0.90	-	0.90
Cynicism (CYi)	0.65	0.88	-	0.86
Inefficacy (INEF)	0.56	0.88	0.34 - 0.49	0.89

goodness-of-fit was observed for the global model ( $\chi^2/df = 3.327$ ; GFI = 0.758; CFI = 0.833; TLI = 0.816; RMSEA = 0.093)

It is worthwhile to notice the perfect correlation between the 2nd order Burnout factor defined by both MBI-SS and MBI-SSi. However, the standardized structural weight of Burnout to INEF is much larger and positive ( $\beta = 0.81$ ;  $p < 0.001$ ) than the similar weight to EF which was smaller and negative ( $\beta = -0.56$ ;  $p < 0.001$ ). That is rewording the items of the 3rd Burnout dimension increased both first order loadings and second order structural weights between the INEF and its items, and Burnout and INEF respectively.





**Figure 2.** Second order hierarchical model of Burnout reflected both in EX, CY and EF as well as EX<sub>i</sub>, CY<sub>i</sub> and EX<sub>i</sub> fitted to both Brazilian and Portuguese college students.

#### 4. Discussion

Given the ubiquitous use of latent variables in the social and health sciences, the assessment of the psychometric properties of the measuring instrument that operationalizes these latent variables or constructs is fundamental when these instruments are used with new data samples, to accurately estimate the data's validity and reliability. Thus, in this paper we suggest a series of analysis required to estimate the construct validity and reliability using structural equation modeling. We exemplify this analysis by studying the psychometric properties of the Maslach Burnout Inventory applied to Brazilians and Portuguese university students considering two theoretical proposals. The first used the original MBI-SS formulation with about 2/3 of the items worded negatively and 1/3 worded positively; the second version (MBI-SSi) used all the items with the same, negative, wording.

The choice of structural equation analysis to evaluate the psychometric characteristics of the instrument and the data gathered with it was guided by the fact that this technique is an extension of generalized linear models that considers measurement errors associated with the latent variables under study. It also allows for the fit of theoretical models established *a priori* to different samples and contexts involving both latent and manifest variables [11]. This study come in due time with recent publications in the Health and Social Sciences that point out the need to review the psychometric properties of the instruments used in research, which must be submitted jointly to the data psychometric evaluation, since results gathered with this type of instruments can only lead to relevant conclusions if the instrument produces reliable and valid data in the study sample. Unfortunately, this recommendation is usually neglected in investigations in the health sciences or performed superficially, thus compromising the quality of the results and conclusions presented [25].

The analysis of the type of different items' wording, in the same instrument, revealed that, as far as the Burnout construct as, estimated by using the MBI-SS, Efficacy is a lesser manifestation of Burnout than Inefficacy (standardized regression weights were  $-0.56$  vs.  $0.81$  respectively) (Figure 2). INEF also correlates stronger with EX and CY than EF does. Positive wording of EF items may cause an artifact in measuring burnout due to the fact that 2/3 of the items are worded negatively and 1/3 are worded positively. Halo effects may explain this



artifact; but also item's perception due to positive wording may be associated with a positive emotional state rather than a negative perceived burnout condition. This fact may also be explained by stereotypes regarding the items wording and patterns associated in response to a previous observed pattern. That is, students, started answer in one direction and failed to perceive that the item's direction was changed along the scale leading to misleading answers [26] [27]. Reliability and validity of the MBI-SSi was also slightly higher than of the MBI-SS. Similar results were reported by Bresó *et al.* [17] with Dutch and Spanish students. These results and results in our study show that when items are worded in the same conceptual direction better convergent validity and reliability was observed.

The tri-factorial structure of the different formats of the MBI has been criticized in aid-related professions. Therefore, other burnout inventories (e.g., OLBI, CBI,...) have proposed a bi-factorial conceptualization of burnout constituted only by Exhaustion and Cynicism. Negative correlations of Exhaustion and Cynicism with Efficacy have led the criticism to the 3-factor structure of the MBI and some scholars have proposed that low efficacy is a consequence and not a dimension of Burnout [28] [29]. More recently, Bresó *et al.* [17] and Schaufeli and Salanova [30] challenged this 2-factor structure conceptualization and proposed that Inefficacy (INEF), assessed by negatively rephrasing the positively worded MBI-GS and MBI-SS efficacy items, is the 3rd burnout dimension. In a sample of Dutch and Spanish students they show that the absolute correlations of EX and CY are higher with INEF than with EF. Simbula and Guglielmi [31] also evaluated the effect of reverting the EF items in teachers and concluded that INEF beliefs relate more strongly to the other burnout dimensions (EX and CY) than EF. Taken together, our observations and data from these authors favour and recommend the use of the inefficacy scale as the "third dimension" of burnout, rather than the efficacy scale. Thus, considering the underlying burnout construct theory and the analysis of construct related validity and reliability of the MBI-SS evaluated on both the original format as well as in new format where the items from Efficacy were reverted to produce a 3rd dimension we termed Inefficacy format. Respecting the theoretical framework of Burnout by Schaufeli, Leiter and Maslach [32], by rewording 1/3 of the items it is possible to improve the psychometric properties of the MBI.

This study aimed also at setting a layout for psychometric data evaluation, paying due attention to item format and wording, for statisticians who have no or little previous experience with psychometric evaluation, especially in the health sciences. A strong theoretical background support regarding the construct is recommended alongside with statistical techniques, like SEM, which can easily produce results required to attest psychometric data validity and reliability.

## 5. Conclusion

Item wording affects the psychometric properties of data gathered with instruments with items formulated both with positive and negative wording. For example, for the MBI-SS, formulating the 3rd dimension of burnout as inefficacy improves the MBI-SS psychometric properties, namely construct reliability and validity, in a sample of Portuguese speaking students. Psychometric data properties, which are a characteristic of the data collected with a given psychometric measurement instrument, can easily be evaluated by SEM following the theoretical framework described in this paper.

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