

*Light Exposure Behavior Assessment (LEBA): Development of a novel instrument to capture light exposure-related behaviours*

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## Abstract

One or two sentences providing a **basic introduction** to the field, comprehensible to a scientist in any discipline.

Two to three sentences of **more detailed background**, comprehensible to scientists in related disciplines.

One sentence clearly stating the **general problem** being addressed by this particular study.

One sentence summarizing the main result (with the words “**here we show**” or their equivalent).

Two or three sentences explaining what the **main result** reveals in direct comparison to what was thought to be the case previously, or how the main result adds to previous knowledge.

One or two sentences to put the results into a more **general context**.

Two or three sentences to provide a **broader perspective**, readily comprehensible to a scientist in any discipline.

*Keywords:* keywords

Word count: X

*Light Exposure Behavior Assessment (LEBA):* Development of a novel instrument to capture light exposure-related behaviours

## Introduction

- Light exposure is important
- Light exposure Behavior is important
- Table: Overview Existing Related Scales: items in total / items on light exposure (behaviour)
- Existing Scales: Review them in text
- None of these do light exposure behavior.

Name	Author
Visual Light Sensitivity Questionnaire-8	[@verriotto2017new]
Office Light Survey	[@eklund1996developme]
Harvard Light Exposure Assessment Questionnaire	[@bajaj2011validation]
Hospital Lighting Survey	[@dianat2013objective]
Morningness-Eveningness Questionnaire	[@horne1976self]
Munich Chronotype Questionnaire (MCTQ)	[@roenneberg2003life]
Assessment of Sleep Environment	[@Olivier.2016]   13
The Pittsburgh Sleep Quality Index (PSQI)	[@buysse1989pittsburgh]
Self-Rating of Biological Rhythm Disorder for Adolescents (SBRDA)	[@Xie.2021]   29
Photosensitivity Assessment Questionnaire (PAQ)	[@Wu.2017]   16

## Methods

### Ethical approval

The cantonal ethics commission (Ethikkommission Nordwest- und Zentralschweiz, project ID Req-2021-00488) reviewed this project and issued an official clarification of responsibility (full document see Suppl. Fig X in appendix) stating: “The research project does not fall under the scope of the Human Research Act, because your project is using only anonymised data. An authorisation from the ethics committee is therefore not required and the EKNZ is not responsible for its review.”

### Data Availability

### Survey characteristics

Data was collected in a quantitative cross-sectional approach via a fully anonymous online survey hosted on REDCap (Harris et al., 2019, 2009) by way of the University of Basel sciCORE. Participants were recruited via the website of a Comic co-released with the survey (Weinzaepflen & Spitschan, 2021), social media (i.e., LinkedIn, Twitter, Facebook), mailing lists, word of mouth, the investigators’ personal contacts, and supported by distribution of the survey link via f.lux software (F.lux Software LLC, 2021).

Completing the online survey took approx. 15 to 20 minutes and was not compensated. The first page of the survey comprised a participant information sheet, where participants’ informed consent to participate was obtained before any of the questions were displayed. Underaged participants (<18 years) were urged to obtain assent from their parents/legal guardians, before filling in the survey. Information on the first page included the objectives of the study, inclusion criteria, estimated duration, the use, storage and sharing of the data, compensation (none), and information about the type of questions in the survey. Moreover, participants needed to confirm that they were

participating the survey for the first time. To ensure high data quality, five attention check items were included in the survey (e.g., “We want to make sure you are paying attention. What is 4+5?”). The data analysed in this study was collected between 17.05.2021 and 03.09.2021. Questions incorporating retrospective recall were all aligned to the period of “past four weeks,” matching the presented LEBA instrument.

In addition to the LEBA questionnaire, which is subject of the current study, the following variables and items were assessed but not included in the analysis:

- Sleep disturbance and sleep-related impairment (adult and pediatric versions) (Bevans et al., 2019; Buysse et al., 2010; Forrest et al., 2018; Harb, Hidalgo, & Martau, 2015; L. Yu et al., 2011)
- Sleep duration, timing, and latency, chronotype, social jetlag, time in bed, work/sleep schedule and outdoor light exposure duration (version for adults and adolescents) (Roenneberg, Wirz-Justice, & Mellow, 2003)
- Sleep environment (Olivier et al., 2016)
- Meal timing & caffeine consumption [custom items]
- Light sensitivity (photophobia vs. photophilia) (Wu & Hallett, 2017)
- Self-reported pubertal stage (only if younger than 18 years old) (Petersen, Crockett, Richards, & Boxer, 1988)

Furthermore, the following 1-item demographic variables were assessed:

- Age
- Sex
- Gender identity
- Occupational Status
- COVID-19 related Occupational setting during the past four weeks
- Time zone & country of residence
- English as native language

## Participants

Table 1 summarizes the survey participants' demographic characteristics. Only participants completing the full LEBA questionnaire were included, thus there are no missing values in the item analyses. XX participants were excluded from analysis due to not passing at least one of the "attention check" items. For exploring initial factor structure (EFA), a sample of 250-300 is recommended (Comrey & Lee, 1992; Schönbrodt & Perugini, 2013). For estimating the sample size for the confirmatory factor analysis (CFA) we followed the N:q rule (Bentler & Chou, 1987; Jackson, 2003; Kline, 2015; Worthington & Whittaker, 2006), where ten participants per parameter is required to earn trustworthiness of the result. Our sample size exceeds these requirements: Anonymous responses from a total of  $n = 690$  participants were included in the analysis of the current study, split into samples for exploratory (EFA:  $n = 428$ ) and confirmatory factor analysis (CFA:  $n = 262$ ). The EFA sample included participants filling out the questionnaire from 17.05.2021 to XX.XX.XXXX, whereas participants who filled out the questionnaire from YY.YY.YYYY to 03.09.2021 were included in the CFA analysis. Participants indicated filling out the online survey from a diverse range of geographic locations. The four most common geographic locations included:

	x
United States - America/New_York (UTC -04:00)	63
United Kingdom - Europe/London (UTC)	57
Germany - Europe/Berlin (UTC +01:00)	53
India - Asia/Kolkata (UTC +05:30)	38

For a full list of geographic locations, see Suppl. Table X in the appendix.

Age among all participants ranged from 11 years to 84 years [EFA:  $min = 11$ ,  $max = 84$ ; CFA:  $min = 12$ ,  $max = 74$ ], with an overall mean of  $\sim 33$  years of age [Overall:  $M = 32.95$ ,  $SD = 14.57$ ; EFA:  $M = 32.99$ ,  $SD = 15.11$ ; CFA:  $M = 32.89$ ,  $SD = 13.66$ ]. In total

325 (47%) of the participants indicated female sex [EFA: 189 (44%); CFA: 136 (52%)], 351 (51%) indicated male [EFA: 230 (54%); CFA: 121 (46%)] and 14 (2.0%) indicated other sex [EFA: 9 (2.1%), CFA: 5 (1.9%)]. Overall, 49 (7.2%) [EFA: 33 (7.8%); CFA: 16 (6.2%)] participants indicated a gender-variant identity. In a “Yes/No” question regarding native language, 320 (46%) of respondents [EFA: 191 (45%); CFA: 129 (49%)] indicated to be native English speakers. For their “Occupational Status,” more than half of the overall sample reported that they currently work [Overall: 396 (57%); EFA: 235 (55%); CFA: 161 (61%)], whereas 174 (25%) [EFA: 122 (29%); CFA: 52 (20%)] reported that they go to school and 120 (17%) [EFA: 71 (17%); CFA: 49 (19%)] responded that they do “Neither.” With respect to the COVID-19 pandemic we asked participants to indicate their occupational setting during the last four weeks: In the overall sample 303 (44%) [EFA: 194 (45%); CFA: 109 (42%)] of the participants indicated that they were in a home office/home schooling setting., while 109 (16%) overall [EFA: 68 (16%) ; CFA: 41 (16%)] reported face-to-face work/schooling. Lastly, 147 (21%) overall [EFA: 94 (22%) ; CFA: 53 (20%)] reported a combination of home- and face-to-face work/schooling, whereas 131 (19%) overall [EFA: 72 (17%); CFA: 59 (23%)] filled in the “Neither (no work or school, or indication)” response option. We tested all demographic variables in Table 1 for significant group differences between the EFA and CFA sample, applying Wilcoxon rank sum test for the continuous variable “Age” and Pearson’s Chi-squared test for all other categorical variables via the gtsummary R package’s “add\_p” function (Sjoberg et al., 2021a) . The p-values were corrected for multiple testing applying false discovery rate (FDR) via the “add\_q” function of the same package. After p-value (FDR) correction for multiple testing, none of the demographic variables were significantly different between the EFA sample and the CFA sample (all q-values  $q \geq 0.2$  , indicating equivalence).

1. Describe EFA and CFA sample separately.
2. Sampling technique: Convince sampling (non-probability sample)
3. Method: cross-sectional survey



169 4. How many missing data?

170 5. How incomplete data were addressed.

171 6. Why such sample was chosen?

## 172 **Procedure**

### 173 **Development of the Scale.**

174 1. How the items were generated

175 2. How the literature was reviewed to identify construct adequacy of the items.

176 3. Discuss the expert panel review process to assess content validity

177 **Data Collection.** Timeline of data collection, mode of data collection.

## 178 **Analytic Strategies**

179 We used R (version 4.1.0), including several R packages, for our analyses. Initially,  
180 our tool have six poin Likert type response scale(0:Does not apply/I don't know; 1:Never,  
181 2:Rarely; 3:Sometimes; 4:Often; 5: Alsways). As our purpose was to capture light  
182 exposure related behavior, "Does not apply/I don't know" and "Never" were providing  
183 similar information. As such we decided to collapse "Does not apply/I don't know" and  
184 "Never" options into one making it a 5 point Likert type response scale. Necessary  
185 assumptions of EFA, including sample adequacy, normality assumptions, quality of  
186 correlation matrix, were assessed. Our data violated both the univariate and multivariate  
187 normality assumptions. Due to these violations and the ordinal nature of our response  
188 data, we used a polychoric correlation matrix (C. Desjardins & Bulut, 2018) for the EFA.  
189 We employed principal axis (PA) as a factor extraction method with varimax rotation. PA  
190 is robust to the normality assumption violations (Watkins, 2020). The obtained latent  
191 structure was confirmed by another factor extraction method: the minimum residuals

extraction method as well. We used a combination factor identification method including scree plot (Cattell, 1966), Horn's parallel analysis (Horn, 1965), minimum average partials method (Velicer, 1976), and hull method (Lorenzo-Seva, Timmerman, & Kiers, 2011) to identify factor numbers. Additionally, to determine the simple structure, we followed the following guidelines recommended by psychometricians (i) no factors with fewer than three items (ii) no factors with a factor loading  $<0.3$  (iii) no items with cross-loading greater than .3 across factors (Bandalos & Finney, 2018). We also conducted psychometric analysis on non-merged response options data (supplementary analysis) and rejected the latent structure obtained as the factors were less interpretable.

## Results

### Exploratory Factor Analysis

Sampling adequacy was checked using Kaiser-Meyer-Olkin (KMO) measures of sampling adequacy (Kaiser, 1974). The overall KMO value for 48 items was 0.63 which was above the cutoff value (.50) indicating a mediocre sample (Hutcheson, 1999). Table 2 summarizes the univariate descriptive statistics for the 48 items. Some of the items were skewed with high Kurtosis values. Our data violated both univariate normality (Shapiro-Wilk statistics; (Shapiro & Wilk, 1965)) and multivariate normality assumptions (Marida's test; (Mardia, 1970)). Multivariate skew was = 583.80 ( $p < 0.001$ ) and multivariate kurtosis was = 2,749.15 ( $p < 0.001$ ). Due to these violations and ordinal nature of the response data polychoric correlations over Pearson's correlations was chosen (C. Desjardins & Bulut, 2018). Bartlett's test of sphericity (Bartlett, 1954),  $\chi^2(1128) = 5042.86$ ,  $p < .001$  indicated the correlations between items are adequate for the EFA. However only 4.96% of the inter-item correlation coefficients were greater than .30. The inter item correlation ranged between .44 to .91. And the corrected item-total correlations ranged between .10 to .44.

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218 Scree plot ( Figure 3) suggested a six-factor solution. Horn's parallel analysis  
219 (Horn, 1965) with 500 iterations also indicated a six-factor solution. However, the  
220 minimum average partial (MAP) method (Velicer, 1976) and Hull method (Lorenzo-Seva  
221 et al., 2011) suggested a five-factor solution. As a result, we tested both five-factor and  
222 six-factor solutions.

223 With initial 48 items we conducted three rounds of EFA gradually discarded  
224 problematic items. (cross-loading items and poor factor loading ( $<.30$ ) items). Finally, a  
225 five-factor EFA solution with 25 items was accepted with low RMSR = 0.08 (Brown,  
226 2015), all factor-loading higher than .30 and no cross-loading greater than .30. We  
227 confirmed this five-factor latent structure using varimax rotation with a minimum residual  
228 extraction method (Table??). Table3 displays the factor-loading (structural coefficients)  
229 and communality of the items. The absolute value of the factor-loading ranged from -.49  
230 to .99 indicating strong coefficients. The communalities ranged between .11 to .99.  
231 However, the histogram of the absolute values of non-redundant residual-correlations  
232 Fig5 showed 26% correlations greater than the absolute value of .05, indicating a  
233 possible under-factoring. (C. D. Desjardins, 2018). Subsequently, we fitted a six-factor  
234 solution. However, a factor emerged with only one salient variable loading in the  
235 six-factor solution, thus disqualifying the six-factor solution (Table??).

236 In the five-factor solution, the first factor contained three items and explained  
237 10.25% of the total variance with a satisfactory internal reliability coefficient ( $alpha =$   
238 .86). All the items in this factor stemmed from the individual's preference to use blue light  
239 filters in different light environments. The second factor contained six items and  
240 explained 9.93% of the total variance with a satisfactory internal reliability coefficient  
241 ( $alpha = .71$ ). Items under this factor commonly investigate an individual's hours spent  
242 outdoor. The third factor contained five items and explained 8.83% of the total variance.

Items under this factor dealt with the specific behaviors pertaining to sleep. The internal consistency reliability coefficient was,  $\alpha = .68$ . The fourth factor contained five items and explained 8.44% of the total variance with an internal consistency coefficient,  $\alpha = .62$ . These five items stemmed from the behavior related to an individual's cellphone usage during the sleep-wakeup time. Lastly, the fifth factor contained six items and explained 6.14% of the total variance. This factor tried to measure an individual's behavior lead by the awareness of light's influence on health. However, this factor showed unsatisfactory internal consistency reliability ( $\alpha = .53$ ). It is essential to attain a balance between psychometric properties and interpretability of the common themes when exploring the latent structure. As all of the emerged factors are highly interpretable and relevant towards our aim to capture light exposure related behavior, regardless of the apparent low reliability of the two factors, we retain the five-factor solution with 23 items for our confirmatory factor analysis (CFA). Two items showed negative factor-loading (items 44 and 21). Upon inspection, it was understood that these items are negatively correlated to the common theme, and thus in the CFA analysis, we reversed the response code for these two items.

### Confirmatory Factor Analysis

We conducted a categorical confirmatory factor analysis with robust weighted least square (WLSMV) estimator as our response data was in ordinary nature (C. Desjardins & Bulut, 2018). Several indices are suggested to measure model fit. These indices can be categorized as absolute, comparative and parsimony fit indices (Brown, 2015). Absolute fit assess the model fit at an absolute level using indices including chi-square test statistics and the standardized root mean square (SRMR). parsimony fit indices including the root mean square error of approximation (RMSEA) considers the number of free parameters in the model to assess the parsimony of the model. Comparative fit indices evaluate the fit of the specified model solution in relation to a more restricted baseline

model restricting all covariances among the indicators as zero. Comparative fit index (CFI) and the Tucker Lewis index (TLI) are such two comparative fit indices. Commonly used Model fit guidelines (Hu & Bentler, 1999; Schumacker & Lomax, 2004) includes (i) Reporting of chi-square test statistics (A non-significant test statistics is required to reflect model fit) (i) CFI and TLI (CFI/TLI close to .95 or above/ranging between 90-95 and above) (ii) RMSEA (close to .06 or below), (iii) SRMR (close to .08 or below) to estimate the model fit. Table 4 summarizes the fit indices of our fitted model. Our fitted model failed to attain an absolute fit estimated by the chi-square test. However, the chi-square test is sensitive to sample size and not recommended to be used as the sole index of absolute model fit (Brown, 2015). Another absolute fit index we obtained in our analysis was SRMR which does not work well with categorical data (C.-Y. Yu, 2002). Subsequently, we judged the model fit based on the comparative fit indices: CFI, TLI and parsimony fit index-RMSEA. Our fitted model attained acceptable fit (CFI = .94; TLI = .93; RMSEA = .06, [.05-.07, 90% CI]) with two imposed equality constraints on item pairs 32-33 and 19-17. However SRMR value was higher than the guideline (SRMR = .12). Further by allowing one pair of items (30-41) to covary their error variance and discarding two item (item 37 & 26) for very low r-square value, our model attained best fit (CFI = .97; TLI = .96; RMSEA = .05 [.04-.06, 90% CI]) and SRMR value (SRMR = .09) was also close to the suggestions of Hu and Bentler (1999). Since reliability coefficient Cronbach's alpha tends to mis-measure reliability for multidimensional construct (Sijtsma, 2009) we report McDonald's omega(total) as the internal consistency reliability coefficient. McDonald's omega(total) is reported to be a satisfactory lower bound reliability coefficient that works for both unidimensional and multidimensional construct (Zinbarg, Revelle, Yovel, & Li, 2005). McDonald's omega(total) for the five factors were .90, .80, .61, .72, .45 respectively. McDonald's omega(total) coefficient for the total scale was .73.

## Analysing the quality of items by Item Information Theory

We sought the IRT to gather information regarding the item quality. IRT complements the conventional classical test theory-based analysis by gathering information on item discrimination and item difficulty (Baker, 2017). Here, an item's quality is judged based on item information in relation to participants' latent trait level ( $\theta$ ). We gathered evidence on item quality by fitting each factor of LEBA with the graded response model (6 to the combined EFA sample and CFA sample ( $n = 690$ )). Item discrimination indicates the pattern of variation in the categorical responses with the changes in latent trait, and item information curve (IIC) indicates the amount of information an item carries along the latent trait continuum. Here, we reported the item discrimination parameter and only discarded the items with relatively flat item information curve (information  $< .2$ ) to develop the short form of LEBA. Baker (2017) categorized the item discrimination in as none = 0; very low = 0.01 to 0.34; low = 0.35 to 0.64; moderate = 0.65 to 1.34; high = 1.35 to 1.69; very high  $> 1.70$ . Item discrimination parameters of our scale fell in very high (10 items), high (4 items), moderate (4 items), low (5 items) indicating a good range of discrimination along the latent trait. Examination of the item information curve indicated 6 items (1, 25, 9, 38, 30, & 41) had relatively flat information curves thus discarded. We also gathered evidence of item fit and person fit to our fitted model.

Test information curve (TIC) indicate the amount of information an the full-scale carry along the latent trait continuum. As we treated each factor of LEBA as an unidimensional construct we obtain 5 TICs. These information curves indicated except blue filter factor, the other factor's TICs are roughly centered on the center of the trait continuum ( $\theta$ ). Also the amount of information changed rather steadily with the change of ( $\theta$ ). Thus we conferred the LEBA scale (except blue filter) estimated the light exposure related behavior with precision near the center of trait continuum (Baker,

2017) which is sufficient to discriminate between latent trait measured by the each factor. The blue filter factor had a peak to the right side of the center of latent trait indicating its ability to providing information only for people who already have some preference towards using blue-filters.

Our result also indicated all the items fitted well to the respective models as assessed by assessed by RMSEA value obtained from Signed-X2 index implementation. All of the items had RMSEA value  $<.06$  indicating adequate fit. Person fit indicates the validity and meaningfulness of the fitted model at the participants latent trait level (C. Desjardins & Bulut, 2018). We estimated the person fit statistics using standardized fit index Zh statistics (Drasgow, Levine, & Williams, 1985).  $Zh < -2$  should be considered as a misfit. Fig indicates that Zh is larger than -2 for most participants, suggesting a good fit of the selected IRT models.

The overall we can concluded that IRT analysis indicated LEBA is a psychometrically sound measure. Item fit indexes and person fit index for all five fitted model were acceptable. Items had diverse slope parameters indicating a good range of discrimination- the ability to differentiate respondents with different levels of the light exposure related behavior. All-in-all we can recommend the LEBA to be used to capture light exposure related behavior.

## Discussion

## References

- Anwar Siraji, M. (2021). *Tabledown: A companion pack for the book "basic & advanced psychometrics in r"*. Retrieved from <https://github.com/masiraji/taledown>
- Aust, F., & Barth, M. (2020). *papaja: Prepare reproducible APA journal articles with R Markdown*. Retrieved from <https://github.com/crsh/papaja>
- Baker, F. B. (2017). *The Basics of Item Response Theory Using R* (1st ed. 2017.). Springer.
- Bandalos, D. L., & Finney, S. J. (2018). Factor analysis: Exploratory and confirmatory. In *The reviewer's guide to quantitative methods in the social sciences* (pp. 98–122). Routledge.
- Barnier, J., Briatte, F., & Larmarange, J. (2020). *Questionr: Functions to make surveys processing easier*. Retrieved from <https://CRAN.R-project.org/package=questionr>
- Barth, M. (2021). *tinylabls: Lightweight variable labels*. Retrieved from <https://github.com/mariusbarth/tinylabls>
- Bartlett, M. (1954). A Note on the Multiplying Factors for Various Chi-square Approximations. *Journal of the Royal Statistical Society. Series B, Methodological*, 16(2), 296–298.
- Bentler, P. M., & Chou, C.-P. (1987). Practical Issues in Structural Modeling. *Sociological Methods & Research*, 16(1), 78–117. <https://doi.org/10.1177/0049124187016001004>
- Bevans, K. B., Meltzer, L. J., La Motte, A. de, Kratchman, A., Viél, D., & Forrest, C. B. (2019). Qualitative development and content validation of the PROMIS pediatric sleep health items. *Behavioral Sleep Medicine*, 17(5), 657–671. <https://doi.org/10.1080/15402002.2018.1461102>
- Brown, T. A. (2015). *Confirmatory factor analysis for applied research* (2nd ed.).



New York, NY, US: The Guilford Press.

Bryer, J., & Speerschneider, K. (2016). *Likert: Analysis and visualization likert items*. Retrieved from <https://CRAN.R-project.org/package=likert>

Buchanan, E. M., Gillenwaters, A., Scofield, J. E., & Valentine, K. D. (2019).

*MOTE: Measure of the Effect: Package to assist in effect size calculations and their confidence intervals*. Retrieved from <http://github.com/doomlab/MOTE>

Buyse, D. J., Yu, L., Moul, D. E., Germain, A., Stover, A., Dodds, N. E., ...

Pilkonis, P. A. (2010). Development and validation of patient-reported outcome measures for sleep disturbance and sleep-related impairments. *Sleep*, 33(6), 781–792. <https://doi.org/10.1093/sleep/33.6.781>

Cattell, R. B. (1966). The Scree Test For The Number Of Factors. *Multivariate Behavioral Research*, 1(2), 245–276.

[https://doi.org/10.1207/s15327906mbr0102\\_10](https://doi.org/10.1207/s15327906mbr0102_10)

Chalmers, R. P. (2012). mirt: A multidimensional item response theory package for the R environment. *Journal of Statistical Software*, 48(6), 1–29.

<https://doi.org/10.18637/jss.v048.i06>

Chang, W., Cheng, J., Allaire, J., Sievert, C., Schloerke, B., Xie, Y., ... Borges, B. (2021). *Shiny: Web application framework for r*. Retrieved from

<https://CRAN.R-project.org/package=shiny>

Comrey, A. L., & Lee, H. B. (1992). *A first course in factor analysis, 2nd ed.*

Hillsdale, NJ, US: Lawrence Erlbaum Associates, Inc.

Conigrave, J. (2020). *Corx: Create and format correlation matrices*. Retrieved from <https://CRAN.R-project.org/package=corx>

Dahl, D. B., Scott, D., Roosen, C., Magnusson, A., & Swinton, J. (2019). *Xtable: Export tables to LaTeX or HTML*. Retrieved from

<https://CRAN.R-project.org/package=xtable>

Desjardins, C. D. (2018). *Handbook of educational measurement and*

393 *psychometrics using R*. (O. Bulut & ProQuest (Firm), Eds.). Boca Raton, FL :  
394 CRC Press.

395 Desjardins, C., & Bulut, O. (2018). *Handbook of Educational Measurement and*  
396 *Psychometrics Using R*. <https://doi.org/10.1201/b20498>

397 Dinno, A. (2018). *Paran: Horn's test of principal components/factors*. Retrieved  
398 from <https://CRAN.R-project.org/package=paran>

399 Drasgow, F., Levine, M. V., & Williams, E. A. (1985). Appropriateness  
400 measurement with polychotomous item response models and standardized  
401 indices. *British Journal of Mathematical and Statistical Psychology*, 38(1),  
402 67–86.

403 Epskamp, S. (2019). *semPlot: Path diagrams and visual analysis of various SEM*  
404 *packages' output*. Retrieved from  
405 <https://CRAN.R-project.org/package=semPlot>

406 Epskamp, S., Cramer, A. O. J., Waldorp, L. J., Schmittmann, V. D., & Borsboom,  
407 D. (2012). qgraph: Network visualizations of relationships in psychometric  
408 data. *Journal of Statistical Software*, 48(4), 1–18.

409 F.lux Software LLC. (2021). F.lux (Version 4.120). Retrieved from  
410 <https://justgetflux.com/>

411 Forrest, C. B., Meltzer, L. J., Marcus, C. L., La Motte, A. de, Kratchman, A.,  
412 Buysse, D. J., ... Bevens, K. B. (2018). Development and validation of the  
413 PROMIS pediatric sleep disturbance and sleep-related impairment item banks.  
414 *Sleep*, 41(6). <https://doi.org/10.1093/sleep/zsy054>

415 Fox, J., & Weisberg, S. (2019). *An R companion to applied regression* (Third).  
416 Thousand Oaks CA: Sage. Retrieved from  
417 <https://socialsciences.mcmaster.ca/jfox/Books/Companion/>

418 Fox, J., Weisberg, S., & Price, B. (2020). *carData: Companion to applied*  
419 *regression data sets*. Retrieved from

420 <https://CRAN.R-project.org/package=carData>

421 Harb, F., Hidalgo, M. P., & Martau, B. (2015). Lack of exposure to natural light in  
422 the workspace is associated with physiological, sleep and depressive  
423 symptoms. *Chronobiology International*, 32(3), 368–375.

424 <https://doi.org/10.3109/07420528.2014.982757>

425 Harrell Jr, F. E., Charles Dupont, with contributions from, & others., many. (2021).

426 *Hmisc: Harrell miscellaneous*. Retrieved from

427 <https://CRAN.R-project.org/package=Hmisc>

428 Harris, P. A., Taylor, R., Minor, B. L., Elliott, V., Fernandez, M., O’Neal, L., ...

429 others. (2019). The REDCap consortium: Building an international community  
430 of software platform partners. *Journal of Biomedical Informatics*, 95, 103208.

431 Harris, P. A., Taylor, R., Thielke, R., Payne, J., Gonzalez, N., & Conde, J. G.

432 (2009). Research electronic data capture (REDCap)—a metadata-driven  
433 methodology and workflow process for providing translational research  
434 informatics support. *Journal of Biomedical Informatics*, 42(2), 377–381.

435 Henry, L., & Wickham, H. (2020). *Purrr: Functional programming tools*. Retrieved

436 from <https://CRAN.R-project.org/package=purrr>

437 Horn, J. L. (1965). A rationale and test for the number of factors in factor analysis.

438 *Psychometrika*, 30(2), 179–185. <https://doi.org/10.1007/BF02289447>

439 Hu, L., & Bentler, P. M. (1999). Cutoff criteria for fit indexes in covariance structure  
440 analysis: Conventional criteria versus new alternatives. *Structural Equation*  
441 *Modeling: A Multidisciplinary Journal*, 6(1), 1–55.

442 <https://doi.org/10.1080/10705519909540118>

443 Hutcheson, G. D. (1999). *The multivariate social scientist : Introductory statistics*  
444 *using generalized linear models*. London : SAGE.

445 Iannone, R. (2016). *DiagrammeRsvg: Export DiagrammeR graphviz graphs as*

446 *SVG*. Retrieved from <https://CRAN.R-project.org/package=DiagrammeRsvg>

- Iannone, R. (2021). *DiagrammeR: Graph/network visualization*. Retrieved from <https://github.com/rich-iannone/DiagrammeR>
- Jackson, D. L. (2003). Revisiting Sample Size and Number of Parameter Estimates: Some Support for the N:q Hypothesis. *Structural Equation Modeling*, 10(1), 128–141. [https://doi.org/10.1207/S15328007SEM1001\\_6](https://doi.org/10.1207/S15328007SEM1001_6)
- Johnson, P., & Kite, B. (2020). *semTable: Structural equation modeling tables*. Retrieved from <https://CRAN.R-project.org/package=semTable>
- Johnson, P., Kite, B., & Redmon, C. (2020). *Kutils: Project management tools*. Retrieved from <https://CRAN.R-project.org/package=kutils>
- Jorgensen, T. D., Pornprasertmanit, S., Schoemann, A. M., & Rosseel, Y. (2021). *semTools: Useful tools for structural equation modeling*. Retrieved from <https://CRAN.R-project.org/package=semTools>
- Kaiser, H. F. (1974). An index of factorial simplicity. *Psychometrika*, 39(1), 31–36. <https://doi.org/10.1007/bf02291575>
- Kassambara, A. (2019). *Ggcorrplot: Visualization of a correlation matrix using 'ggplot2'*. Retrieved from <https://CRAN.R-project.org/package=ggcorrplot>
- Kline, R. B. (2015). *Principles and practice of structural equation modeling*. The Guilford Press.
- Kowarik, A., & Templ, M. (2016). Imputation with the R package VIM. *Journal of Statistical Software*, 74(7), 1–16. <https://doi.org/10.18637/jss.v074.i07>
- Lishinski, A. (2021). *lavaanPlot: Path diagrams for 'lavaan' models via 'DiagrammeR'*. Retrieved from <https://CRAN.R-project.org/package=lavaanPlot>
- Lorenzo-Seva, U., Timmerman, M., & Kiers, H. (2011). The Hull Method for Selecting the Number of Common Factors. *Multivariate Behavioral Research*, 46, 340–364. <https://doi.org/10.1080/00273171.2011.564527>
- Makowski, D., Ben-Shachar, M. S., Patil, I., & Lüdtke, D. (2020). Methods and

algorithms for correlation analysis in r. *Journal of Open Source Software*,  
5(51), 2306. <https://doi.org/10.21105/joss.02306>

Mardia, K. V. (1970). Measures of multivariate skewness and kurtosis with  
applications. *Biometrika*, 57(3), 519–530.

<https://doi.org/10.1093/biomet/57.3.519>

Mock, T. (2021). *gtExtras: A collection of helper functions for the gt package*.

Retrieved from <https://github.com/jthomasmock/gtExtras>

Müller, K., & Wickham, H. (2021). *Tibble: Simple data frames*. Retrieved from  
<https://CRAN.R-project.org/package=tibble>

Navarro-Gonzalez, D., & Lorenzo-Seva, U. (2021). *EFA.MRFA: Dimensionality  
assessment using minimum rank factor analysis*. Retrieved from  
<https://CRAN.R-project.org/package=EFA.MRFA>

Olivier, K., Gallagher, R. A., Killgore, W. D. S., Carrazco, N., Alfonso-Miller, P., ...  
Grandner, M. A. (2016). Development and initial validation of the assessment  
of sleep environment: A novel inventory for describing and quantifying the  
impact of environmental factors on sleep. *Sleep*, 39(Abstract Supplement:  
A367).

Ooms, J. (2021a). *Magick: Advanced graphics and image-processing in r*.  
Retrieved from <https://CRAN.R-project.org/package=magick>

Ooms, J. (2021b). *Rsvg: Render SVG images into PDF, PNG, PostScript, or  
bitmap arrays*. Retrieved from <https://CRAN.R-project.org/package=rsvg>

Petersen, A. C., Crockett, L., Richards, M., & Boxer, A. (1988). A self-report  
measure of pubertal status: Reliability, validity, and initial norms. *Journal of  
Youth and Adolescence*, 17(2), 117–133. <https://doi.org/10.1007/BF01537962>

Pornprasertmanit, S., Miller, P., Schoemann, A., & Jorgensen, T. D. (2021).  
*Simsem: SIMulated structural equation modeling*. Retrieved from  
<https://CRAN.R-project.org/package=simsem>

501 R Core Team. (2021). *R: A language and environment for statistical computing*.

502 Vienna, Austria: R Foundation for Statistical Computing. Retrieved from

503 <https://www.R-project.org/>

504 Revelle, W. (2021). *Psych: Procedures for psychological, psychometric, and*

505 *personality research*. Evanston, Illinois: Northwestern University. Retrieved

506 from <https://CRAN.R-project.org/package=psych>

507 Roenneberg, T., Wirz-Justice, A., & Mellow, M. (2003). Life between clocks: Daily

508 temporal patterns of human chronotypes. *Journal of Biological Rhythms*,

509 *18*(1), 80–90.

510 Rosseel, Y. (2012). lavaan: An R package for structural equation modeling.

511 *Journal of Statistical Software*, *48*(2), 1–36. Retrieved from

512 <https://www.jstatsoft.org/v48/i02/>

513 Ryu, C. (2021). *Dlookr: Tools for data diagnosis, exploration, transformation*.

514 Retrieved from <https://CRAN.R-project.org/package=dlookr>

515 Sarkar, D. (2008). *Lattice: Multivariate data visualization with r*. New York:

516 Springer. Retrieved from <http://lmdvr.r-forge.r-project.org>

517 Schönbrodt, F. D., & Perugini, M. (2013). At what sample size do correlations

518 stabilize? *Journal of Research in Personality*, *47*(5), 609–612.

519 <https://doi.org/10.1016/j.jrp.2013.05.009>

520 Schumacker, R. E., & Lomax, R. G. (2004). *A beginner's guide to structural*

521 *equation modeling*. psychology press.

522 Shapiro, S. S., & Wilk, M. B. (1965). An analysis of variance test for normality

523 (complete samples). *Biometrika*, *52*(3-4), 591–611.

524 <https://doi.org/10.1093/biomet/52.3-4.591>

525 Sijtsma, K. (2009). On the use, the misuse, and the very limited usefulness of

526 cronbach's alpha. *Psychometrika*, *74*(1), 107.

527 Sjöberg, D. D., Curry, M., Hannum, M., Larmarange, J., Whiting, K., & Zabor, E. C.

- (2021b). *Gtsummary: Presentation-ready data summary and analytic result tables*. Retrieved from <https://CRAN.R-project.org/package=gtsummary>
- Sjoberg, D. D., Curry, M., Hannum, M., Larmarange, J., Whiting, K., & Zabor, E. C. (2021a). *Gtsummary: Presentation-ready data summary and analytic result tables*. Retrieved from <https://CRAN.R-project.org/package=gtsummary>
- Stauffer, R., Mayr, G. J., Dabernig, M., & Zeileis, A. (2009). Somewhere over the rainbow: How to make effective use of colors in meteorological visualizations. *Bulletin of the American Meteorological Society*, 96(2), 203–216. <https://doi.org/10.1175/BAMS-D-13-00155.1>
- Terry M. Therneau, & Patricia M. Grambsch. (2000). *Modeling survival data: Extending the Cox model*. New York: Springer.
- Ushey, K., McPherson, J., Cheng, J., Atkins, A., & Allaire, J. (2021). *Packrat: A dependency management system for projects and their r package dependencies*. Retrieved from <https://CRAN.R-project.org/package=packrat>
- van Lissa, C. J. (2021). *tidySEM: Tidy structural equation modeling*. Retrieved from <https://CRAN.R-project.org/package=tidySEM>
- Velicer, W. (1976). Determining the Number of Components from the Matrix of Partial Correlations. *Psychometrika*, 41, 321–327. <https://doi.org/10.1007/BF02293557>
- Venables, W. N., & Ripley, B. D. (2002). *Modern applied statistics with s* (Fourth). New York: Springer. Retrieved from <https://www.stats.ox.ac.uk/pub/MASS4/>
- Watkins, M. (2020). *A Step-by-Step Guide to Exploratory Factor Analysis with R and RStudio*. <https://doi.org/10.4324/9781003120001>
- Weinzaepflen, C., & Spitschan, M. (2021). Enlighten your clock: How your body tells time. Open Science Framework. <https://doi.org/10.17605/OSF.IO/ZQXVH>
- Wickham, H. (2011). The split-apply-combine strategy for data analysis. *Journal of Statistical Software*, 40(1), 1–29. Retrieved from

555 <http://www.jstatsoft.org/v40/i01/>

556 Wickham, H. (2016). *ggplot2: Elegant graphics for data analysis*. Springer-Verlag  
557 New York. Retrieved from <https://ggplot2.tidyverse.org>

558 Wickham, H. (2019). *Stringr: Simple, consistent wrappers for common string*  
559 *operations*. Retrieved from <https://CRAN.R-project.org/package=stringr>

560 Wickham, H. (2021a). *Forcats: Tools for working with categorical variables*  
561 *(factors)*. Retrieved from <https://CRAN.R-project.org/package=forcats>

562 Wickham, H. (2021b). *Tidyr: Tidy messy data*. Retrieved from  
563 <https://CRAN.R-project.org/package=tidyr>

564 Wickham, H., Averick, M., Bryan, J., Chang, W., McGowan, L. D., François, R., ...  
565 Yutani, H. (2019). Welcome to the tidyverse. *Journal of Open Source*  
566 *Software*, 4(43), 1686. <https://doi.org/10.21105/joss.01686>

567 Wickham, H., & Bryan, J. (2019). *Readxl: Read excel files*. Retrieved from  
568 <https://CRAN.R-project.org/package=readxl>

569 Wickham, H., François, R., Henry, L., & Müller, K. (2021). *Dplyr: A grammar of*  
570 *data manipulation*. Retrieved from <https://CRAN.R-project.org/package=dplyr>

571 Wickham, H., & Hester, J. (2021). *Readr: Read rectangular text data*. Retrieved  
572 from <https://CRAN.R-project.org/package=readr>

573 Wilke, C. O. (2020). *Cowplot: Streamlined plot theme and plot annotations for*  
574 *'ggplot2'*. Retrieved from <https://CRAN.R-project.org/package=cowplot>

575 Winston Chang. (2014). *Extrafont: Tools for using fonts*. Retrieved from  
576 <https://CRAN.R-project.org/package=extrafont>

577 Worthington, R. L., & Whittaker, T. A. (2006). Scale Development Research: A  
578 Content Analysis and Recommendations for Best Practices. *The Counseling*  
579 *Psychologist*, 34(6), 806–838. <https://doi.org/10.1177/0011000006288127>

580 Wu, Y., & Hallett, M. (2017). Photophobia in neurologic disorders. *Translational*  
581 *Neurodegeneration*, 6(1), 26. <https://doi.org/10.1186/s40035-017-0095-3>



- Yu, C.-Y. (2002). *Evaluating cutoff criteria of model fit indices for latent variable models with binary and continuous outcomes* (Thesis). ProQuest Dissertations Publishing.
- Yu, L., Buysse, D. J., Germain, A., Moul, D. E., Stover, A., Dodds, N. E., ... Pilkonis, P. A. (2011). Development of short forms from the PROMIS™ sleep disturbance and sleep-related impairment item banks. *Behavioral Sleep Medicine, 10*(1), 6–24. <https://doi.org/10.1080/15402002.2012.636266>
- Zeileis, A., & Croissant, Y. (2010). Extended model formulas in R: Multiple parts and multiple responses. *Journal of Statistical Software, 34*(1), 1–13. <https://doi.org/10.18637/jss.v034.i01>
- Zeileis, A., Fisher, J. C., Hornik, K., Ihaka, R., McWhite, C. D., Murrell, P., ... Wilke, C. O. (2020). colorspace: A toolbox for manipulating and assessing colors and palettes. *Journal of Statistical Software, 96*(1), 1–49. <https://doi.org/10.18637/jss.v096.i01>
- Zeileis, A., Hornik, K., & Murrell, P. (2009). Escaping RGBland: Selecting colors for statistical graphics. *Computational Statistics & Data Analysis, 53*(9), 3259–3270. <https://doi.org/10.1016/j.csda.2008.11.033>
- Zhu, H. (2021). *kableExtra: Construct complex table with 'kable' and pipe syntax*. Retrieved from <https://CRAN.R-project.org/package=kableExtra>
- Zinbarg, R. E., Revelle, W., Yovel, I., & Li, W. (2005). Cronbach's  $\alpha$ , Revelle's  $\beta$ , and McDonald's  $\omega$  h: Their relations with each other and two alternative conceptualizations of reliability. *Psychometrika, 70*(1), 123–133.

Table 1

*Dempgraphics*

Variable	Overall, N = 690	1. EFA Sample, N = 428	2. CFA Sample, N = 262	p-value	q-value
Age	32.95 (14.57)	32.99 (15.11)	32.89 (13.66)	0.5	0.5
Sex				0.14	0.4
Female	325 (47%)	189 (44%)	136 (52%)		
Male	351 (51%)	230 (54%)	121 (46%)		
Other	14 (2.0%)	9 (2.1%)	5 (1.9%)		
Gender-Variant Identity	49 (7.2%)	33 (7.8%)	16 (6.2%)	0.4	0.5
Native English Speaker	320 (46%)	191 (45%)	129 (49%)	0.2	0.5
Occupational Status				0.040	0.2
Work	396 (57%)	235 (55%)	161 (61%)		
School	174 (25%)	122 (29%)	52 (20%)		
Neither	120 (17%)	71 (17%)	49 (19%)		
Occupational setting				0.3	0.5
Home office/Home schooling	303 (44%)	194 (45%)	109 (42%)		
Face-to-face work/Face-to-face schooling	109 (16%)	68 (16%)	41 (16%)		
Combination of home- and face-to-face- work/schooling	147 (21%)	94 (22%)	53 (20%)		
Neither (no work or school, or in vacation)	131 (19%)	72 (17%)	59 (23%)		

<sup>1</sup> Mean (SD); n (%)<sup>2</sup> Wilcoxon rank sum test; Pearson's Chi-squared test<sup>3</sup> False discovery rate correction for multiple testing

Table 2

*Descriptive Statistics*

	Mean	SD	Skew	Kurtosis	Shapiro-Wilk Statistics	Item-Total Correlation
Item1	2.27	1.39	0.74	-0.81	0.81*	.25
Item2	2.87	1.59	0.08	-1.60	0.83*	.19
Item3	3.36	1.38	-0.48	-1.03	0.87*	.16
Item4	1.47	1.18	2.38	4.00	0.43*	.28
Item5	4.01	1.40	-1.22	0.07	0.70*	.13
Item6	2.79	1.55	0.19	-1.48	0.85*	.20
Item7	2.26	1.25	0.70	-0.60	0.85*	.19
Item8	2.97	1.20	-0.06	-0.94	0.91*	-.10
Item9	2.94	1.03	-0.12	-0.40	0.91*	.10
Item10	2.74	1.04	0.09	-0.74	0.91*	.28
Item11	2.18	0.90	0.60	0.12	0.86*	.26
Item12	2.36	1.22	0.59	-0.62	0.87*	.25
Item13	2.73	1.46	0.20	-1.36	0.87*	.33
Item14	2.14	1.31	0.77	-0.78	0.80*	.26
Item15	3.26	1.09	-0.26	-0.45	0.91*	.14
Item16	1.56	1.23	2.00	2.45	0.50*	.32
Item17	1.54	1.21	2.07	2.75	0.49*	.31
Item18	1.12	0.49	5.02	27.80	0.25*	.16
Item19	1.05	0.36	7.23	52.98	0.13*	.18
Item20	1.04	0.33	8.99	85.28	0.10*	.16
Item21	1.14	0.59	4.79	24.05	0.25*	.16
Item22	3.57	1.07	-0.65	-0.17	0.88*	.21
Item23	2.56	1.27	0.33	-1.00	0.89*	.11

Table 2 continued

	Mean	SD	Skew	Kurtosis	Shapiro-Wilk Statistics	Item-Total Correlation
Item24	4.14	0.99	-1.23	1.14	0.79*	.19
Item25	2.59	1.41	0.27	-1.27	0.86*	.19
Item26	2.25	1.27	0.69	-0.64	0.84*	.18
Item27	3.80	1.29	-0.87	-0.42	0.82*	.17
Item28	3.76	1.14	-0.68	-0.45	0.86*	.00
Item29	2.44	1.31	0.38	-1.14	0.86*	.11
Item30	1.48	1.11	2.18	3.35	0.48*	.24
Item31	3.00	1.62	-0.08	-1.61	0.83*	.44
Item32	3.55	1.65	-0.60	-1.34	0.76*	.43
Item33	3.62	1.64	-0.68	-1.25	0.74*	.32
Item34	3.42	1.83	-0.45	-1.69	0.69*	.33
Item35	3.86	1.67	-0.99	-0.85	0.65*	.23
Item36	1.54	1.25	2.13	2.86	0.46*	.36
Item37	1.33	0.91	3.03	8.43	0.41*	.01
Item38	4.30	1.08	-1.79	2.53	0.67*	.22
Item39	1.96	0.98	1.02	0.69	0.82*	.05
Item40	2.16	1.19	0.71	-0.54	0.84*	.14
Item41	1.31	0.81	2.75	6.92	0.43*	.21
Item42	3.93	1.48	-1.06	-0.44	0.71*	.18
Item43	1.64	1.18	1.79	2.02	0.60*	.15
Item44	3.51	1.30	-0.70	-0.59	0.85*	.39
Item45	2.22	1.48	0.71	-1.02	0.76*	.30
Item46	1.76	1.23	1.35	0.44	0.66*	.38
Item47	2.11	1.17	0.77	-0.39	0.83*	.32

Table 2 continued

	Mean	SD	Skew	Kurtosis	Shapiro-Wilk Statistics	Item-Total Correlation
Item48	2.60	1.25	0.29	-0.86	0.89*	.35

*Note.* \* $p < .001$

Table 3

*Factor loadings and communality of the retained items*

item	PA1	PA2	PA3	PA4	PA5	Communality	Uniqueness
item16	0.99					0.993	0.007
item36	0.94					0.899	0.101
item17	0.8					0.658	0.342
item11		0.79				0.642	0.358
item10		0.76				0.592	0.408
item12		0.65				0.465	0.535
item7		0.5				0.267	0.733
item8		-0.49				0.252	0.748
item9		0.32				0.113	0.887
item27			0.8			0.658	0.342
item3			0.8			0.682	0.318
item40			0.65			0.464	0.536
item30			0.45			0.353	0.647
item41			0.36			0.329	0.671
item33				0.74		0.555	0.445
item32				0.73		0.624	0.376
item35				0.66		0.454	0.546
item37				-0.39		0.174	0.826
item38				0.38		0.178	0.822
item46					0.6	0.422	0.578
item45					0.59	0.374	0.626
item25					0.41	0.193	0.807
item4					0.41	0.219	0.781
item1					0.4	0.17	0.83
item26					0.35	0.165	0.835
% of Variance	0.1	0.1	0.09	0.08	0.06		

*Note.* Only loading higher than .30 is reported

Table 4

*Fit indices of CFA*

Model	Chi-Square	df	CFI	TLI	RMSEA	RMSEA 90% Lower CI	RMSEA 90% Upper CI	SRMR
Five factor model:25	448.51	222.00	.94	0.93	0.06	0.05	0.07	0.12
Five factor model:23	346.59	221.00	.97	0.96	0.05	0.04	0.06	0.09

*Note.* df: Degrees of Freedom; CFI: Comparative Fit Index; TLI: Tucker Lewis Index; RMSEA: Root Mean Square Error of Approximation; CI: Confidence Interval; SRMR: Standardized Root Mean Square

Table 5

*Invariance Analysis*

	Chi-Square	df	CFI	TLI	RMSEA	RMSEA 90% Lower CI	RMSEA 90% Upper	SRMR	Chi-Sqr comparison	df*	p
Configural	632.20	442.00	0.95	0.94	0.06	0.05	0.07	0.13	-	-	-
Metric	644.58	458.00	0.95	0.95	0.06	0.05	0.07	0.13	18.019a	16	0.323
Scalar	714.19	522.00	0.95	0.95	0.05	0.04	0.06	0.13	67.961b	64	0.344
Residual	714.19	522.00	0.95	0.95	0.05	0.04	0.06	0.13	0c	0	NA
Structural	691.49	542.00	0.96	0.96	0.05	0.04	0.06	0.13	12.617d	20	0.893

*Note.* a = Metric vs Configural; b = Scalar vs Metric; c = Residual vs Scalar; d = Structural vs Residual; \* = df of model comparison



Table 6

*IRT Item parameters for the LEBA Scale*

	a	b1	b2	b3	b4
item16	28.55	0.78	0.90	1.06	1.40
item36	4.49	0.94	1.08	1.23	1.40
item17	2.81	0.97	1.11	1.38	1.62
item11	3.27	-0.79	0.65	1.54	2.31
item10	3.07	-1.27	-0.09	0.82	2.00
item12	1.72	-0.67	0.44	1.28	2.11
item7	1.09	-0.50	0.73	1.63	2.97
Ritem8	1.19	-2.26	-0.48	0.64	1.91
item9	0.91	-2.63	-0.96	1.11	3.49
item27	2.21	-1.88	-1.19	-0.73	0.30
item3	3.03	-1.24	-0.77	-0.20	0.66
item40	1.55	-0.51	0.46	1.32	2.22
item30	0.49	3.27	3.74	4.64	6.52
item41	0.51	3.87	4.78	6.39	8.91
item32	1.62	-1.03	-0.78	-0.42	0.16
item35	1.36	-1.09	-0.98	-0.75	-0.40
item38	0.40	-7.50	-5.58	-4.25	-0.91
item33	13.51	-0.66	-0.48	-0.24	0.13
item46	2.22	0.68	0.89	1.38	2.17
item45	1.51	0.30	0.55	1.17	1.91
item25	0.52	-1.37	-0.04	1.89	4.22
item4	0.84	2.44	2.80	3.18	3.67
item1	0.39	-0.91	1.52	3.25	5.53

*Note.* a = item discrimination parameter; b(1-4)  
= response category difficulty parameter

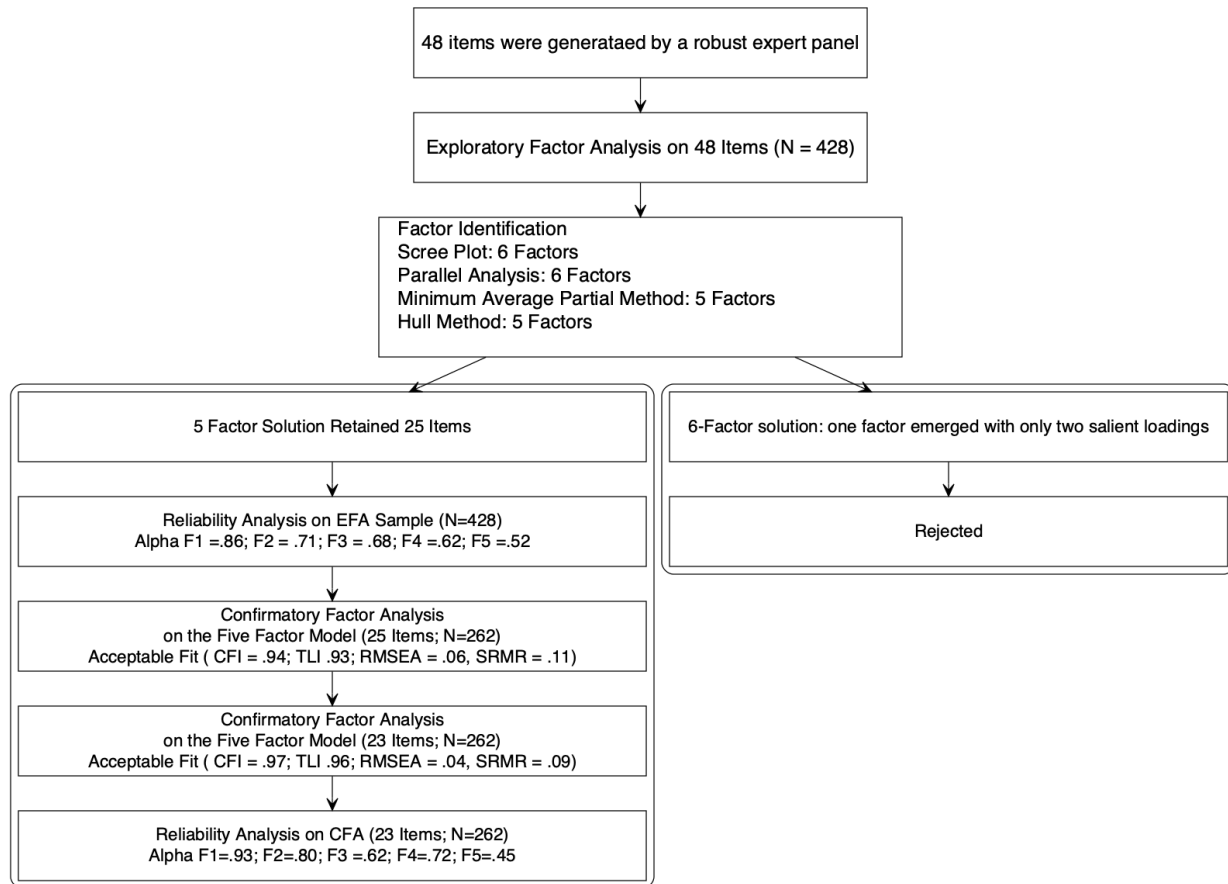


Figure 1. Development

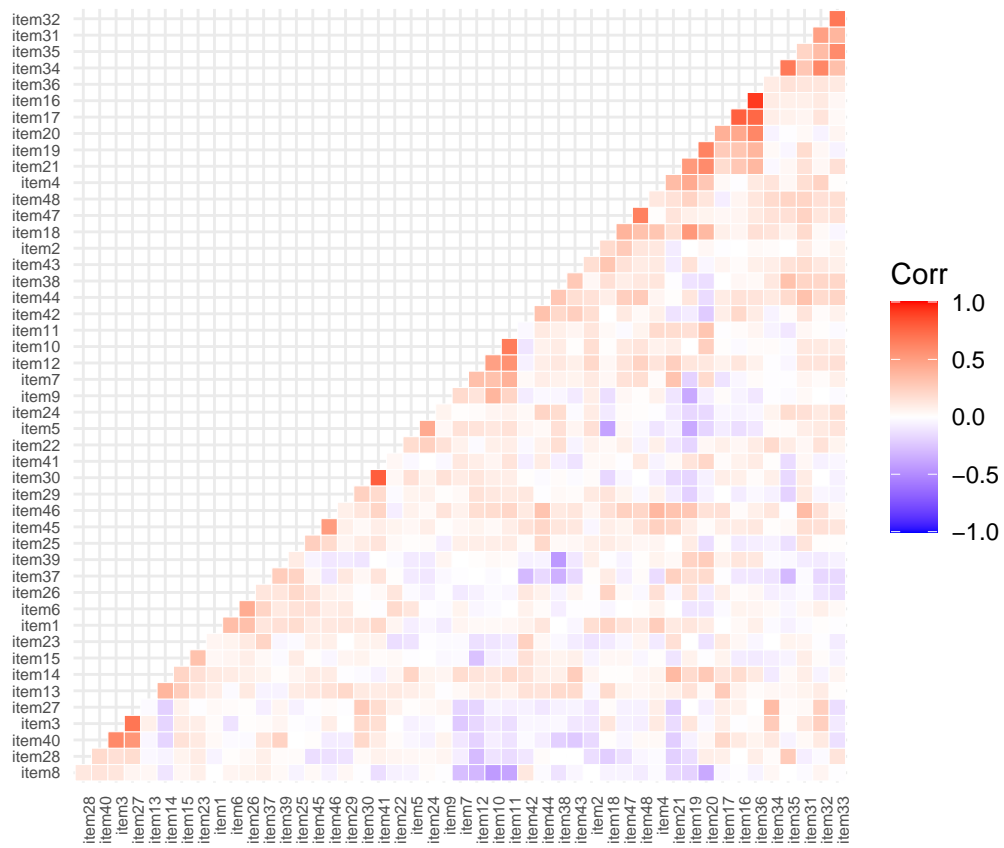


Figure 2. Correlation plot of the items

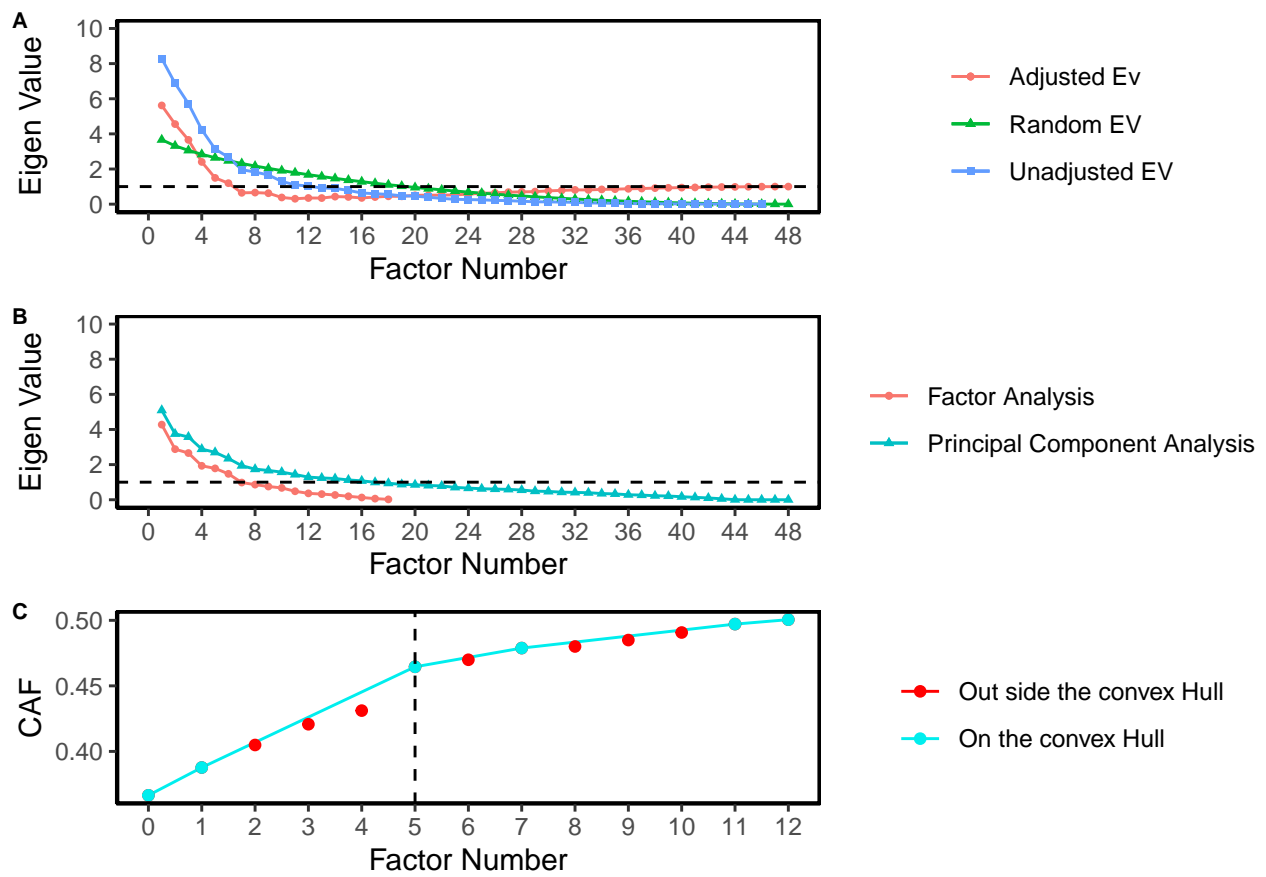


Figure 3. Factor Identification (A) Parallel analysis (B) Scree Plot (C) Hull Method

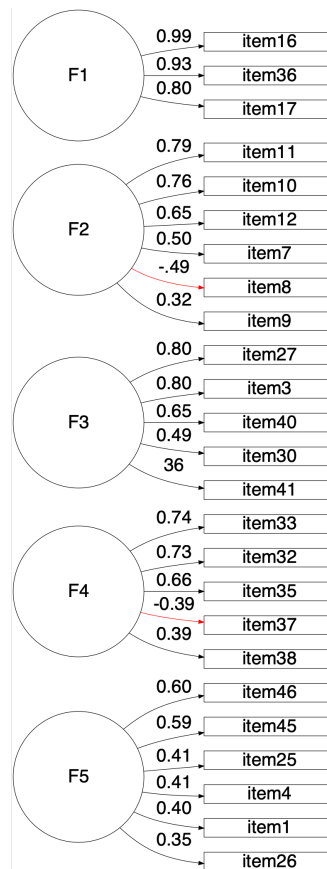


Figure 4. Five Factor Solution

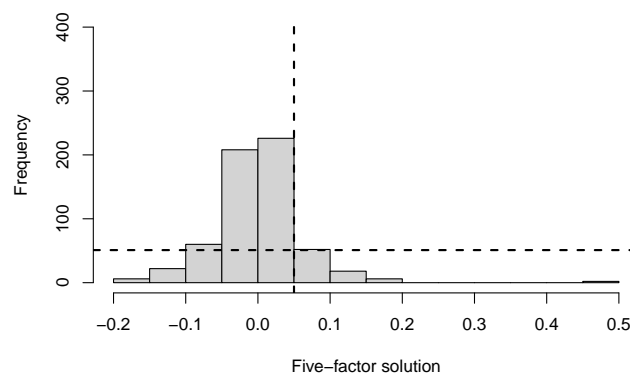


Figure 5. Histogram of residulas: five-factor solution

































































































Summary results of LEBA (N =690)											
Items	Summary Statistics				Graphics		Response Pattern				
LEBA Items	n	Mean	Median	SD	Histogram <sup>1</sup>	Density <sup>2</sup>	Never	Rarely	Sometimes	Often	Always
EFA (n = 428)											
Item01	428	2.3	2.0	1.4			22.20% (95)	12.38% (53)	10.51% (45)	12.62% (54)	42.29% (181)
Item03	428	3.4	4.0	1.4			11.45% (49)	31.07% (133)	24.30% (104)	17.29% (74)	15.89% (68)
Item04	428	1.5	1.0	1.2			3.50% (15)	2.10% (9)	8.18% (35)	2.10% (9)	84.11% (360)
Item07	428	2.3	2.0	1.2			27.80% (119)	12.38% (53)	6.54% (28)	17.29% (74)	35.98% (154)
Item08	428	3.0	3.0	1.2			25.93% (111)	22.20% (95)	13.79% (59)	27.80% (119)	10.28% (44)
Item09	428	2.9	3.0	1.0			19.63% (84)	22.43% (96)	5.84% (25)	41.82% (179)	10.28% (44)
Item10	428	2.7	3.0	1.0			31.31% (134)	21.96% (94)	3.50% (15)	31.31% (134)	11.92% (51)
Item11	428	2.2	2.0	0.9			46.26% (198)	7.01% (30)	1.17% (5)	23.13% (99)	22.43% (96)
Item12	428	2.4	2.0	1.2			29.67% (127)	12.15% (52)	6.78% (29)	21.50% (92)	29.91% (128)
Item16	428	1.6	1.0	1.2			4.21% (18)	4.67% (20)	7.48% (32)	3.97% (17)	79.67% (341)
Item17	428	1.5	1.0	1.2			3.27% (14)	3.27% (14)	7.71% (33)	5.14% (22)	80.61% (345)
Item25	428	2.6	3.0	1.4			13.79% (59)	17.99% (77)	11.68% (50)	22.20% (95)	34.35% (147)
Item26	428	3.7	4.0	1.3			10.98% (47)	23.36% (100)	38.32% (164)	20.09% (86)	7.24% (31)
Item27	428	3.8	4.0	1.3			11.21% (48)	30.37% (130)	38.79% (166)	11.21% (48)	8.41% (36)
Item30	428	1.5	1.0	1.1			3.27% (14)	5.37% (23)	4.67% (20)	4.91% (21)	81.78% (350)
Item32	428	3.6	4.0	1.6			7.01% (30)	14.95% (64)	46.73% (200)	8.18% (35)	23.13% (99)
Item33	428	3.6	4.0	1.6			7.01% (30)	14.49% (62)	49.30% (211)	7.24% (31)	21.96% (94)
Item35	428	3.9	5.0	1.7			1.87% (8)	9.35% (40)	62.15% (266)	3.74% (16)	22.90% (98)
Item36	428	1.5	1.0	1.3			3.04% (13)	2.34% (10)	9.35% (40)	3.04% (13)	82.24% (352)
Item37	428	2.3	2.0	1.3			23.36% (100)	10.98% (47)	7.24% (31)	20.09% (86)	38.32% (164)
Item38	428	4.3	5.0	1.1			3.50% (15)	27.57% (118)	58.18% (249)	5.37% (23)	5.37% (23)
Item40	428	2.2	2.0	1.2			25.00% (107)	11.45% (49)	4.44% (19)	19.63% (84)	39.49% (169)
Item41	428	1.3	1.0	0.8			4.67% (20)	3.04% (13)	1.17% (5)	6.07% (26)	85.05% (364)
Item45	428	2.2	1.0	1.5			7.01% (30)	11.92% (51)	11.68% (50)	16.36% (70)	53.04% (227)
Item46	428	1.8	1.0	1.2			7.71% (33)	8.88% (38)	4.67% (20)	11.68% (50)	67.06% (287)
CFA (n =262)											
Item01	262	2.3	2.0	1.4			22.52% (59)	10.69% (28)	11.83% (31)	14.50% (38)	40.46% (106)
Item03	262	3.7	4.0	1.3			7.25% (19)	28.24% (74)	35.11% (92)	17.56% (46)	11.83% (31)
Item04	262	1.3	1.0	0.8			2.29% (6)	3.05% (8)	1.91% (5)	3.44% (9)	89.31% (234)
Item07	262	2.1	2.0	1.2			23.66% (62)	14.12% (37)	4.58% (12)	14.50% (38)	43.13% (113)
Item08	262	3.0	3.0	1.2			32.06% (84)	22.90% (60)	14.12% (37)	20.99% (55)	9.92% (26)
Item09	262	2.9	3.0	1.1			22.14% (58)	26.34% (69)	4.20% (11)	34.35% (90)	12.98% (34)
Item10	262	2.6	3.0	1.1			29.39% (77)	21.37% (56)	2.67% (7)	29.01% (76)	17.56% (46)
Item11	262	2.1	2.0	0.9			46.56% (122)	5.34% (14)	1.91% (5)	20.23% (53)	25.95% (68)
Item12	262	2.3	2.0	1.2			30.92% (81)	11.45% (30)	6.49% (17)	19.08% (50)	32.06% (84)
Item16	262	1.6	1.0	1.3			3.44% (9)	5.73% (15)	8.40% (22)	4.20% (11)	78.24% (205)
Item17	262	1.6	1.0	1.2			3.44% (9)	2.67% (7)	8.40% (22)	5.34% (14)	80.15% (210)
Item25	262	2.5	2.0	1.4			18.32% (48)	16.79% (44)	10.31% (27)	21.76% (57)	32.82% (86)
Item27	262	4.0	4.0	1.2			7.25% (19)	33.59% (88)	45.04% (118)	8.02% (21)	6.11% (16)
Item30	262	1.4	1.0	1.1			2.67% (7)	6.11% (16)	3.44% (9)	4.20% (11)	83.59% (219)
Item32	262	3.4	4.0	1.7			4.20% (11)	16.79% (44)	41.60% (109)	11.45% (30)	25.95% (68)
Item33	262	3.1	3.0	1.7			6.11% (16)	14.12% (37)	35.50% (93)	11.83% (31)	32.44% (85)
Item35	262	3.6	5.0	1.8			2.67% (7)	6.49% (17)	56.11% (147)	7.25% (19)	27.48% (72)
Item36	262	1.6	1.0	1.3			3.44% (9)	3.44% (9)	9.54% (25)	3.05% (8)	80.53% (211)
Item38	262	4.3	5.0	1.1			7.53% (20)	21.37% (56)	60.31% (158)	6.49% (17)	4.20% (11)
Item40	262	2.5	2.0	1.3			27.10% (71)	12.21% (32)	11.07% (29)	18.70% (49)	30.92% (81)
Item41	262	1.2	1.0	0.7			3.82% (10)	2.67% (7)	1.15% (3)	2.29% (6)	90.08% (236)
Item45	262	2.0	1.0	1.4			5.34% (14)	11.83% (31)	9.16% (24)	9.54% (25)	64.12% (168)
Item46	262	1.6	1.0	1.2			2.67% (7)	9.54% (25)	4.20% (11)	8.02% (21)	75.57% (198)
<sup>1</sup> Histogram											
<sup>2</sup> Density											

Figure 6

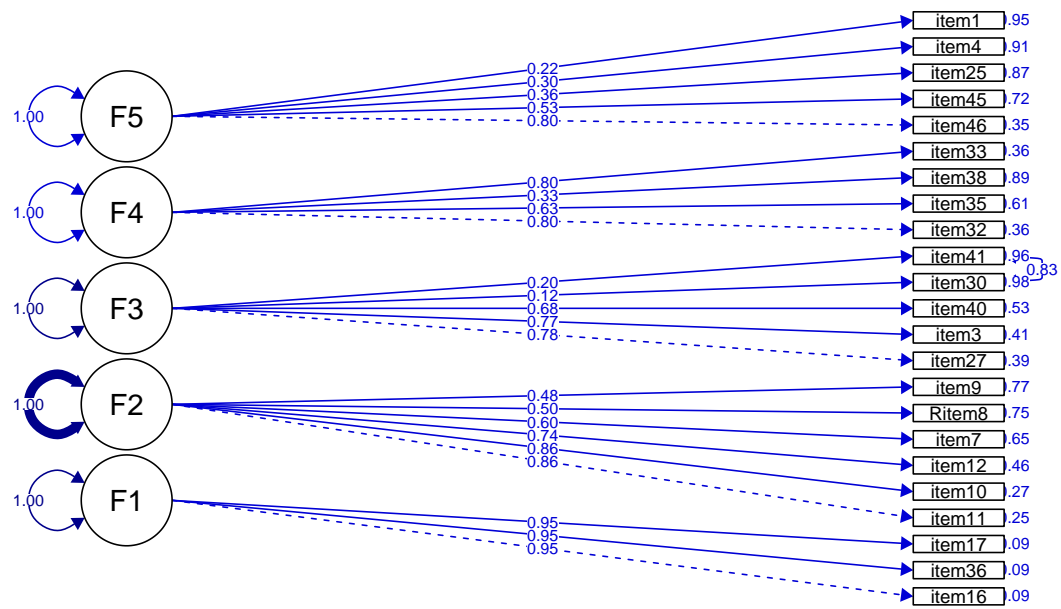


Figure 7. (A) Five Factor Model of LEBA

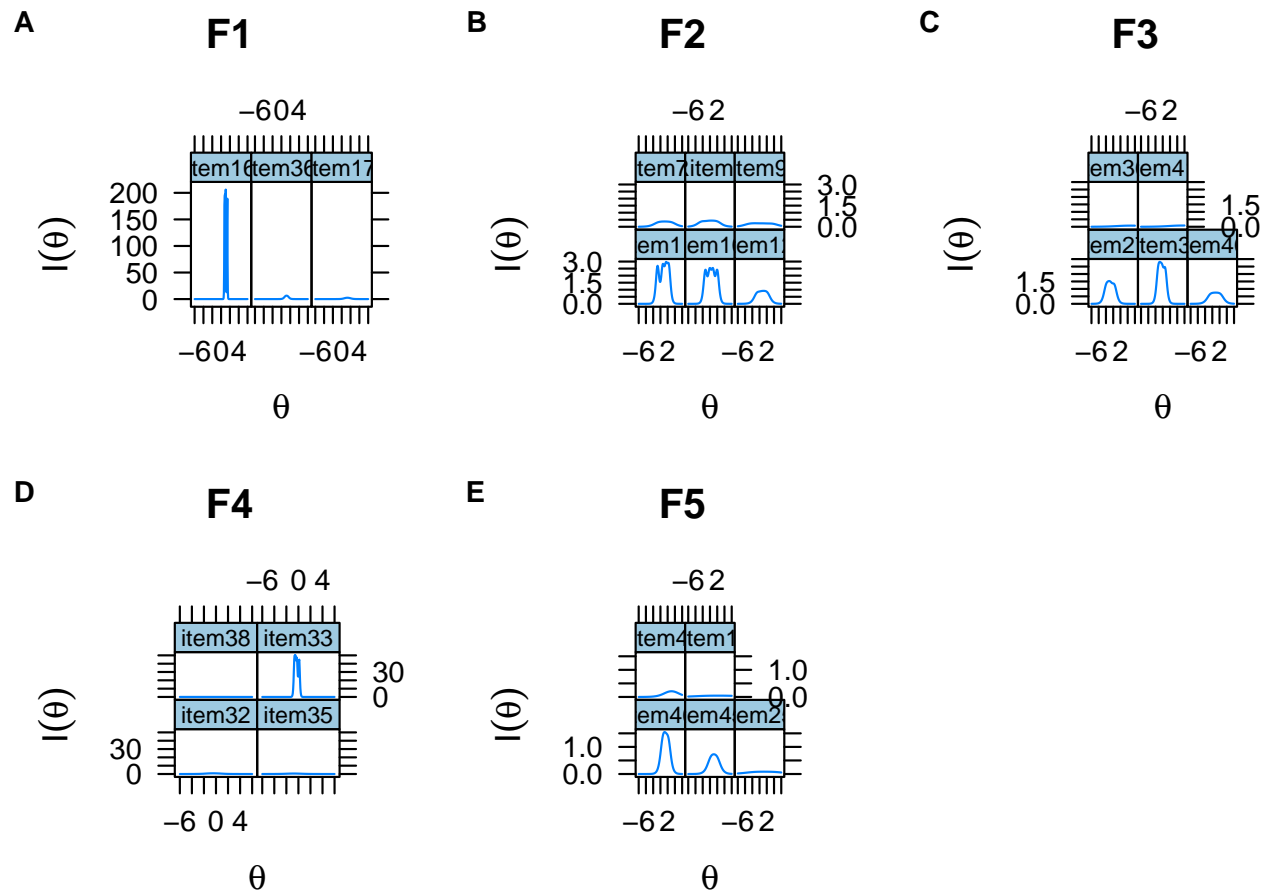


Figure 8. Item information curves (A) blue filter (B) natural light (C) smart device (D) sleep environment (E) electric light



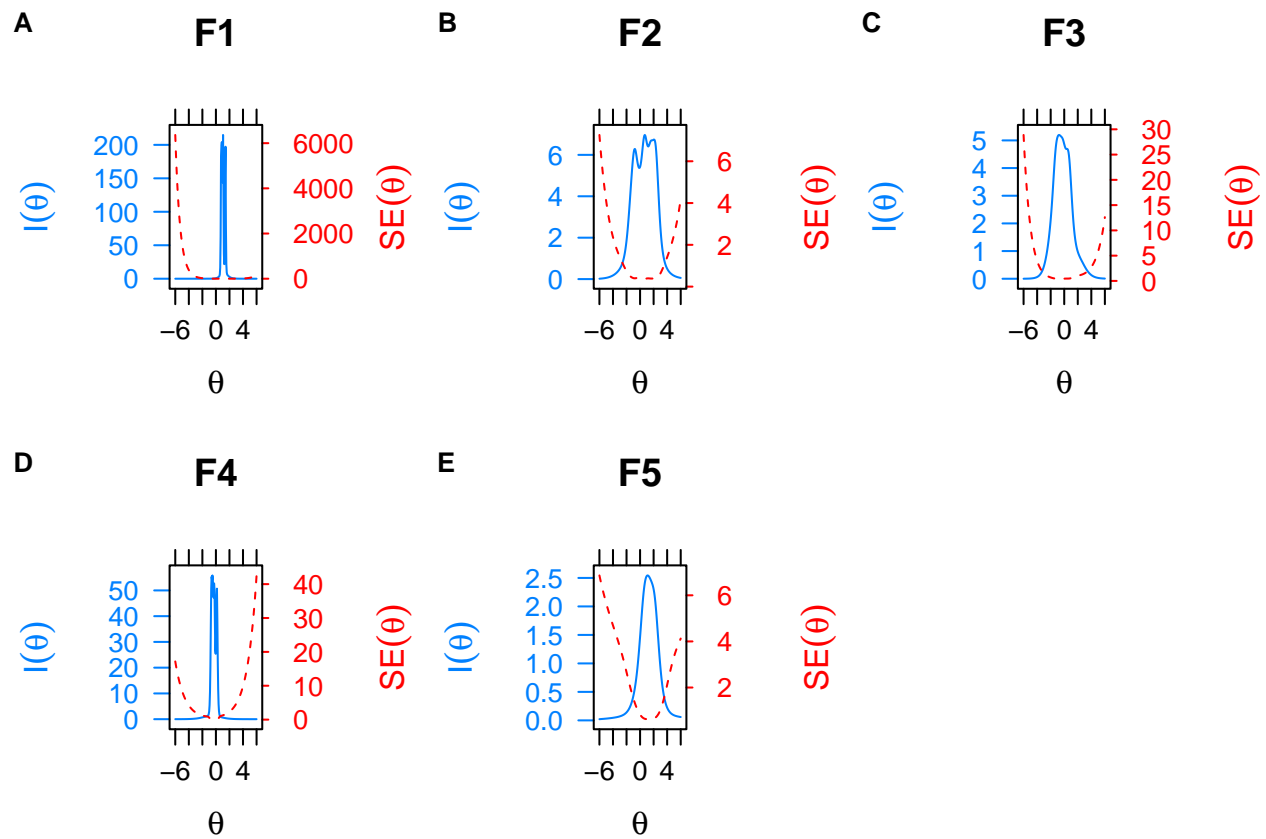


Figure 9. Test information curves (A) blue filter (B) natural light (C) smart device (D) sleep environment (E) electric light

## Appendix A

Table A1

*Factor loadings and communality of the retained items(Minmum Residual)*

item	MR1	MR2	MR3	MR4	MR5	Communality	Uniqueness
item16	1					0.996	0.004
item36	0.94					0.897	0.103
item17	0.8					0.658	0.342
item11		0.79				0.642	0.358
item10		0.76				0.592	0.408
item12		0.65				0.465	0.535
item7		0.5				0.267	0.733
item8		-0.49				0.252	0.748
item9		0.32				0.113	0.887
item27			0.8			0.659	0.341
item3			0.8			0.683	0.317
item40			0.65			0.464	0.536
item30			0.45			0.353	0.647
item41			0.36			0.329	0.671
item33				0.74		0.555	0.445
item32				0.73		0.623	0.377
item35				0.66		0.455	0.545
item37				-0.39		0.175	0.825
item38				0.38		0.178	0.822
item46					0.6	0.422	0.578
item45					0.59	0.374	0.626
item25					0.41	0.193	0.807
item4					0.41	0.219	0.781
item1					0.4	0.17	0.83
item26					0.35	0.165	0.835
% of Variance	0.1	0.1	0.09	0.08	0.06		

*Note.* Only loading higher than .30 is reported



Appendix B

Factor Analysis with Unmerged Response Option

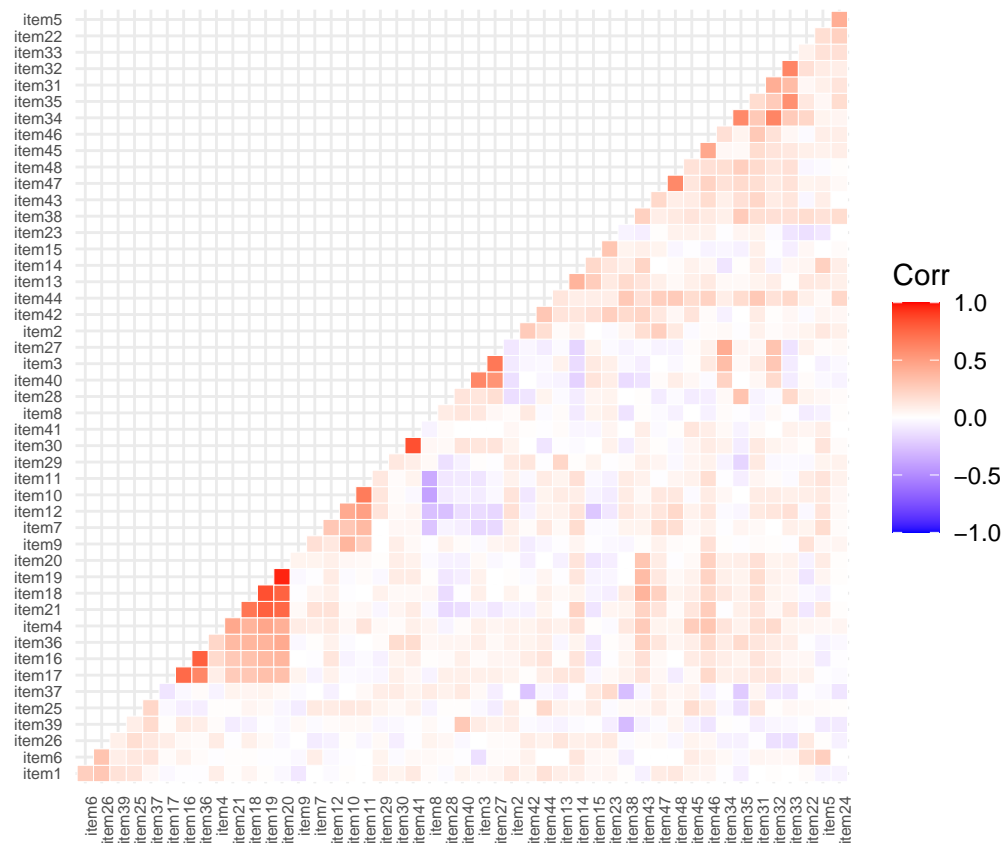


Figure B1. Correlation plot of the items

Horn’s parallel analysis with 500 iterations indicated a five-factor solution. However, Scree plot and the MAP method suggested 6-factor solution. five-factor solution . As a result, we tested both five-factor and six-factor solutions.

**Five Factor Solution[Unmerged Responses] (24 Items)**

**F1**

- I use light therapy applying a blue light box.
- I use light therapy applying a light visor.
- I use light therapy applying a white light box.
- I use light therapy applying another form of light device.

---

**Five Factor Solution[Unmerged Responses] (24 Items)**

---

I use an alarm with a dawn simulation light.

**F2**

I spend more than 3 hours per day (in total) outside.

I spend between 1 and 3 hours per day (in total) outside.

I spend as much time outside as possible.

I spend 30 minutes or less per day (in total) outside.

I go for a walk or exercise outside within 2 hours after waking up.

I spend between 30 minutes and 1 hour per day (in total) outside.

**F3**

I look at my mobile phone screen immediately after waking up.

I use my mobile phone within 1 hour before attempting to fall asleep.

I check my phone when I wake up at night.

**F4**

I use a blue-filter app on my computer screen within 1 hour before attempting to fall asleep.

I seek out knowledge on how to improve my light exposure.

I dim my computer screen within 1 hour before attempting to fall asleep.

I discuss the effects of light on my body with other people.

I modify my light environment to match my current needs.

I dim my room light within 1 hour before attempting to fall asleep.

I use as little light as possible when I get up during the night.

**F5**

I wear blue-filtering, orange-tinted, and/or red-tinted glasses indoors during the day.

I wear blue-filtering, orange-tinted, and/or red-tinted glasses outdoors during the day.

---

**Five Factor Solution[Unmerged Responses] (24 Items)**

---

I wear blue-filtering, orange-tinted, and/or red-tinted glasses within 1 hour before attempting to fall asleep.

---

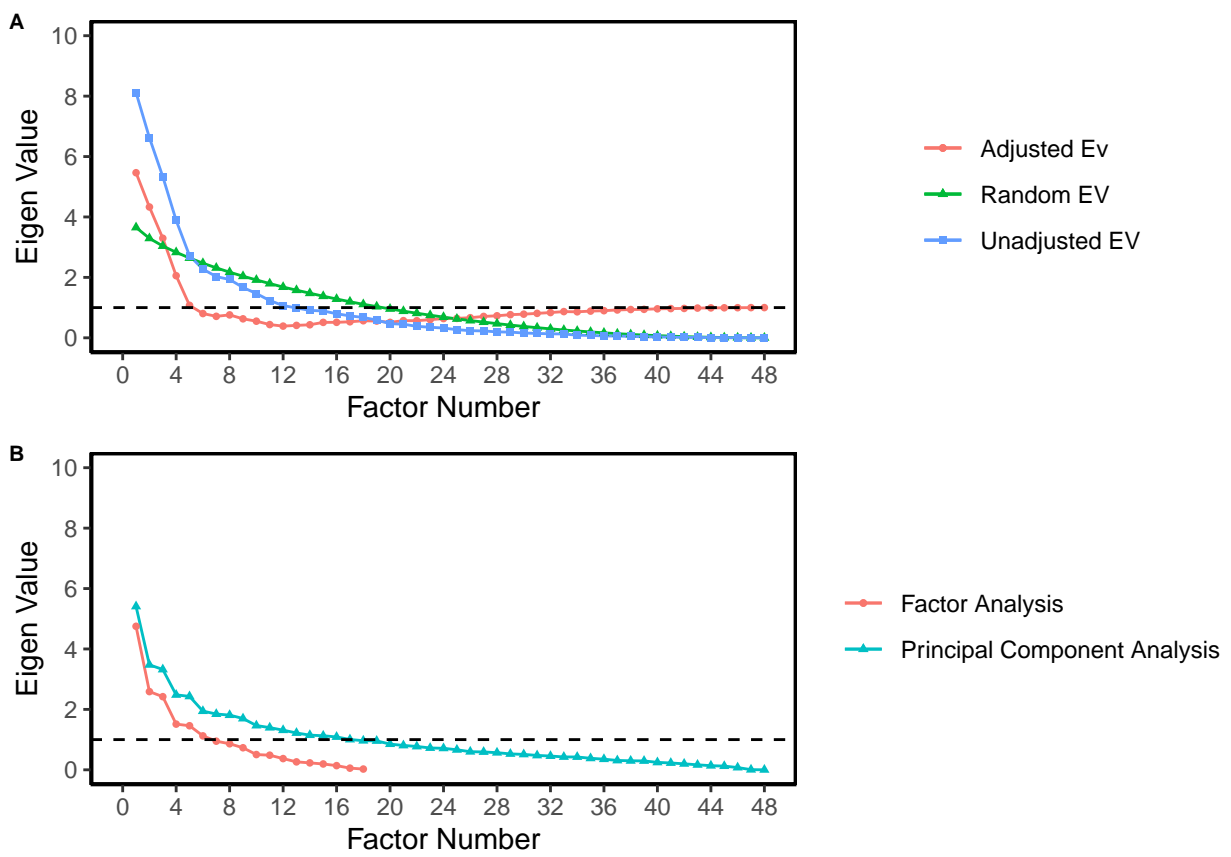


Figure B2. Factor Identification (A) Parallel analysis (B) Scree Plot



Table B1

*Descriptive Statistics for Unmerged response options*

	Mean	SD	Skew	Kurtosis	Shapiro-Wilk Statistics	Item-Total Correlation
Item1	2.16	1.51	0.49	-0.86	0.90*	.21
Item2	2.76	1.75	-0.10	-1.42	0.88*	.20
Item3	3.34	1.43	-0.58	-0.77	0.88*	.18
Item4	1.30	1.31	1.93	2.92	0.62*	.32
Item5	3.95	1.56	-1.42	0.75	0.70*	.19
Item6	2.70	1.66	0.02	-1.33	0.90*	.18
Item7	2.23	1.28	0.60	-0.59	0.89*	.18
Item8	2.95	1.24	-0.19	-0.70	0.93*	-.07
Item9	2.92	1.09	-0.37	0.11	0.91*	.14
Item10	2.73	1.07	-0.03	-0.52	0.92*	.27
Item11	2.17	0.93	0.44	0.20	0.89*	.25
Item12	2.34	1.26	0.46	-0.58	0.91*	.24
Item13	2.71	1.49	0.14	-1.29	0.89*	.28
Item14	2.11	1.34	0.68	-0.78	0.84*	.24
Item15	3.26	1.11	-0.34	-0.21	0.91*	.11
Item16	1.46	1.31	1.71	1.90	0.65*	.33
Item17	1.43	1.30	1.76	2.12	0.64*	.30
Item18	0.92	0.67	2.00	9.41	0.62*	.32
Item19	0.85	0.56	1.71	10.74	0.55*	.34
Item20	0.83	0.54	1.76	13.92	0.53*	.31
Item21	0.94	0.75	2.46	10.66	0.58*	.27
Item22	3.57	1.08	-0.72	0.08	0.88*	.19
Item23	2.53	1.31	0.22	-0.91	0.92*	.11
Item24	4.13	1.01	-1.39	2.01	0.78*	.19
Item25	2.57	1.43	0.22	-1.23	0.88*	.17
Item26	2.22	1.22	0.52	0.62	0.82*	.16

Table B2

*Factor loadings and communality of the retained items [Unmerged Responses]*

item	PA1	PA2	PA5	PA3	PA4	Communality	Uniqueness	Complexity
item19	0.99					1.007	-0.007	1.058
item20	0.91					0.874	0.126	1.114
item18	0.82					0.711	0.289	1.123
item21	0.8					0.683	0.317	1.163
item4	0.47					0.25	0.75	1.298
item11		0.83				0.687	0.313	1.007
item10		0.81				0.67	0.33	1.031
item12		0.56				0.371	0.629	1.374
item8		-0.44				0.206	0.794	1.106
item7		0.42				0.226	0.774	1.614
item9		0.33				0.115	0.885	1.1
item16			0.95			0.946	0.054	1.097
item17			0.74			0.595	0.405	1.168
item36	0.3		0.73			0.653	0.347	1.431
item3				0.85		0.746	0.254	1.048
item27				0.78		0.624	0.376	1.028
item40				0.71		0.512	0.488	1.05
item35					0.58	0.351	0.649	1.091
item48					0.57	0.354	0.646	1.144
item33					0.55	0.32	0.68	1.085
item47					0.52	0.294	0.706	1.186
item44					0.45	0.216	0.784	1.145
item31					0.41	0.206	0.794	1.477
item38					0.33	0.129	0.871	1.317
% of Variance	0.15	0.09	0.09	0.08	0.08			

*Note.* Only loading higher than .30 is reported

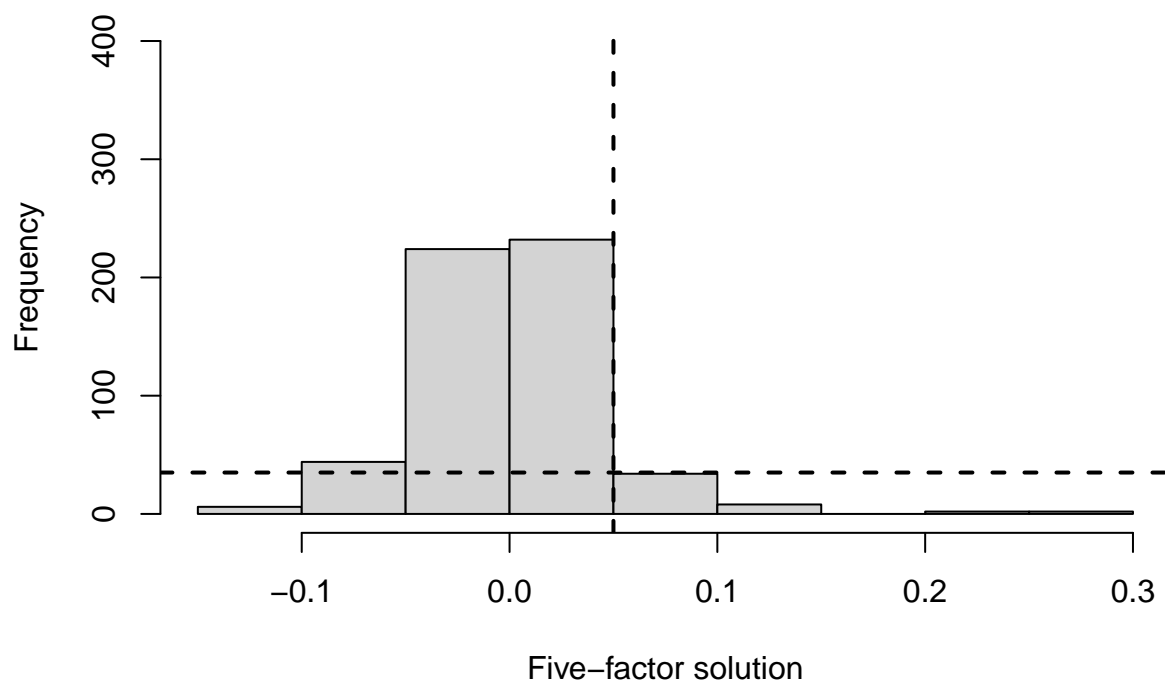


Figure B3. Histogram of residulas: five-factor solution

## Appendix C

**Disclaimer:** This is a non-public version of LEBA (dated November 12, 2021) and still a work in progress. Please do not distribute!

LEBA captures light exposure-related behaviours on a 5 point Likert type scale ranging from 1 to 5 (Never/Does not apply/I don't know = 1; Rarely = 2; Sometimes = 3; Often = 4; Always = 5). The score of each factor is calculated by the summation of scores of items belonging to the corresponding factor. The following instruction is given before displaying the items: "Please indicate how often you performed the following behaviours in the past 4 weeks."

Appendix D  
LEBA Long Form (23 Items)

	Items	Never/Does not apply/I don't know	Rarely	Sometimes	Often	Always
1	I wear blue-filtering, orange-tinted, and/or red-tinted glasses indoors during the day.					
2	I wear blue-filtering, orange-tinted, and/or red-tinted glasses outdoors during the day.					
3	I wear blue-filtering, orange-tinted, and/or red-tinted glasses within 1 hour before attempting to fall asleep.					
4	I spend 30 minutes or less per day (in total) outside.					

	Items	Never/Does not apply/I don't know	Rarely	Sometimes	Often	Always
5	I spend between 1 and 3 hours per day (in total) outside.					
6	I spend between 30 minutes and 1 hour per day (in total) outside.					
7	I spend more than 3 hours per day (in total) outside.					
8	I spend as much time outside as possible.					
9	I go for a walk or exercise outside within 2 hours after waking up.					
10	I use my mobile phone within 1 hour before attempting to fall asleep.					

	Items	Never/Does not apply/I don't know	Rarely	Sometimes	Often	Always
11	I look at my mobile phone screen immediately after waking up.					
12	I check my phone when I wake up at night.					
13	I look at my smartwatch within 1 hour before attempting to fall asleep.					
14	I look at my smartwatch when I wake up at night.					
15	I dim my mobile phone screen within 1 hour before attempting to fall asleep.					
16	I use a blue-filter app on my computer screen within 1 hour before attempting to fall asleep.					

	Items	Never/Does not apply/I don't know	Rarely	Sometimes	Often	Always
17	I use as little light as possible when I get up during the night.					
18	I dim my computer screen within 1 hour before attempting to fall asleep.					
19	I use tunable lights to create a healthy light environment.					
20	I use LEDs to create a healthy light environment.					
21	I use a desk lamp when I do focused work.					
22	I use an alarm with a dawn simulation light.					
23	I turn on the lights immediately after waking up.					



## 615 Latent Structure, Reliability and Structural Validity

616 The long form of LEBA consists 23 items with five factors.

Factor names	Items	Reliability Coefficients:	Reliability Coefficients:
		McDonald's Omega	Cronbach's alpha
<b>F1: Wearing blue light filters</b>	1-3	.93	.90
F2: Spending time outdoors	4-9 (Item 4 is reversed)	.80	.78
<b>F3: Using phone and smartwatch in bed</b>	10-14	.61	.62
<b>F4: Using light before bedtime</b>	15-18	.72	.62
<b>F5: Using light in the morning and during daytime</b>	19-23	.45	.41
		.73(Total scale)	

617 LEBA -long form showed satisfactory structural validity (CFI =.97; TLI = .96; RMSEA = .05[.04-.06, 90% CI]; SRMR =  
618 .09).

619 **How to cite:**

Appendix E  
LEBA Short Form (17 Items)

	Short Form (17 Items)	Never/Does not apply/I don't know	Rarely	Sometimes	Often	Always
01	I wear blue-filtering, orange-tinted, and/or red-tinted glasses indoors during the day.					
02	I wear blue-filtering, orange-tinted, and/or red-tinted glasses outdoors during the day.					
03	I wear blue-filtering, orange-tinted, and/or red-tinted glasses within 1 hour before attempting to fall asleep.					
04	I spend 30 minutes or less per day (in total) outside.					

	Short Form (17 Items)	Never/Does not apply/I don't know	Rarely	Sometimes	Often	Always
05	I spend between 1 and 3 hours per day (in total) outside.					
06	I spend more than 3 hours per day (in total) outside.					
07	I spend as much time outside as possible.					
08	I go for a walk or exercise outside within 2 hours after waking up.					
09	I use my mobile phone within 1 hour before attempting to fall asleep.					
10	I look at my mobile phone screen immediately after waking up.					
11	I check my phone when I wake up at night.					

	Short Form (17 Items)	Never/Does not apply/I don't know	Rarely	Sometimes	Often	Always
12	I dim my mobile phone screen within 1 hour before attempting to fall asleep.					
13	I use a blue-filter app on my computer screen within 1 hour before attempting to fall asleep.					
14	I dim my computer screen within 1 hour before attempting to fall asleep.					
15	I use tunable lights to create a healthy light environment.					
16	I use LEDs to create a healthy light environment.					
17	I use an alarm with a dawn simulation light.					

## 620 Latent Structure, Reliability and Structural Validity

621 The short form of LEBA consists 23 items with five factors.

Factor names	Items
<b>F1: Wearing blue light filters</b>	1-3
<b>F2: Spending time outdoors</b>	4-8 (Item 4 is reversed)
<b>F3: Using phone and smart-watch in bed</b>	9-11
<b>F4: Using light before bedtime</b>	12-14
<b>F5: Using light in the morning and during daytime</b>	15-17

622 **How to cite:**