

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

Mushfiquel Anwar Siraji<sup>1, \*</sup>, Rafael Robert Lazar<sup>2, 3, \*</sup>, Juliëtte van Duijnhoven<sup>4, 5</sup>, Luc Schlangen<sup>5, 6</sup>, Shamsul Haque<sup>1</sup>, Vineetha Kalavally<sup>7</sup>, Céline Vetter<sup>8, 9</sup>, Gena Glickman<sup>10</sup>, Karin Smolders<sup>5, 6</sup>, & Manuel Spitschan<sup>11, 12, 13</sup>

<sup>1</sup> Monash University, Department of Psychology, Jeffrey Cheah School of Medicine and Health Sciences, Malaysia

<sup>2</sup> Psychiatric Hospital of the University of Basel (UPK), Centre for Chronobiology, Basel, Switzerland

<sup>3</sup> University of Basel, Transfaculty Research Platform Molecular and Cognitive Neurosciences, Basel, Switzerland

<sup>4</sup> Eindhoven University of Technology, Department of the Built Environment, Building Lighting, Eindhoven, Netherlands

<sup>5</sup> Eindhoven University of Technology, Intelligent Lighting Institute, Eindhoven, Netherlands

<sup>6</sup> Eindhoven University of Technology, Department of Industrial Engineering and Innovation Sciences, Human-Technology Interaction, Eindhoven, Netherlands

<sup>7</sup> Monash University, Department of Electrical and Computer Systems Engineering, Selangor, Malaysia

<sup>8</sup> University of Colorado Boulder, Department of Integrative Physiology, Boulder, USA

<sup>9</sup> XIMES GmbH, Vienna, Austria

<sup>10</sup> Uniformed Services University of the Health Sciences, Department of Psychiatry,  
Bethesda, USA

<sup>11</sup> Max Planck Institute for Biological Cybernetics, Tübingen, Germany

<sup>12</sup> Technical University of Munich, Department of Sport and Health Sciences (TUM SG),  
Munich, Germany

<sup>13</sup> University of Oxford, Department of Experimental Psychology, Oxford, United Kingdom

\* Joint first author

This research is supported by funding from the Wellcome Trust (204686/Z/16/Z), the European Training Network LIGHTCAP (project number 860613) under the Marie Skłodowska-Curie actions framework H2020-MSCA-ITN-2019, the BioClock project (number 1292.19.077) of the research program Dutch Research Agenda: Onderzoek op Routes door Consortia (NWA-ORC) which is (partly) financed by the Dutch Research Council (NWO), and the European Union and the nationals contributing in the context of the ECSEL Joint Undertaking programme (2021-2024) under the grant #101007319.

The authors made the following contributions. Mushfiqul Anwar Siraji: Formal Analysis, Visualization, Writing – original draft, Writing – review & editing;; Rafael Robert Lazar: Data curation, Investigation, Project administration, Visualization, Writing – original draft, Writing – review & editing;; Juliëtte van Duijnhoven: Conceptualization, Methodology, Investigation, Writing – review & editing; Luc Schlangen: Conceptualization, Methodology, Investigation, Writing – review & editing; Shamsul Haque: Conceptualization, Supervision, Writing – review & editing; Vineetha Kalavally: Supervision, Writing – review & editing; Céline Vetter: Conceptualization, Writing – review & editing; Gena Glickman: Conceptualization, Methodology, Writing – review & editing; Karin Smolders: Conceptualization, Methodology, Writing – review & editing; Manuel Spitschan: Conceptualization, Data curation, Investigation, Project administration, Visualization, Methodology, Writing – original draft, Writing – review & editing.

## Abstract

Light exposure is an important driver of health and well-being. Many aspects of light exposure are modulated by our behaviour. How these light-related behaviours can be shaped to optimise personal light exposure is currently unknown. Here, we present a novel, self-reported and psychometrically validated instrument to capture light exposure-related behaviour, the Light Exposure Behavior Assessment (LEBA). An expert panel prepared the initial 48 item pool. Responses to these items were then collected in an online survey producing responses from an international sample (690 completed responses, 74 countries, 28 time zones). Exploratory factor analysis on an initial subset of our sample (n=428) rendered a five-factor solution with 25 items (Wearing blue light filters, spending time outdoors, using phone and smart-watch in bed, using light before bedtime, using light in the morning and during daytime). Confirmatory factor analysis on another subset of participants (n=262) yielded the best fit for the five-factor solution after discarding another two items (CFI=0.97, TLI=0.96, RMSEA=0.05, SRMR=0.09). The internal consistency reliability coefficient for the total instrument was McDonald's omega =0.73. Measurement model invariance analysis between native and non-native English speakers showed our model attained the highest level of invariance (residual invariance; CFI=0.95, TLI =0.95, RMSEA=0.05). Lastly, a short form of LEBA (n=18) was developed using Item Response Theory on the complete sample (n=690). The psychometric properties of the LEBA instrument indicate the usability to measure the light exposure-related behaviours across a variety of settings and may offer a scalable solution to characterize light exposure-related behaviours in remote samples.

**Keywords:** light exposure, light-related behaviours, non-visual effects of light, psychometrics

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.

## 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

All code and data underlying this article is available on a public GitHub repository (<https://github.com/leba-instrument/leba-manuscript>).

## 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; Daniel J. 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 2 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 ten most common country + timezone combinations included:

- United States - America/New\_York (UTC -04:00): 63 (9.1%)
- United Kingdom - Europe/London (UTC): 57 (8.3%)
- Germany - Europe/Berlin (UTC +01:00): 53 (7.7%)
- India - Asia/Kolkata (UTC +05:30): 38 (5.5%)
- United States - America/Los\_Angeles (UTC -07:00): 37 (5.4%)
- United States - America/Chicago (UTC -05:00): 30 (4.3%)
- France - Europe/Paris (UTC +01:00): 22 (3.2%)
- Switzerland - Europe/Zurich (UTC +01:00): 21 (3.0%)
- Brazil - America/Sao\_Paulo (UTC -03:00): 19 (2.8%)
- Netherlands - Europe/Amsterdam (UTC +01:00): 19 (2.8%)

For a complete 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 on vacation)” 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  $\chi^2$  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  $q \geq 0.2$ ).

## Item Generation

To ensure construct adequacy we thoroughly assessed the current status of literature and identified a variety of light exposure related scales. However, no scales specifically measuring the behavioral component of light exposure were found (Table 1). Consequentially we pursued to introduce a new openly available scale to address this research gap. For this purpose an expert researcher panel from the fields of

chronobiology, light research, neuroscience and psychology (including seven of the authors, see authors roles) generated and collected preliminary item ideas. Special attention was paid to design items circumscribed to assess light exposure *behavior* as opposed to subjective measurements of the light environment (cf.(Eklund & Boyce, 1996) & (Dianat, Sedghi, Bagherzade, Jafarabadi, & Stedmon, 2013)) and semi-quantitative assessments of light sources' illuminance (cf. (Bajaj, Rosner, Lockley, & Schernhammer, 2011)) in order to maintain content validity. In a collective effort the generated items were then peer-reviewed, amended, unified, and complemented with a suitable response scale (5 point Likert-scale ranging from 1 "Never/Does not apply/I don't know" to 5 "Always"). This process was finalized when all experts were in agreement, resulting in 48 items to implement in the data collection.

## **Analytic Strategies**

Figure 1 summarizes the steps of our psychometric analysis. In our analysis we used R (version 4.1.0), with several R packages. Initially, our tool had six point Likert type response format (0:Does not apply/I don't know; 1:Never, 2:Rarely; 3:Sometimes; 4:Often; 5:Always). Our purpose was to capture light exposure related behavior and these two response options: "Does not apply/I don't know" and "Never" were providing similar information. As such we decided to collapse them into one, making it a 5 point Likert type response format. Necessary assumptions of EFA, including sample adequacy, normality assumptions, quality of correlation matrix, were assessed. Our data violated both the univariate and multivariate normality assumptions. Due to these violations and the ordinal nature of our response data, we used polychoric correlation matrix (Desjardins & Bulut, 2018) for the EFA. We employed principal axis (PA) as factor extraction method with varimax rotation. PA is robust to the normality assumption violations (Watkins, 2020). The obtained latent structure was confirmed by another factor extraction method: "the minimum residuals extraction" as well. We used a combination

of 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 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 confirmed the latent structure obtained in the EFA by conducting a categorical "Confirmatory Factor Analysis" (CFA) using "robust weighted least square estimator" (WLSMV). We established the measurement invariance of our tool across the native and non-native English speakers using structural equation model framework. To assess the possible semantic overlap of our tool with the existing tools, we sought to "Semantic Scale Network" (Rosenbusch, Wanders, & Pit, 2020). To assess the possible semantic overlap of our tool with the existing tools, we sought to "Semantic Scale Network" (Rosenbusch et al., 2020). Lastly, we sought "Item Response Theory" (IRT) based analysis on developing a short form of LEBA. We also conducted psychometric analysis on non-merged response options data (Supp. Table C2) and rejected the latent structure obtained as the factors were less interpretable.

## Results

### Item Analysis

Table 3 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 (Desjardins & Bulut, 2018). The corrected item-total correlation ranges between .03 -.48. However, no item was discarded based on descriptive statistics or item analysis.

### Exploratory Factor Analysis

Sampling adequacy was checked using Kaiser-Meyer-Olkin (KMO) measures of sampling adequacy (Kaiser, 1974) . The overall KMO vale for 48 items was 0.63 which was above the cutoff value (.50) indicating a mediocre sample (Hutcheson, 1999). 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 absolute value of inter-item correlation ranged between .00 to .91. Figure 2 depicts the correlation matrix.

Scree plot ( Figure 3) suggested a six-factor solution. Horn's parallel analysis (Horn, 1965) with 500 iterations also indicated a six-factor solution. However, the minimum average partial (MAP) method (Table A1) (Velicer, 1976) and Hull method (Lorenzo-Seva et al., 2011) ( Figure 3) suggested a five-factor solution. As a result, we tested both five-factor and six-factor solutions.

With the initial 48 items we conducted three rounds of EFA and gradually discarded problematic items. (cross-loading items and poor factor loading ( $<.30$ ) items). Finally, a five-factor EFA solution with 25 items was accepted with low RMSR = 0.08 (Brown, 2015), all factor-loading higher than .30 and no cross-loading greater than .30. We further confirmed this five-factor latent structure by another EFA using varimax rotation with a minimum residual extraction method (Sup.Table B1). Table 4 displays the factor-loading (structural coefficients) and communality of the items. The absolute value of the factor-loading ranged from .49 to .99 indicating strong coefficients. The

commonalities ranged between .11 to .99. Figure 4 depicts the obtained five-five factor structure. However, the histogram of the absolute values of non-redundant residual-correlations (Sup.Figure A1) showed 26% correlations were greater than the absolute value of .05, indicating a possible under-factoring. (Desjardins & Bulut, 2018). Subsequently, we fitted a six-factor solution. However, a factor emerged with only two salient variables, thus disqualifying the six-factor solution (Sup.Table C1). Internal consistency reliability coefficient Cronbach's alpha assumes all the factor-loadings of the items under a factor are equal (Graham, 2006; Novick & Lewis, 1967) which is not the case in our sample. Additionally Cronbach's alpha coefficient has a tendency to deflate the estimates for Likert type data as the calculation is based on pearson-correlation matrix which requires that response data should be in continuous of nature (Gadermann, Guhn, & Zumbo, 2012; Zumbo, Gadermann, & Zeisser, 2007). Subsequently to get better estimates of reliability we reported ordinal alpha which used polychoric-correlation matrix and assumed that the responses data were ordered in nature instead of continuous (Zumbo et al., 2007). Ordinal alpha coefficient value ranges from 0 to 1 and higher value represents better reliability. In the five-factor solution, the first factor contained three items and explained 10.25% of the total variance with a internal reliability coefficient ordinal  $\alpha = .94$ . All the items in this factor stemmed from the individual's preference to use blue light filters in different light environments. The second factor contained six items and explained 9.93% of the total variance with a internal reliability coefficient ordinal  $\alpha = .76$ . Items under this factor commonly investigated an individual's hours spent 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 using phone and smart-watch in bed. The internal consistency reliability coefficient was, ordinal  $\alpha = .75$ . The fourth factor contained five items and explained 8.44% of the total variance with an internal consistency coefficient, ordinal  $\alpha = .72$ . These five items investigated the behaviors related to individual's light exposure before bedtime. Lastly,

the fifth factor contained six items and explained 6.14% of the total variance. This factor captured individual's morning and daytime light exposure related behavior. The internal consistency reliability was, ordinal  $\alpha = .62$ . 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 fifth factor, we retain all the five-factors 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. Figure 5 depicts the data distribution and endorsement pattern for the included items in our LEBA tool for both the EFA and CFA sample.

### Confirmatory Factor Analysis

We conducted categorical confirmatory factor analysis with robust weighted least square (WLSMV) estimator since our response data was of ordinary nature (Desjardins & Bulut, 2018). Several indices are suggested to measure model fit which 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  $\chi^2$  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 assesses 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  $\chi^2$  test statistics (A non-significant test statistics is required to reflect model

fit) (ii) CFI and TLI (CFI/TLI close to .95 or above/ranging between 90-95 and above) (iii) RMSEA (close to .06 or below), (iv) SRMR (close to .08 or below) to estimate the model fit. Table 5 summarizes the fit indices of our fitted model. Our fitted model failed to attain an absolute fit estimated by the  $\chi^2$  test. However, the  $\chi^2$  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. Yu, 2002). 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 equity constrain on item pairs 32-33 [I dim my mobile phone screen within 1 hour before attempting to fall asleep.;I dim my computer screen within 1 hour before attempting to fall asleep.] and 16-17 [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.]. Items pair 32-33 stemmed from the preference of dimming electric device's brightness before bed time and items pair 16 and 19 stemmed from the preference of using blue filtering or colored glasses during the daytime. Nevertheless, SRMR value was higher than the guideline (SRMR = .12). Further by allowing one pair of items (30-41) [I look at my smartwatch within 1 hour before attempting to fall asleep.;I look at my smartwatch when I wake up at night.] 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). Internal consistency ordinal  $\alpha$  for the five factors of LEBA were .96, .83, .70, .69, .52 respectively. We also estimated the internal consistency reliability of the total scale using McDonald's  $\omega_t$  coefficient which is a better reliability estimate for multidimensional constructs (Dunn, Baguley, & Brunsden, 2014; Sijtsma, 2009). McDonald's  $\omega_t$  coefficient for the total scale was .73. Figure 6 depicts the obtained CFA structure.

## Measurement Invariance

Measurement invariance (MI) evaluates whether a construct has the psychometric equivalence and same meaning across groups or measurement occasions (Kline, 2015; Putnick & Bornstein, 2016). We used structural equation modeling framework to assess the measurement invariance of our developed tool across two groups: native English speakers (n= 129) and non-native English speakers (n = 133). For a detailed description these two groups please see Sup. Table ???. Our measurement invariance testing involved successively comparing the nested models: configural, metric, scalar, and residual invariance models with each others (Widaman & Reise, 1997). Among these nested models configural model is the first and least restrictive model. The configural model assumes that the number of factors and item number under each factor will be equal across two groups. The metric invariance model assumes configural invariance of the fitted model and requires the factor-loadings of the items across the two groups to be equal. Having the factor-loadings equal across groups indicates each item contributes to the measured construct equivalently. Scalar invariance assumes the metric invariance of the fitted model demands the item intercepts to be equivalent across groups. This equity of item intercepts indicates the equivalence of response scale across the groups, i.e., persons with the same level of the underlying construct will score the same across the groups. The residual invariance model assumes metric invariance for the fitted model and adds the assumption of equality in error variances and covariances across the groups. This model is the highest level of MI and assures the equivalence of precision of items across the groups in measuring the underlying constructs. The invariance model fit of our tool was assessed using the fit indices including  $\chi^2$  test, CFI and TLI (close to .95 or above), RMSEA (close to .06 or below) (Hu & Bentler, 1999). We excluded SRMR from our consideration as it does not behave optimally for categorical variables (C. Yu, 2002). Table 6 summarized the fit indices. The comparison among different measurement invariance models was made using the  $\chi^2$  difference test ( $\Delta\chi^2$ ) to



assess whether our obtained latent structure of “LEBA” attained the highest level of the MI. A non-significant  $\Delta\chi^2$  test between two MI models fit indicates model fit does not significantly decrease for the superior model (Dimitrov, 2010) thus allowing the superior level of invariance model to be accepted. We started our analysis by comparing the model fit of the least restrictive model: configural model to metric MI model and continued successive comparisons. Table 6 indicates that our fitted model had acceptable fit indices for all of the fitted MI models. The model fit did not significantly decrease across the nested models up to the scalar MI model. The chi-square value difference between the scalar and residual model is zero, indicating model fit remained the same for both: scalar and residual MI model, indicating the acceptability of the residual MI model.

### Semantic Analysis

To find out if our developed tool (23 items) is overlapping with existing instruments, we subjected the items of LEBA to the “Semantic Scale Network”(SSN) analysis (Rosenbusch et al., 2020). The SSN detects semantically related scales and provides cosine similarity index ranging between -.66 to 1 (Rosenbusch et al., 2020). Pair of scales with a cosine similarity index value of 1 indicates they are perfectly semantically similar scales indicating redundancy. LEBA appeared most strongly related to scales about sleep: “Sleep Disturbance Scale For Children” (Bruni et al., 1996) and “WHO-Composite International Diagnostic Interview (CIDI): Insomnia”(WHO, 1990). The cosine similarities lie between .47 to .51. Flesch-Kincaid Grade Level (Flesch, 1948) analysis on the the 23 items of our scale indicated required educational grade level was 3.33 and with a age above 8.33.

## Developing Short form of LEBA

We sought the item response theory (IRT) to develop the short form of LEBA. IRT the conventional classical test theory-based analysis by gathering information on item quality by indices like item difficulty, item discrimination, and item information (Baker, 2017). Item is judged based on item information in relation to participants' latent trait level ( $\theta$ ). We fitted each factor of LEBA with the graded response model (Samejima, Liden, & Hambleton, 1997) to the combined EFA and CFA sample ( $n = 690$ ). Item discrimination indicates the pattern of variation in the categorical responses with the changes in latent trait level ( $\theta$ ), 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  $> 1.70$ . Table 7 summarizes the IRT parameters of our tool. Item discrimination parameters of our tool fell in very high (10 items), high (4 items), moderate (4 items), and low ( 5 items) categories indicating a good range of discrimination along the latent trait level ( $\theta$ ). Examination of the item information curve (Sup.fig A3-A6) indicated 5 items (1, 25, 38, 30, & 41) had relatively flat information curves ( $I(\theta) < .20$ ) thus discarded creating a short form of LEBA with 5 factors and 18 items.

Test information curve (TIC) (Figure 7) indicate the amount of information an the full-scale carry along the latent trait continuum. As we treated each factor of short-LEBA as an unidimensional construct we obtain 5 TICs (Figure 7). These information curves indicated except the first and fifth factors, the other three factor's TICs are roughly centered on the center of the trait continuum ( $\theta$ ). The first and fifth factor had a peak to the right side of the center of latent trait. Thus we conferred the LEBA tool estimated the

light exposure related behavior with precision near the center of trait continuum for 2nd, 3rd and 4th factors and near the right side of the center of trait continuum for 1st and 5th factors (Baker, 2017).

Table 8 summarizes the item fit indexes of the items. All the items fitted well to the respective models as assessed by RMSEA value obtained from Signed- $\chi^2$  index implementation. All of the items had RMSEA value  $\leq .06$  indicating adequate fit. Sup.Figure A7 depicts the person fit of our fitted models. Person fit indicates the validity and meaningfulness of the fitted model at the participants latent trait level (Desjardins & Bulut, 2018). We estimated the person fit statistics using standardized fit index Zh statistics (Dragow, Levine, & Williams, 1985).  $Z_h < -2$  should be considered as a misfit. Fig indicates that  $Z_h$  is larger than -2 for most participants, suggesting a good fit of the selected IRT models.

## Discussion

We developed a self-reported tool to capture different light exposure related behavior and evaluated its psychometric properties using classical test theory and item response theory based analysis.

48 items were generated by an expert panel and administered to a large sample ( $n = 428$ ) to explore the latent structure. Exploratory factor analysis revealed a five factor solution with 25 items. ("Wearing blue light filters," "Spending time outdoors," "Using phone and smart-watch in bed," "Using light before bedtime," and "Using light in the morning and during daytime"). The internal consistency reliability coefficient ordinal alpha ranged between .62-.94. As all the retained factors were meaningful and contributed essentially towards our aim we retained all five factors.

A CFA on a separate sample ( $n = 262$ ) gave a five-factor solution ( $CFI = .97$ ;  $TLI = .96$ );  $RMSEA = .05$  [.04-.06, 90% CI] and  $SRMR = .09$ ) after discarding two items. The

internal consistency McDonald's  $\omega_t$  of the five factors were satisfactory (.96, .83, .70, .69, .52) Internal consistency reliability of the total scale (23 items) was also satisfactory, McDonald's  $\omega_t = .73$ . In the same sample, our measurement invariance analysis revealed that the latent structure attained the residual measurement invariance across subgroups: male and female (CFI: .98, TLI: .98, SRMR: .98).

The "Semantic Scale Network"(SSN) analysis (Rosenbusch et al., 2020) on the retained 23 items showed "LEBA" was related to "Sleep Disturbance Scale For Children" (SDSC) (Bruni et al., 1996) and "WHO-Composite International Diagnostic Interview (CIDI): Insomnia"(WHO, 1990). Upon inspecting the item contents we found items under "Using phone and smart-watch in bed" and "Using light before bedtime" have semantic overlap with the items of SDSC and CIDI. Items in those two scales were looking into behaviors related to sleep. As such the similarity index obtained is expected. Flesch-Kincaid Grade Level (Flesch, 1948) analysis on the the 23 items of our scale indicated required educational grade level was 3.33 and with a age above 8.33.

Lastly, we developed a short-LEBA (n=23) using IRT analysis. We fitted a graded response model model to the combined EFA and CFA sample (n =690). We discarded 5 items with relatively flat item information curve [ $I(\theta) < .20$ ]. IRT analysis indicated short form of 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. Test information curve also indicated a good coverage of underlying trait continuum with precision.

## Conclusion

"The Light exposure behavior assessment"(LEBA) gave a five solution with 25 items in an exploratory factor analysis. A confirmatory factor analysis with this 25-item

478 scale again offered a five-factor solution, but this time two more item was discarded. The  
479 23-item “LEBA” was found reliable and valid. A short-form of LEBA was developed using  
480 IRT analysis. IRT analysis gave a 18-item scale with a good range of coverage across  
481 the underlying trait continuum. All-in-all, we can recommend both forms to be used to  
482 capture individual’s light exposure related behavior

## References

- Aust, F., & Barth, M. (2020). *papaja: Prepare reproducible APA journal articles with R Markdown*. Retrieved from <https://github.com/crsh/papaja>
- Bajaj, A., Rosner, B., Lockley, S. W., & Schernhammer, E. S. (2011). Validation of a light questionnaire with real-life photopic illuminance measurements: The harvard light exposure assessment questionnaire. *Cancer Epidemiology and Prevention Biomarkers*, 20(7), 1341–1349.
- 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>

- Bossini, L., Valdagno, M., Padula, L., De Capua, A., Pacchierotti, C., & Castrogiovanni, P. (2006). Sensibilità alla luce e psicopatologia: Validazione del questionario per la valutazione della fotosensibilità (QVF). *Med Psicosomatica*, 51, 167–176.
- Brown, T. A. (2015). *Confirmatory factor analysis for applied research* (2nd ed.). New York, NY, US: The Guilford Press.
- Bruni, O., Ottaviano, S., Guidetti, V., Romoli, M., Innocenzi, M., Cortesi, F., & Giannotti, F. (1996). The sleep disturbance scale for children (SDSC) construction and validation of an instrument to evaluate sleep disturbances in childhood and adolescence. *Journal of Sleep Research*, 5(4), 251–261.
- 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>
- Buysse, Daniel J., Reynolds III, C. F., Monk, T. H., Berman, S. R., & Kupfer, D. J. (1989). The pittsburgh sleep quality index: A new instrument for psychiatric practice and research. *Psychiatry Research*, 28(2), 193–213.
- Buysse, Daniel 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.

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

538 Chang, W., Cheng, J., Allaire, J., Sievert, C., Schloerke, B., Xie, Y., ... Borges, B.

539 (2021). *Shiny: Web application framework for r*. Retrieved from

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

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

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

543 Conigrave, J. (2020). *Corx: Create and format correlation matrices*. Retrieved

544 from <https://CRAN.R-project.org/package=corx>

545 Dahl, D. B., Scott, D., Roosen, C., Magnusson, A., & Swinton, J. (2019). *Xtable:*

546 *Export tables to LaTeX or HTML*. Retrieved from

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

548 Desjardins, C., & Bulut, O. (2018). *Handbook of Educational Measurement and*

549 *Psychometrics Using R*. <https://doi.org/10.1201/b20498>

550 Dianat, I., Sedghi, A., Bagherzade, J., Jafarabadi, M. A., & Stedmon, A. W. (2013).

551 Objective and subjective assessments of lighting in a hospital setting:

552 Implications for health, safety and performance. *Ergonomics*, 56(10),

553 1535–1545.

554 Dimitrov, D. M. (2010). Testing for factorial invariance in the context of construct

555 validation. *Measurement and Evaluation in Counseling and Development*,

556 43(2), 121–149.

557 Dinno, A. (2018). *Paran: Horn's test of principal components/factors*. Retrieved

558 from <https://CRAN.R-project.org/package=paran>

559 Drasgow, F., Levine, M. V., & Williams, E. A. (1985). Appropriateness

560 measurement with polychotomous item response models and standardized

561 indices. *British Journal of Mathematical and Statistical Psychology*, 38(1),

562 67–86.

563 Dunn, T. J., Baguley, T., & Brunsden, V. (2014). From alpha to omega: A practical



564 solution to the pervasive problem of internal consistency estimation. *British*  
565 *Journal of Psychology*, 105(3), 399–412.

566 Eklund, N., & Boyce, P. (1996). The development of a reliable, valid, and simple  
567 office lighting survey. *Journal of the Illuminating Engineering Society*, 25(2),  
568 25–40.

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

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

575 Flesch, R. (1948). A new readability yardstick. *Journal of Applied Psychology*,  
576 32(3), 221.

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

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

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

586 Fox, J., Weisberg, S., & Price, B. (2020). *carData: Companion to applied*  
587 *regression data sets*. Retrieved from  
588 <https://CRAN.R-project.org/package=carData>

589 Gadermann, A. M., Guhn, M., & Zumbo, B. D. (2012). Estimating ordinal reliability  
590 for likert-type and ordinal item response data: A conceptual, empirical, and

- practical guide. *Practical Assessment, Research, and Evaluation*, 17(1), 3.
- Graham, J. M. (2006). Congeneric and (essentially) tau-equivalent estimates of score reliability: What they are and how to use them. *Educational and Psychological Measurement*, 66(6), 930–944.
- Harb, F., Hidalgo, M. P., & Martau, B. (2015). Lack of exposure to natural light in the workspace is associated with physiological, sleep and depressive symptoms. *Chronobiology International*, 32(3), 368–375.  
<https://doi.org/10.3109/07420528.2014.982757>
- Harrell Jr, F. E., Charles Dupont, with contributions from, & others., many. (2021). *Hmisc: Harrell miscellaneous*. Retrieved from  
<https://CRAN.R-project.org/package=Hmisc>
- Harris, P. A., Taylor, R., Minor, B. L., Elliott, V., Fernandez, M., O’Neal, L., ... others. (2019). The REDCap consortium: Building an international community of software platform partners. *Journal of Biomedical Informatics*, 95, 103208.
- Harris, P. A., Taylor, R., Thielke, R., Payne, J., Gonzalez, N., & Conde, J. G. (2009). Research electronic data capture (REDCap)—a metadata-driven methodology and workflow process for providing translational research informatics support. *Journal of Biomedical Informatics*, 42(2), 377–381.
- Henry, L., & Wickham, H. (2020). *Purrr: Functional programming tools*. Retrieved from <https://CRAN.R-project.org/package=purrr>
- Horn, J. L. (1965). A rationale and test for the number of factors in factor analysis. *Psychometrika*, 30(2), 179–185. <https://doi.org/10.1007/BF02289447>
- Horne, J. A., & Östberg, O. (1976). A self-assessment questionnaire to determine morningness-eveningness in human circadian rhythms. *International Journal of Chronobiology*.
- Hu, L., & Bentler, P. M. (1999). Cutoff criteria for fit indexes in covariance structure analysis: Conventional criteria versus new alternatives. *Structural Equation*

618 *Modeling: A Multidisciplinary Journal*, 6(1), 1–55.

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

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

622 Iannone, R. (2016). *DiagrammeRsvg: Export DiagrammeR graphviz graphs as*  
623 *SVG*. Retrieved from <https://CRAN.R-project.org/package=DiagrammeRsvg>

624 Iannone, R. (2021). *DiagrammeR: Graph/network visualization*. Retrieved from  
625 <https://github.com/rich-iannone/DiagrammeR>

626 Irribarra, D. T., & Freund, R. (2014). *Wright map: IRT item-person map with*  
627 *ConQuest integration*. Retrieved from <http://github.com/david-ti/wrightmap>

628 Jackson, D. L. (2003). Revisiting Sample Size and Number of Parameter  
629 Estimates: Some Support for the N:q Hypothesis. *Structural Equation*  
630 *Modeling*, 10(1), 128–141. [https://doi.org/10.1207/S15328007SEM1001\\_6](https://doi.org/10.1207/S15328007SEM1001_6)

631 Johnson, P., & Kite, B. (2020). *semTable: Structural equation modeling tables*.  
632 Retrieved from <https://CRAN.R-project.org/package=semTable>

633 Johnson, P., Kite, B., & Redmon, C. (2020). *Kutils: Project management tools*.  
634 Retrieved from <https://CRAN.R-project.org/package=kutils>

635 Jorgensen, T. D., Pornprasertmanit, S., Schoemann, A. M., & Rosseel, Y. (2021).  
636 *semTools: Useful tools for structural equation modeling*. Retrieved from  
637 <https://CRAN.R-project.org/package=semTools>

638 Kaiser, H. F. (1974). An index of factorial simplicity. *Psychometrika*, 39(1), 31–36.  
639 <https://doi.org/10.1007/bf02291575>

640 Kassambara, A. (2019). *Ggcorrplot: Visualization of a correlation matrix using*  
641 *'ggplot2'*. Retrieved from <https://CRAN.R-project.org/package=ggcorrplot>

642 Kline, R. B. (2015). *Principles and practice of structural equation modeling*. The  
643 Guilford Press.

644 Kowarik, A., & Templ, M. (2016). Imputation with the R package VIM. *Journal of*

- 645           *Statistical Software*, 74(7), 1–16. <https://doi.org/10.18637/jss.v074.i07>
- 646 Lishinski, A. (2021). *lavaanPlot: Path diagrams for 'lavaan' models via*
- 647           *'DiagrammeR'*. Retrieved from
- 648           <https://CRAN.R-project.org/package=lavaanPlot>
- 649 Lorenzo-Seva, U., Timmerman, M., & Kiers, H. (2011). The Hull Method for
- 650           Selecting the Number of Common Factors. *Multivariate Behavioral Research*,
- 651           46, 340–364. <https://doi.org/10.1080/00273171.2011.564527>
- 652 Makowski, D., Ben-Shachar, M. S., Patil, I., & Lüdtke, D. (2020). Methods and
- 653           algorithms for correlation analysis in R. *Journal of Open Source Software*,
- 654           5(51), 2306. <https://doi.org/10.21105/joss.02306>
- 655 Mardia, K. V. (1970). Measures of multivariate skewness and kurtosis with
- 656           applications. *Biometrika*, 57(3), 519–530.
- 657           <https://doi.org/10.1093/biomet/57.3.519>
- 658 Michalke, M. (2020a). *koRpus.lang.en: Language support for 'koRpus' package:*
- 659           *english*. Retrieved from <https://reaktanz.de/?c=hacking&s=koRpus>
- 660 Michalke, M. (2020b). *Sylly: Hyphenation and syllable counting for text analysis*.
- 661           Retrieved from <https://reaktanz.de/?c=hacking&s=sylly>
- 662 Michalke, M. (2021). *koRpus: Text analysis with emphasis on POS tagging,*
- 663           *readability, and lexical diversity*. Retrieved from
- 664           <https://reaktanz.de/?c=hacking&s=koRpus>
- 665 Mock, T. (2021). *gtExtras: A collection of helper functions for the gt package*.
- 666           Retrieved from <https://github.com/jthomasmock/gtExtras>
- 667 Müller, K., & Wickham, H. (2021). *Tibble: Simple data frames*. Retrieved from
- 668           <https://CRAN.R-project.org/package=tibble>
- 669 Navarro-Gonzalez, D., & Lorenzo-Seva, U. (2021). *EFA.MRFA: Dimensionality*
- 670           *assessment using minimum rank factor analysis*. Retrieved from
- 671           <https://CRAN.R-project.org/package=EFA.MRFA>

- Neuwirth, E. (2014). *RColorBrewer: ColorBrewer palettes*. Retrieved from <https://CRAN.R-project.org/package=RColorBrewer>
- Novick, M. R., & Lewis, C. (1967). Coefficient alpha and the reliability of composite measurements. *Psychometrika*, 32(1), 1–13.
- 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>
- Peters, G.-J. (2021). *Ufs: Quantitative analysis made accessible*. Retrieved from <https://CRAN.R-project.org/package=ufs>
- 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>
- Putnick, D. L., & Bornstein, M. H. (2016). Measurement invariance conventions and reporting: The state of the art and future directions for psychological research. *Developmental Review*, 41, 71–90.
- R Core Team. (2021). *R: A language and environment for statistical computing*. Vienna, Austria: R Foundation for Statistical Computing. Retrieved from <https://www.R-project.org/>

- 699 Revelle, W. (2021). *Psych: Procedures for psychological, psychometric, and*  
700 *personality research*. Evanston, Illinois: Northwestern University. Retrieved  
701 from <https://CRAN.R-project.org/package=psych>
- 702 Roenneberg, T., Wirz-Justice, A., & Mellow, M. (2003). Life between clocks: Daily  
703 temporal patterns of human chronotypes. *Journal of Biological Rhythms*,  
704 18(1), 80–90.
- 705 Rosenbusch, H., Wanders, F., & Pit, I. L. (2020). The semantic scale network: An  
706 online tool to detect semantic overlap of psychological scales and prevent  
707 scale redundancies. *Psychological Methods*, 25(3), 380.
- 708 Rosseel, Y. (2012). lavaan: An R package for structural equation modeling.  
709 *Journal of Statistical Software*, 48(2), 1–36. Retrieved from  
710 <https://www.jstatsoft.org/v48/i02/>
- 711 Ryu, C. (2021). *Dlookr: Tools for data diagnosis, exploration, transformation*.  
712 Retrieved from <https://CRAN.R-project.org/package=dlookr>
- 713 Samejima, F., Liden, W. van der, & Hambleton, R. (1997). Handbook of modern  
714 item response theory. New York, NY: Springer.
- 715 Sarkar, D. (2008). *Lattice: Multivariate data visualization with r*. New York:  
716 Springer. Retrieved from <http://lmdvr.r-forge.r-project.org>
- 717 Schönbrodt, F. D., & Perugini, M. (2013). At what sample size do correlations  
718 stabilize? *Journal of Research in Personality*, 47(5), 609–612.  
719 <https://doi.org/10.1016/j.jrp.2013.05.009>
- 720 Schumacker, R. E., & Lomax, R. G. (2004). *A beginner's guide to structural*  
721 *equation modeling*. psychology press.
- 722 Shapiro, S. S., & Wilk, M. B. (1965). An analysis of variance test for normality  
723 (complete samples). *Biometrika*, 52(3-4), 591–611.  
724 <https://doi.org/10.1093/biomet/52.3-4.591>
- 725 Sijtsma, K. (2009). On the use, the misuse, and the very limited usefulness of

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

Siraji, M. A. (2021). *Tabledown: A companion pack for the book "basic & advanced psychometrics in r"*. Retrieved from

<https://github.com/masiraji/taledown>

Sjoberg, 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/>

Verriotto, J. D., Gonzalez, A., Aguilar, M. C., Parel, J.-M. A., Feuer, W. J., Smith,

A. R., & Lam, B. L. (2017). New methods for quantification of visual  
photosensitivity threshold and symptoms. *Translational Vision Science &  
Technology*, 6(4), 18–18.

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>

WHO. (1990). Composite international diagnostic interview.

Wickham, H. (2011). The split-apply-combine strategy for data analysis. *Journal  
of Statistical Software*, 40(1), 1–29. Retrieved from  
<http://www.jstatsoft.org/v40/i01/>

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

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

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

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

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

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

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

Wickham, H., & Hester, J. (2021). *Readr: Read rectangular text data*. Retrieved



780 from <https://CRAN.R-project.org/package=readr>

781 Widaman, K. F., & Reise, S. P. (1997). Exploring the measurement invariance of  
782 psychological instruments: Applications in the substance use domain.

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

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

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

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

792 Xie, Yihui. (2015). *Dynamic documents with R and knitr* (2nd ed.). Boca Raton,  
793 Florida: Chapman; Hall/CRC. Retrieved from <https://yihui.org/knitr/>

794 Xie, Yang, Wu, X., Tao, S., Wan, Y., & Tao, F. (2021). Development and validation  
795 of the self-rating of biological rhythm disorder for chinese adolescents.

796 *Chronobiology International*, 1–7.

797 <https://doi.org/10.1080/07420528.2021.1989450>

798 Yu, C. (2002). *Evaluating cutoff criteria of model fit indices for latent variable*  
799 *models with binary and continuous outcomes* (Thesis). ProQuest  
800 Dissertations Publishing.

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

802 Pilkonis, P. A. (2011). Development of short forms from the PROMIS™ sleep  
803 disturbance and sleep-related impairment item banks. *Behavioral Sleep*  
804 *Medicine*, 10(1), 6–24. <https://doi.org/10.1080/15402002.2012.636266>

805 Yuan, K.-H., & Zhang, Z. (2020). *Rsem: Robust structural equation modeling with*  
806 *missing data and auxiliary variables*. Retrieved from

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

808 Zeileis, A., & Croissant, Y. (2010). Extended model formulas in R: Multiple parts  
809 and multiple responses. *Journal of Statistical Software*, 34(1), 1–13.

810 <https://doi.org/10.18637/jss.v034.i01>

811 Zeileis, A., Fisher, J. C., Hornik, K., Ihaka, R., McWhite, C. D., Murrell, P., ...

812 Wilke, C. O. (2020). colorspace: A toolbox for manipulating and assessing  
813 colors and palettes. *Journal of Statistical Software*, 96(1), 1–49.

814 <https://doi.org/10.18637/jss.v096.i01>

815 Zeileis, A., Hornik, K., & Murrell, P. (2009). Escaping RGBland: Selecting colors  
816 for statistical graphics. *Computational Statistics & Data Analysis*, 53(9),

817 3259–3270. <https://doi.org/10.1016/j.csda.2008.11.033>

818 Zhang, Z., & Yuan, K.-H. (2020). *Coefficientalpha: Robust coefficient alpha and*  
819 *omega with missing and non-normal data*. Retrieved from

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

821 Zhu, H. (2021). *kableExtra: Construct complex table with 'kable' and pipe syntax*.

822 Retrieved from <https://CRAN.R-project.org/package=kableExtra>

823 Zumbo, B. D., Gadermann, A. M., & Zeisser, C. (2007). Ordinal versions of

824 coefficients alpha and theta for likert rating scales. *Journal of Modern Applied*  
825 *Statistical Methods*, 6(1), 4.

Table 1

*Releated Scales*

Name	Author	Description	Relevant Items	Scale type	Validity evidences
Visual Light Sensitivity Questionnaire- 8	Verriotto et al., 2017	Eight- question survey to assess the presence and severity of photosen- sitivity symptoms	None	5-point Likert scale	Not available
Office Light Survey	Eklundet al., 1996	30 items survey to assess electrical lighting en- vironment in office	Item 29	Mixed response format	Not available

Table 1

*Releated Scales (continued)*

Name	Author	Description	Relevant Items	Scale type	Validity evidences
Harvard Light Exposure Assess- ment Question- naire	Bajaj et al., 2011	1 item semi- quantitative light ques- tionnaire	None	Semi- quantitative	Correlation with physical measure- mernt
Hospital Lighting Survey	Dianat et el., 2013	23 items question- naire to assess light envi- ronment in a hospital	Item 16,17	5-point Likert scale	Face and Content validity
Morningness- Eveningness Question- naire	Horne et al., 1976	19 items question- naire to under- stand your body clock	item 1,2,8,13,14	Mixed response format	Correlation the oral tempera- ture

Table 1

*Related Scales (continued)*

Name	Author	Description	Relevant Items	Scale type	Validity evidences
Munich Chrono- type Question- naire (MCTQ)	Roenneberg et al., 2003	17 items question- naire to under- stand individuals phase of entrain- ment	Time spect outdoors	Mixed response format	Correlation with sleep-logs, actimetry, and physi- ological parame- ters
Sleep Practices and Attitudes Question- naire (SPAQ)	Olivier et.al., 2016	16 Factor question- naire measuring practice, behavior and attitude related sleep	Subscale 8&9	5-point Likert scale	Face and Construct validity

Table 1

*Releated Scales (continued)*

Name	Author	Description	Relevant Items	Scale type	Validity evidences
The Pittsburgh Sleep Quality Index (PSQI)	Buyse et al., 1989	9 items inventory to measure sleep quality and sleeping pattern	item 1-4	Mixed response format	Correlation with clinical measure- ments
Self- Rating of Biological Rhythm Disorder for Disorder for Adoles- cents (SBRDA)	Xie et al., 2021	29 Items question- naire assessing four di- mensions of biological rhythm disorder in adoles- cents	Item 3,6,22-25 and 29	5-point Likert scale	Construct validity

Table 1

*Releated Scales (continued)*

Name	Author	Description	Relevant Items	Scale type	Validity evidences
Photosensitive Assess- ment Question- naire (PAQ)	Rossini et al.,2006	16 dichoto- mous (yes/no) items question- naire to assess "photopho- bia" and "pho- tophilia"	All items	Binary response option	Not available

Table 2

*Demographic Characteristics*

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> False discovery rate correction for multiple testing<sup>3</sup> Wilcoxon rank sum test<sup>4</sup> Pearson's Chi-squared test



Table 3

*Descriptive Statistics*

	Mean	SD	Skew	Kurtosis	Shapiro-Wilk Statistics	Item-Total Correlation
Item1	2.27	1.39	0.74	-0.81	0.81*	0.19
Item2	2.87	1.59	0.08	-1.60	0.83*	0.28
Item3	3.36	1.38	-0.48	-1.03	0.87*	0.23
Item4	1.47	1.18	2.38	4.00	0.43*	0.24
Item5	4.01	1.40	-1.22	0.07	0.70*	0.17
Item6	2.79	1.55	0.19	-1.48	0.85*	0.13
Item7	2.26	1.25	0.70	-0.60	0.85*	0.32
Item8	2.97	1.20	-0.06	-0.94	0.91*	0.25
Item9	2.94	1.03	-0.12	-0.40	0.91*	0.08
Item10	2.74	1.04	0.09	-0.74	0.91*	0.42
Item11	2.18	0.90	0.60	0.12	0.86*	0.41
Item12	2.36	1.22	0.59	-0.62	0.87*	0.48
Item13	2.73	1.46	0.20	-1.36	0.87*	0.25
Item14	2.14	1.31	0.77	-0.78	0.80*	0.28
Item15	3.26	1.09	-0.26	-0.45	0.91*	0.03
Item16	1.56	1.23	2.00	2.45	0.50*	0.28
Item17	1.54	1.21	2.07	2.75	0.49*	0.21
Item18	1.12	0.49	5.02	27.80	0.25*	0.18
Item19	1.05	0.36	7.23	52.98	0.13*	0.17
Item20	1.04	0.33	8.99	85.28	0.10*	0.16
Item21	1.14	0.59	4.79	24.05	0.25*	0.21
Item22	3.57	1.07	-0.65	-0.17	0.88*	0.20
Item23	2.56	1.27	0.33	-1.00	0.89*	0.08

Table 3 continued

	Mean	SD	Skew	Kurtosis	Shapiro-Wilk Statistics	Item-Total Correlation
Item24	4.14	0.99	-1.23	1.14	0.79*	0.22
Item25	2.59	1.41	0.27	-1.27	0.86*	0.15
Item26	2.25	1.27	0.69	-0.64	0.84*	0.08
Item27	3.80	1.29	-0.87	-0.42	0.82*	0.17
Item28	3.76	1.14	-0.68	-0.45	0.86*	0.18
Item29	2.44	1.31	0.38	-1.14	0.86*	0.13
Item30	1.48	1.11	2.18	3.35	0.48*	0.13
Item31	3.00	1.62	-0.08	-1.61	0.83*	0.39
Item32	3.55	1.65	-0.60	-1.34	0.76*	0.33
Item33	3.62	1.64	-0.68	-1.25	0.74*	0.37
Item34	3.42	1.83	-0.45	-1.69	0.69*	0.20
Item35	3.86	1.67	-0.99	-0.85	0.65*	0.20
Item36	1.54	1.25	2.13	2.86	0.46*	0.35
Item37	1.33	0.91	3.03	8.43	0.41*	0.09
Item38	4.30	1.08	-1.79	2.53	0.67*	0.32
Item39	1.96	0.98	1.02	0.69	0.82*	0.07
Item40	2.16	1.19	0.71	-0.54	0.84*	0.25
Item41	1.31	0.81	2.75	6.92	0.43*	0.14
Item42	3.93	1.48	-1.06	-0.44	0.71*	0.15
Item43	1.64	1.18	1.79	2.02	0.60*	0.22
Item44	3.51	1.30	-0.70	-0.59	0.85*	0.40
Item45	2.22	1.48	0.71	-1.02	0.76*	0.29
Item46	1.76	1.23	1.35	0.44	0.66*	0.39
Item47	2.11	1.17	0.77	-0.39	0.83*	0.37

Table 3 continued

	Mean	SD	Skew	Kurtosis	Shapiro-Wilk Statistics	Item-Total Correlation
Item48	2.60	1.25	0.29	-0.86	0.89*	0.36

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

Table 4

*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 5

*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 6

*Invariance Analysis*

	Chi-Square	df	CFI	TLI	RMSEA	RMSEA 90% Lower CI	RMSEA 90% Upper	Chi-Square Difference	df difference*	p
Configural	632.20	442.00	0.95	0.94	0.06	0.05	0.07	-	-	-
Metric	644.58	458.00	0.95	0.95	0.06	0.05	0.07	18.019a	16	0.323
Scalar	714.19	522.00	0.95	0.95	0.05	0.04	0.06	67.961b	64	0.344
Residual	714.19	522.00	0.95	0.95	0.05	0.04	0.06	0c	0	NA

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

Table 7

*IRT Item parameters for the LEBA Scale*

	a	b1	b2	b3	b4
item16	28.13	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	1.37	-1.09	-0.98	-0.75	-0.40
item38	0.40	-7.48	-5.56	-4.23	-0.90
item33	12.31	-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

Table 8

*Item fit statistics for the fitted models*

Item	Signed Chi-square	df	RMSEA	p
item16	2.02	6.00	0.00	0.92
item36	39.07	13.00	0.05	0.00
item17	25.58	13.00	0.04	0.02
item11	55.03	27.00	0.04	0.00
item10	53.19	30.00	0.03	0.01
item12	34.39	42.00	0.00	0.79
item7	67.45	46.00	0.03	0.02
Ritem8	140.90	46.00	0.05	0.00
item9	131.19	45.00	0.05	0.00
item27	16.41	11.00	0.03	0.13
item3	15.09	11.00	0.02	0.18
item40	9.92	9.00	0.01	0.36
item32	41.33	15.00	0.05	0.00
item35	41.71	14.00	0.05	0.00
item33	46.89	14.00	0.06	0.00
item46	19.00	15.00	0.02	0.21
item45	15.05	15.00	0.00	0.45
item25	31.60	15.00	0.04	0.01



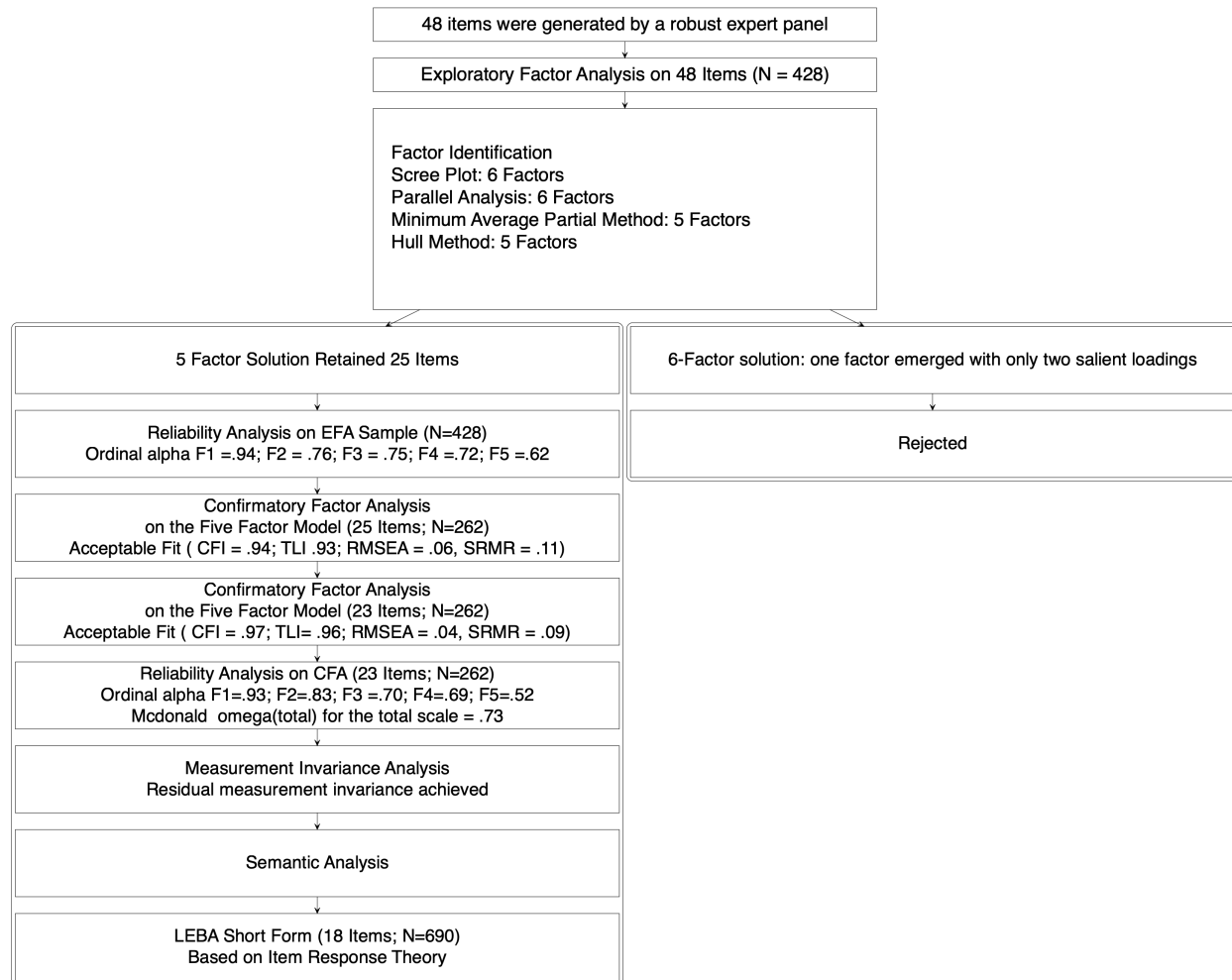


Figure 1. Development of long and short form of LEBA

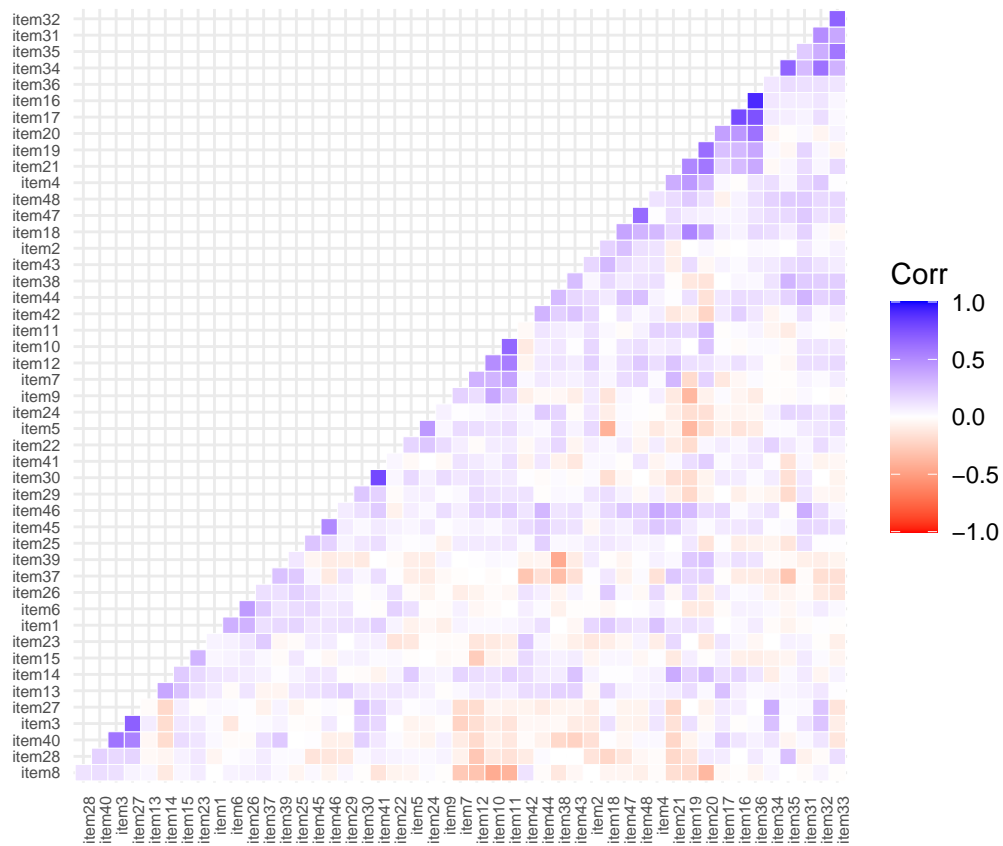
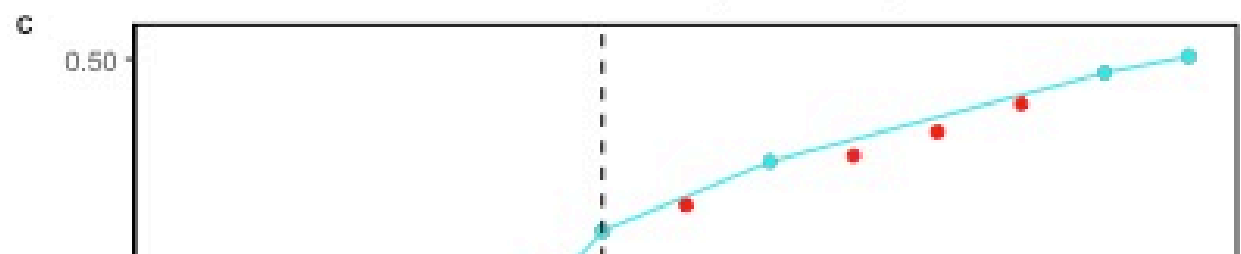
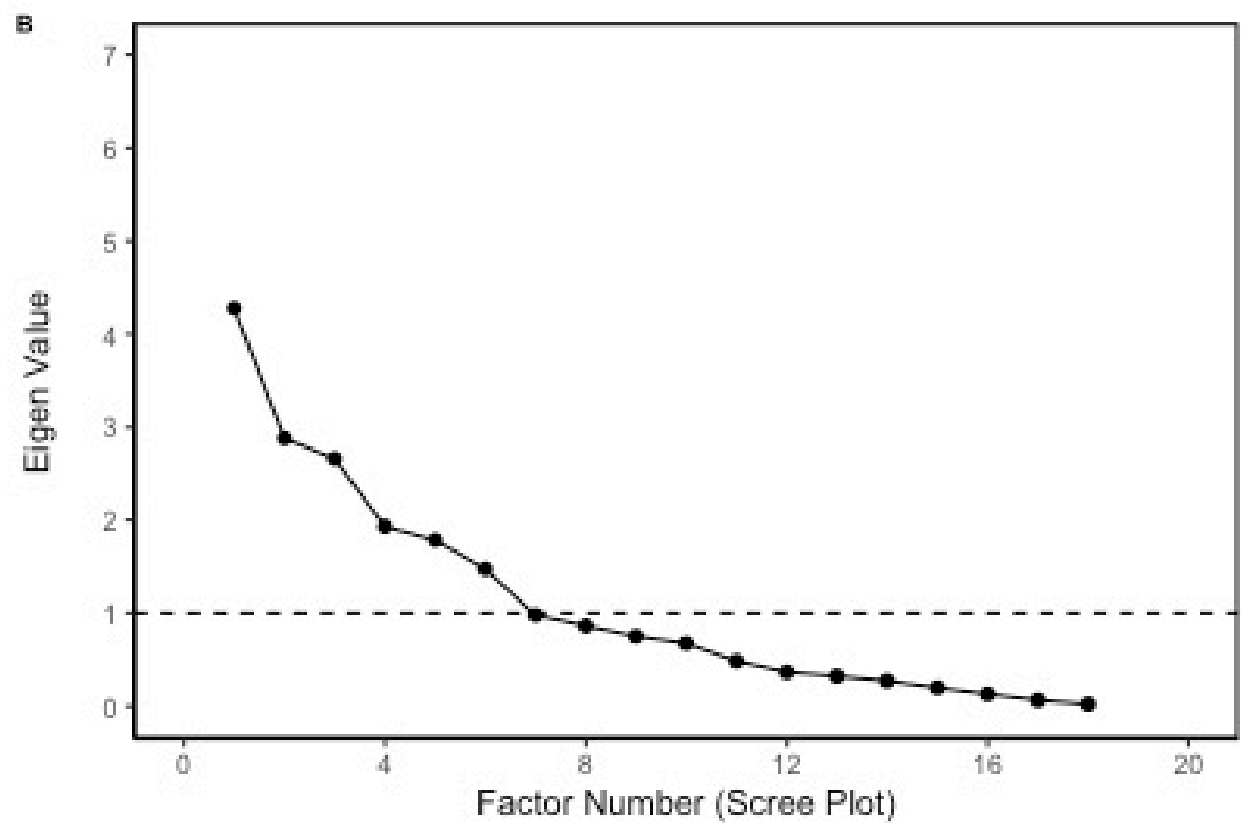
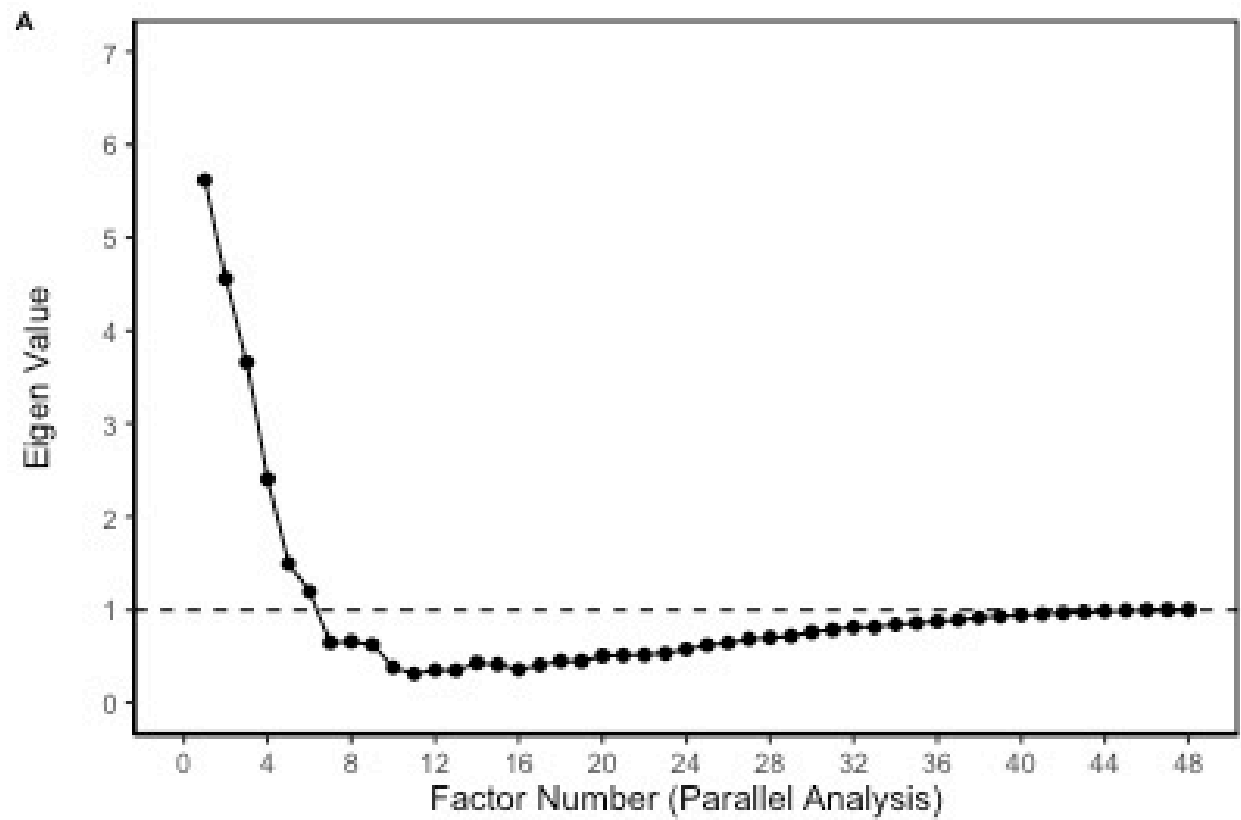


Figure 2. Correlation plot of the items



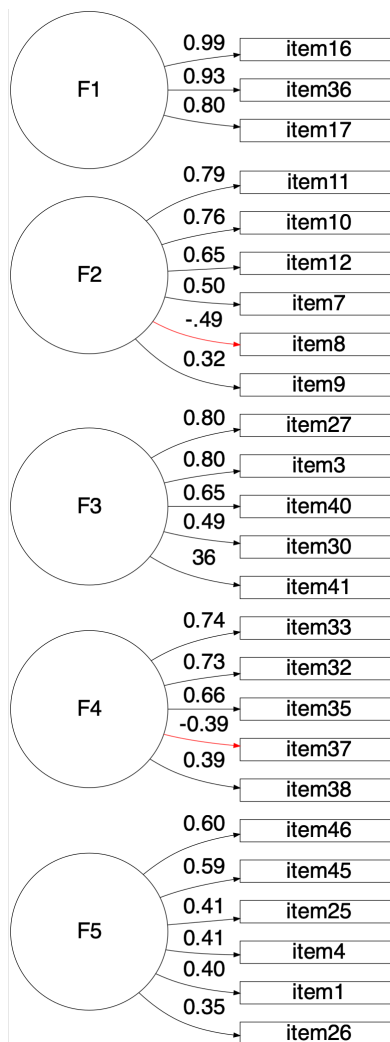


Figure 4. Five Factor Solution

LEBA											
Summary Descriptives of CFA and EFA Sample (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			42.29% (181)	22.20% (95)	12.62% (54)	12.38% (53)	10.51% (45)
● item03	428	3.4	4.0	1.4			15.89% (68)	11.45% (49)	17.29% (74)	31.07% (133)	24.30% (104)
● item04	428	1.5	1.0	1.2			84.11% (360)	3.50% (15)	2.10% (9)	2.10% (9)	8.18% (35)
● item07	428	2.3	2.0	1.2			35.98% (154)	27.80% (119)	17.29% (74)	12.38% (53)	6.54% (28)
● item08	428	3.0	3.0	1.2			13.79% (59)	22.20% (95)	27.80% (119)	25.93% (111)	10.28% (44)
● item09	428	2.9	3.0	1.0			10.28% (44)	19.63% (84)	41.82% (179)	22.43% (96)	5.84% (25)
● item10	428	2.7	3.0	1.0			11.92% (51)	31.31% (134)	31.31% (134)	21.96% (94)	3.50% (15)
● item11	428	2.2	2.0	0.9			22.43% (96)	46.26% (198)	23.13% (99)	7.01% (30)	1.17% (5)
● item12	428	2.4	2.0	1.2			29.91% (128)	29.67% (127)	21.50% (92)	12.15% (52)	6.78% (29)
● item16	428	1.6	1.0	1.2			79.67% (341)	4.21% (18)	3.97% (17)	4.67% (20)	7.48% (32)
● item17	428	1.5	1.0	1.2			80.61% (345)	3.27% (14)	5.14% (22)	3.27% (14)	7.71% (33)
● item25	428	2.6	3.0	1.4			34.35% (147)	13.79% (59)	22.20% (95)	17.99% (77)	11.68% (50)
● item26	428	3.7	4.0	1.3			38.32% (164)	23.36% (100)	20.09% (86)	10.98% (47)	7.24% (31)
● item27	428	3.8	4.0	1.3			8.41% (36)	11.21% (48)	11.21% (48)	30.37% (130)	38.79% (166)
● item30	428	1.5	1.0	1.1			81.78% (350)	3.27% (14)	4.91% (21)	5.37% (23)	4.67% (20)
● item32	428	3.6	4.0	1.6			23.13% (99)	7.01% (30)	8.18% (35)	14.95% (64)	46.73% (200)
● item33	428	3.6	4.0	1.6			21.96% (94)	7.01% (30)	7.24% (31)	14.49% (62)	49.30% (211)
● item35	428	3.9	5.0	1.7			22.90% (98)	1.87% (8)	3.74% (16)	9.35% (40)	62.15% (266)
● item36	428	1.5	1.0	1.3			82.24% (352)	3.04% (13)	3.04% (13)	2.34% (10)	9.35% (40)
● item37	428	2.3	2.0	1.3			38.32% (164)	23.36% (100)	20.09% (86)	10.98% (47)	7.24% (31)
● item38	428	4.3	5.0	1.1			5.37% (23)	3.50% (15)	5.37% (23)	27.57% (118)	58.18% (249)
● item40	428	2.2	2.0	1.2			39.49% (169)	25.00% (107)	19.63% (84)	11.45% (49)	4.44% (19)
● item41	428	1.3	1.0	0.8			85.05% (364)	4.67% (20)	6.07% (26)	3.04% (13)	1.17% (5)
● item45	428	2.2	1.0	1.5			53.04% (227)	7.01% (30)	16.36% (70)	11.92% (51)	11.68% (50)
● item46	428	1.8	1.0	1.2			67.06% (287)	7.71% (33)	11.68% (50)	8.88% (38)	4.67% (20)
CFA (n =262)											
● item01	262	2.3	2.0	1.4			40.46% (106)	22.52% (59)	14.50% (38)	10.69% (28)	11.83% (31)
● item03	262	3.7	4.0	1.3			11.83% (31)	7.25% (19)	17.56% (46)	28.24% (74)	35.11% (92)
● item04	262	1.3	1.0	0.8			89.31% (234)	2.29% (6)	3.44% (9)	3.05% (8)	1.91% (5)
● item07	262	2.1	2.0	1.2			43.13% (113)	23.66% (62)	14.50% (38)	14.12% (37)	4.58% (12)
● item08	262	3.0	3.0	1.2			14.12% (37)	22.90% (60)	20.99% (55)	32.06% (84)	9.92% (26)
● item09	262	2.9	3.0	1.1			12.98% (34)	22.14% (58)	34.35% (90)	26.34% (69)	4.20% (11)
● item10	262	2.6	3.0	1.1			17.56% (46)	29.39% (77)	29.01% (76)	21.37% (56)	2.67% (7)
● item11	262	2.1	2.0	0.9			25.95% (68)	46.56% (122)	20.23% (53)	5.34% (14)	1.91% (5)
● item12	262	2.3	2.0	1.2			32.06% (84)	30.92% (81)	19.08% (50)	11.45% (30)	6.49% (17)
● item16	262	1.6	1.0	1.3			78.24% (205)	3.44% (9)	4.20% (11)	5.73% (15)	8.40% (22)
● item17	262	1.6	1.0	1.2			80.15% (210)	3.44% (9)	5.34% (14)	2.67% (7)	8.40% (22)
● item25	262	2.5	2.0	1.4			32.82% (86)	18.32% (48)	21.76% (57)	16.79% (44)	10.31% (27)
● item27	262	4.0	4.0	1.2			6.11% (16)	7.25% (19)	8.02% (21)	33.59% (88)	45.04% (118)
● item30	262	1.4	1.0	1.1			83.59% (219)	2.67% (7)	4.20% (11)	6.11% (16)	3.44% (9)
● item32	262	3.4	4.0	1.7			25.95% (68)	4.20% (11)	11.45% (30)	16.79% (44)	41.60% (109)
● item33	262	3.1	3.0	1.7			32.44% (85)	6.11% (16)	11.83% (31)	14.12% (37)	35.50% (93)
● item35	262	3.6	5.0	1.8			27.48% (72)	2.67% (7)	7.25% (19)	6.49% (17)	56.11% (147)
● item36	262	1.6	1.0	1.3			80.53% (211)	3.44% (9)	3.05% (8)	3.44% (9)	9.54% (25)
● item38	262	4.3	5.0	1.1			4.20% (11)	7.63% (20)	6.49% (17)	21.37% (56)	60.31% (158)
● item40	262	2.5	2.0	1.3			30.92% (81)	27.10% (71)	18.70% (49)	12.21% (32)	11.07% (29)
● item41	262	1.2	1.0	0.7			90.08% (236)	3.82% (10)	2.29% (6)	2.67% (7)	1.15% (3)
● item45	262	2.0	1.0	1.4			64.12% (168)	5.34% (14)	9.54% (25)	11.83% (31)	9.16% (24)
● item46	262	1.6	1.0	1.2			75.57% (198)	2.67% (7)	8.02% (21)	9.54% (25)	4.20% (11)
<sup>1</sup> Histogram											
<sup>2</sup> Density											

Figure 5. Summary Descriptives of CFA and EFA Sample

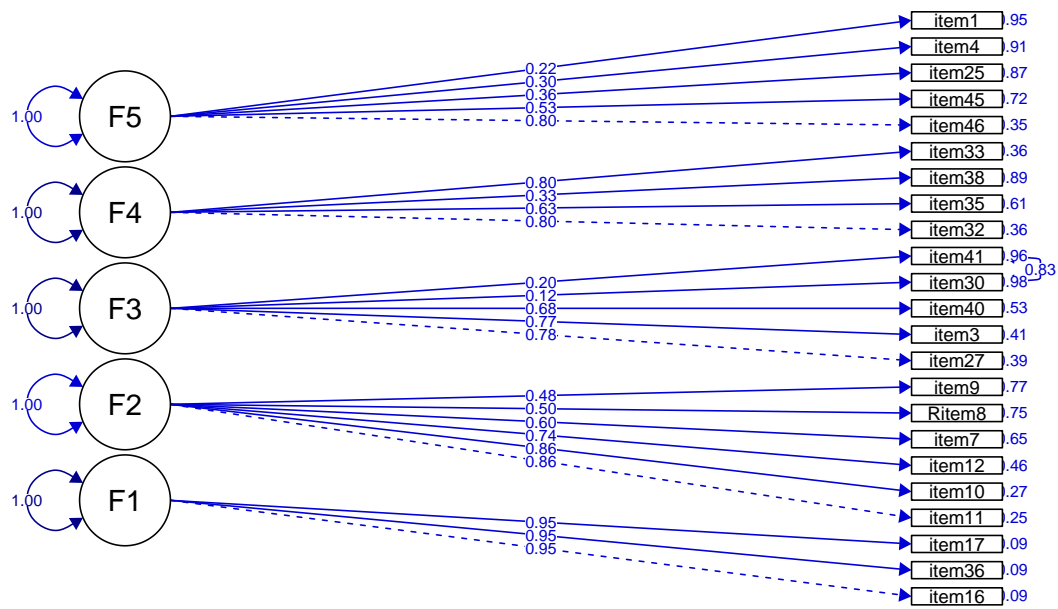
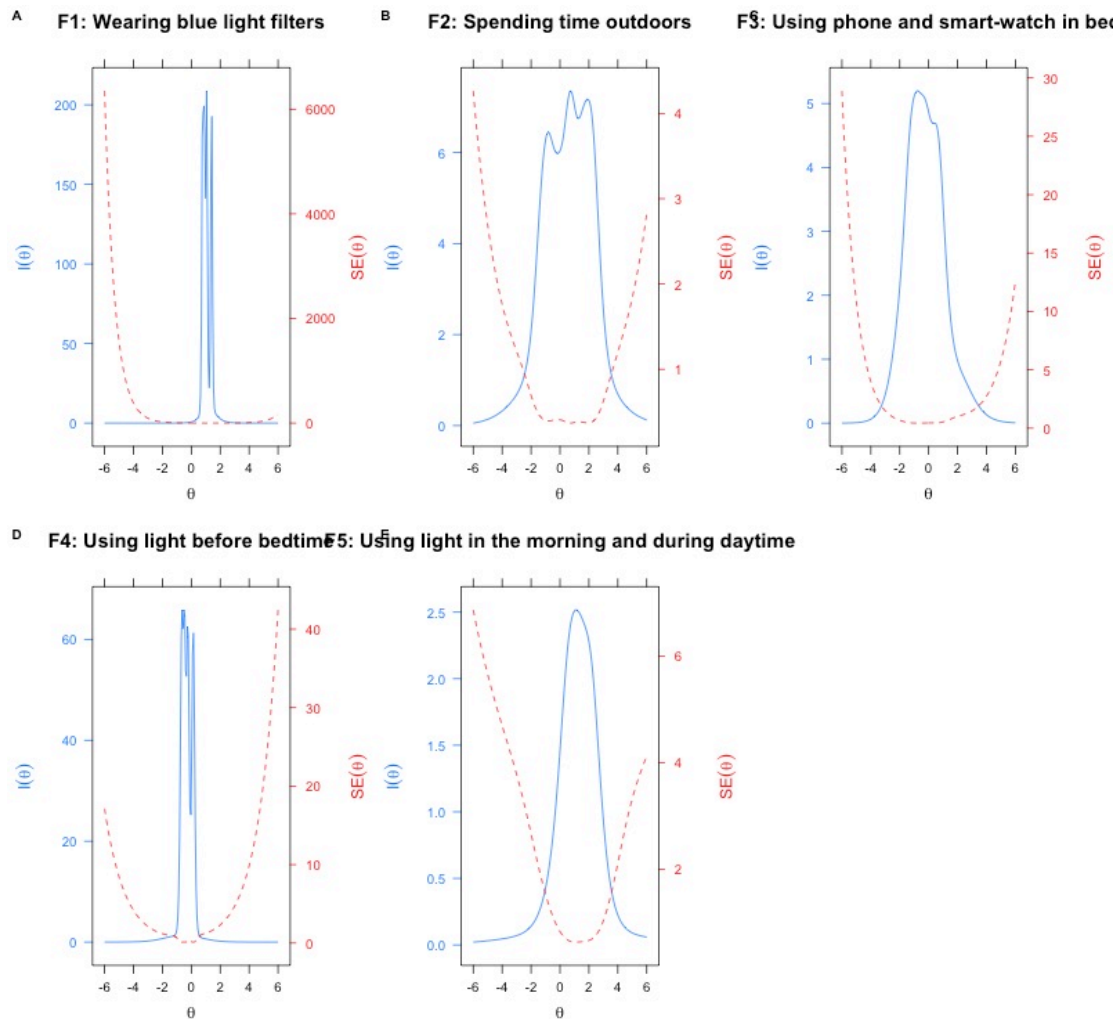


Figure 6. Five Factor CFA Model of LEBA



*Figure 7.* Test information curves (a) Wearing blue light filters (b) Spending time outdoors (c) Using phone and smartwatch in bed (d) Using light before bedtime (e) Using light in the morning and during daytime

Table A1

*Map Statistics*

MAP Statistics	dof	chisq	fit	RMSEA	BIC	eChisq	SRMR
.01125	1,080.00	4,344.31	0.18	0.08	-2,199.54	8,678.73	0.09
.01062	1,033.00	3,735.35	0.30	0.08	-2,523.72	6,414.94	0.08
.01077	987.00	3,065.44	0.38	0.07	-2,914.91	5,022.94	0.07
.01042	942.00	2,661.78	0.45	0.07	-3,045.92	3,969.03	0.06
.00938	898.00	2,237.56	0.51	0.06	-3,203.53	2,971.15	0.06
.00943	855.00	2,040.02	0.56	0.06	-3,140.53	2,441.92	0.05
.00973	813.00	1,861.69	0.59	0.05	-3,064.37	2,063.72	0.05
.00999	772.00	1,620.64	0.62	0.05	-3,057.00	1,707.87	0.04

## Appendix A



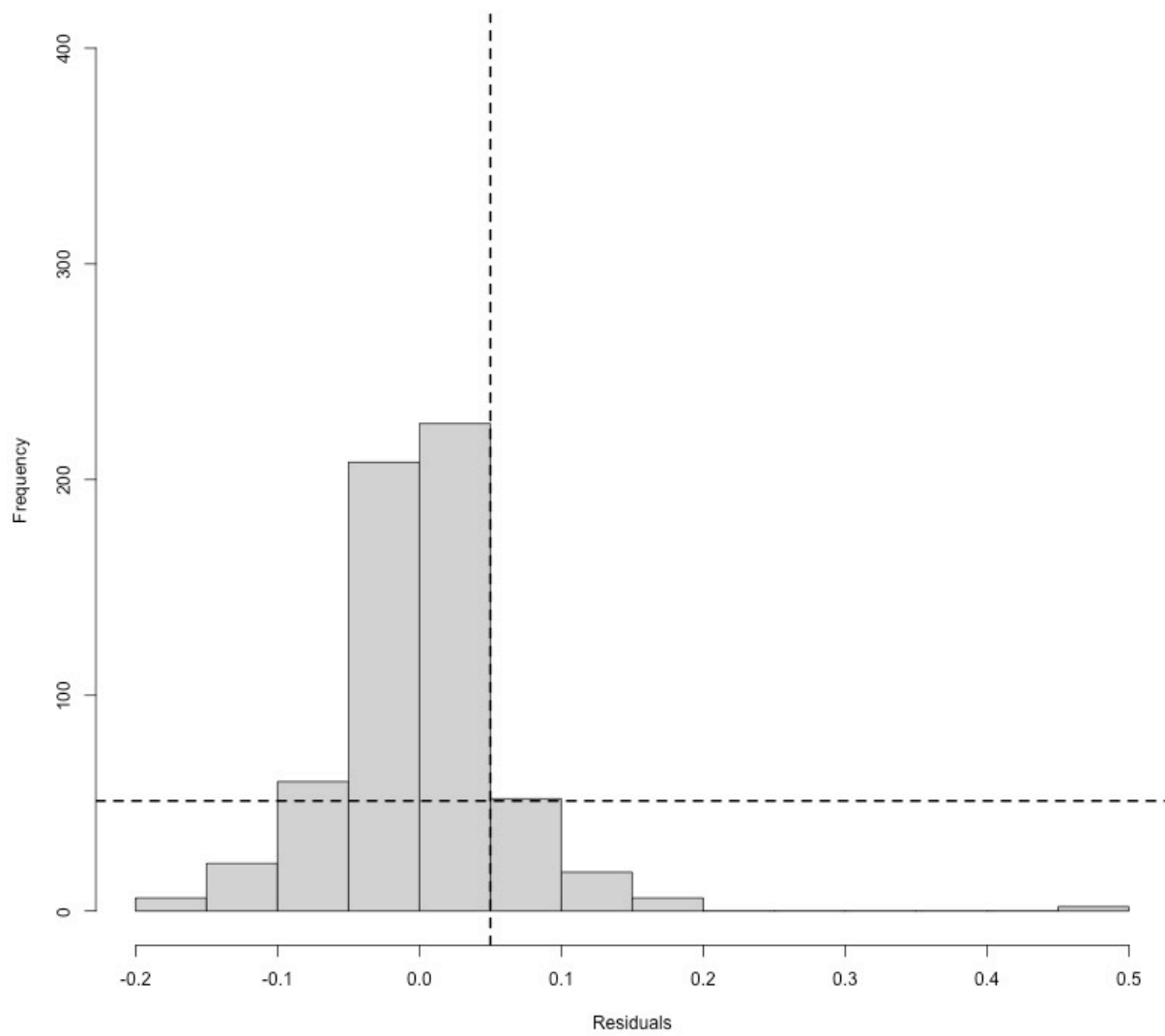
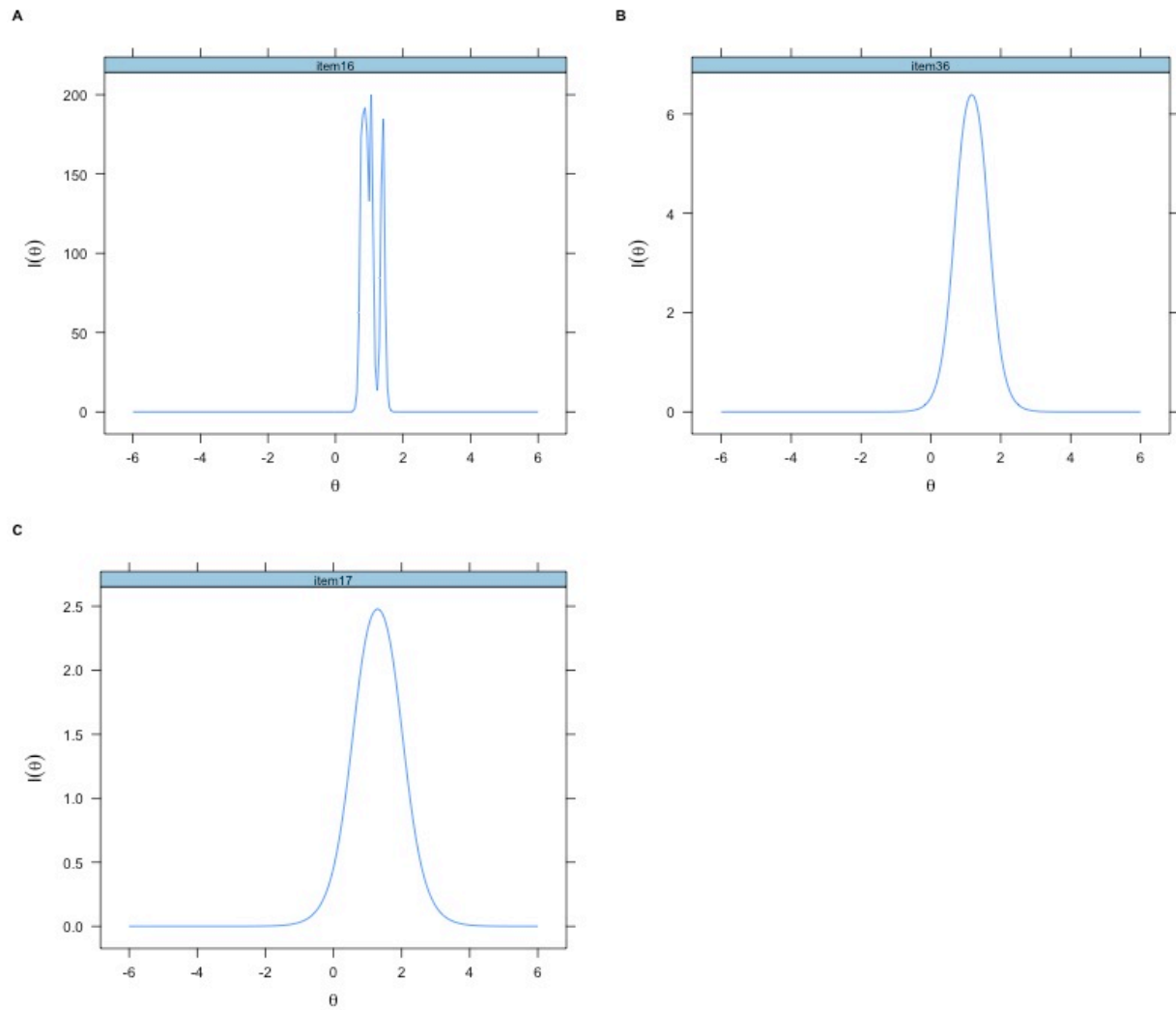


Figure A1. Histogram of residuals: five-factor solution

**F1: Wearing blue light filters**

*Figure A2.* Item information curve of LEBA F1

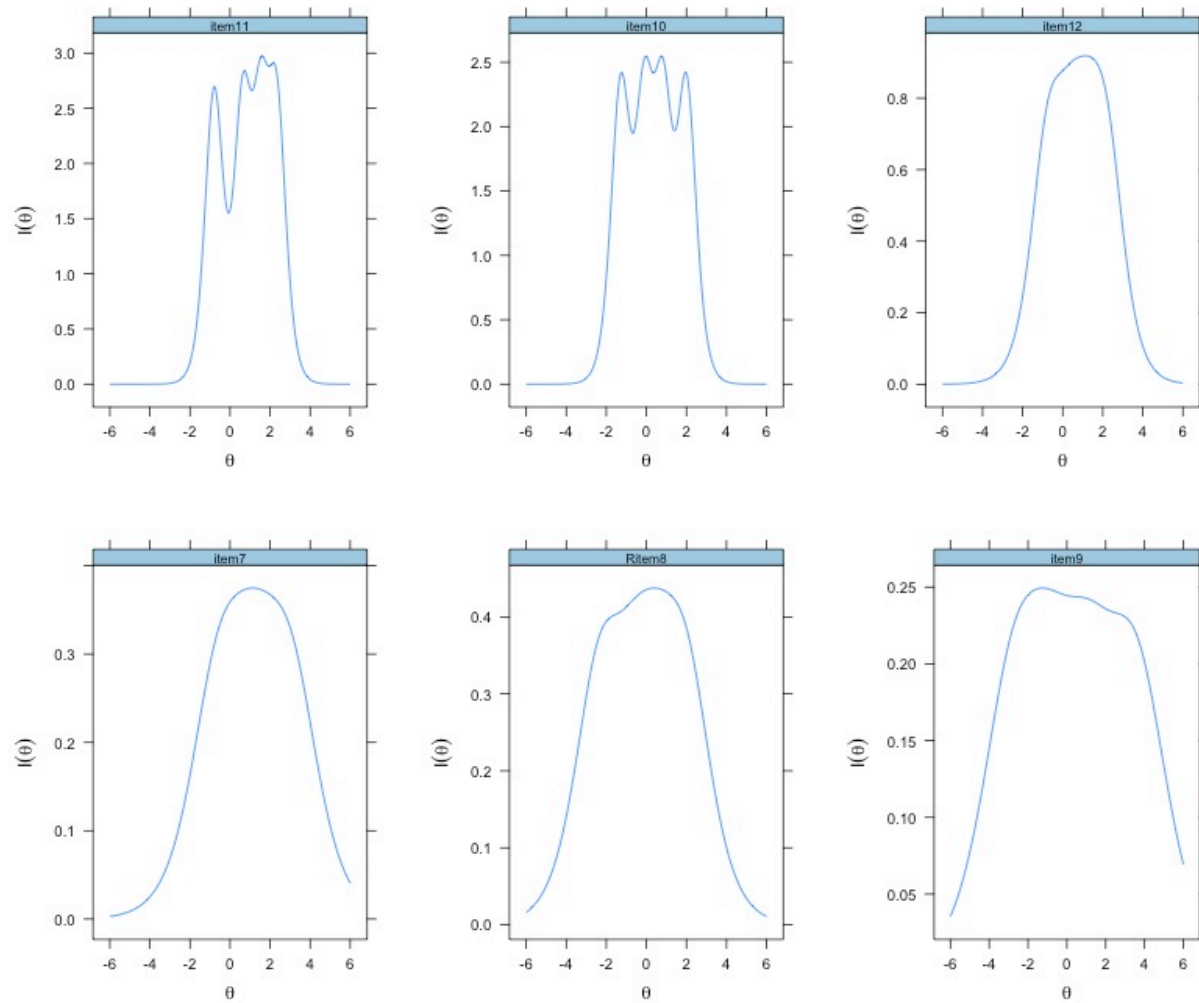
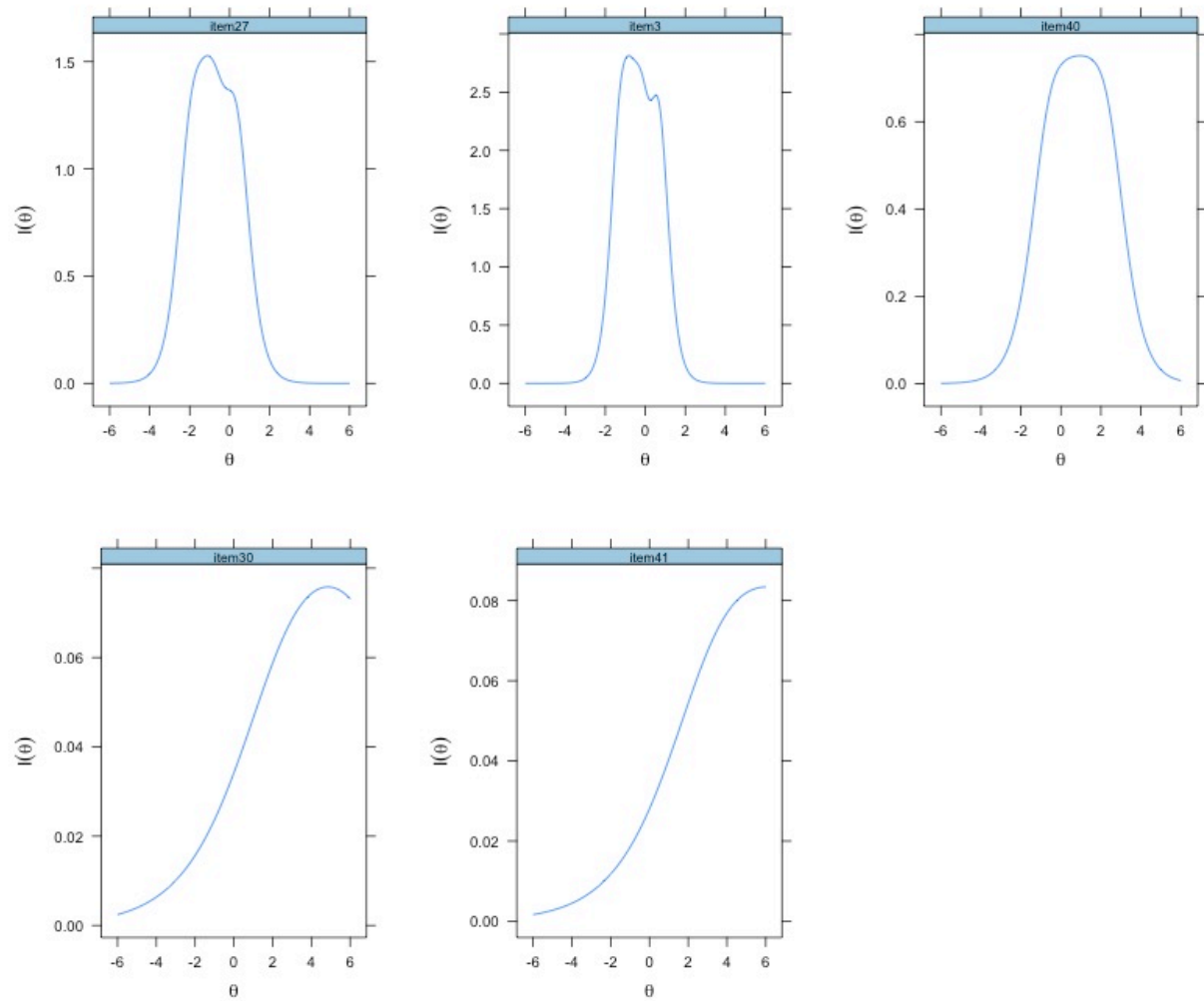
**F2: Spending time outdoors**

Figure A3. Item information curve of LEBA F1

**F3: Using phone and smart-watch in bed**

*Figure A4.* Item information curve of LEBA F1

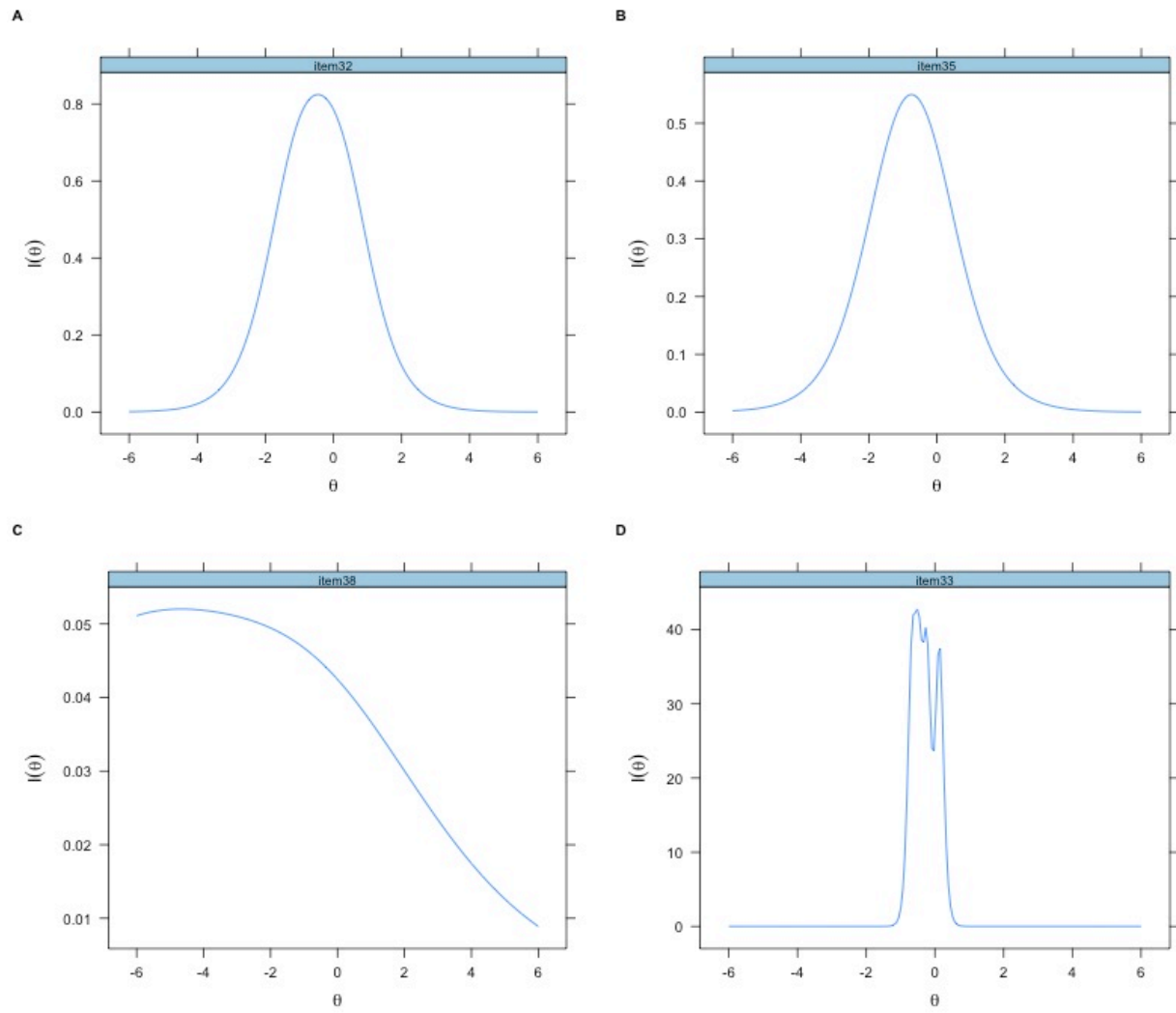
**F4: Using light before bedtime**

Figure A5. Item information curve of LEBA F1

F5: Using light...daytime

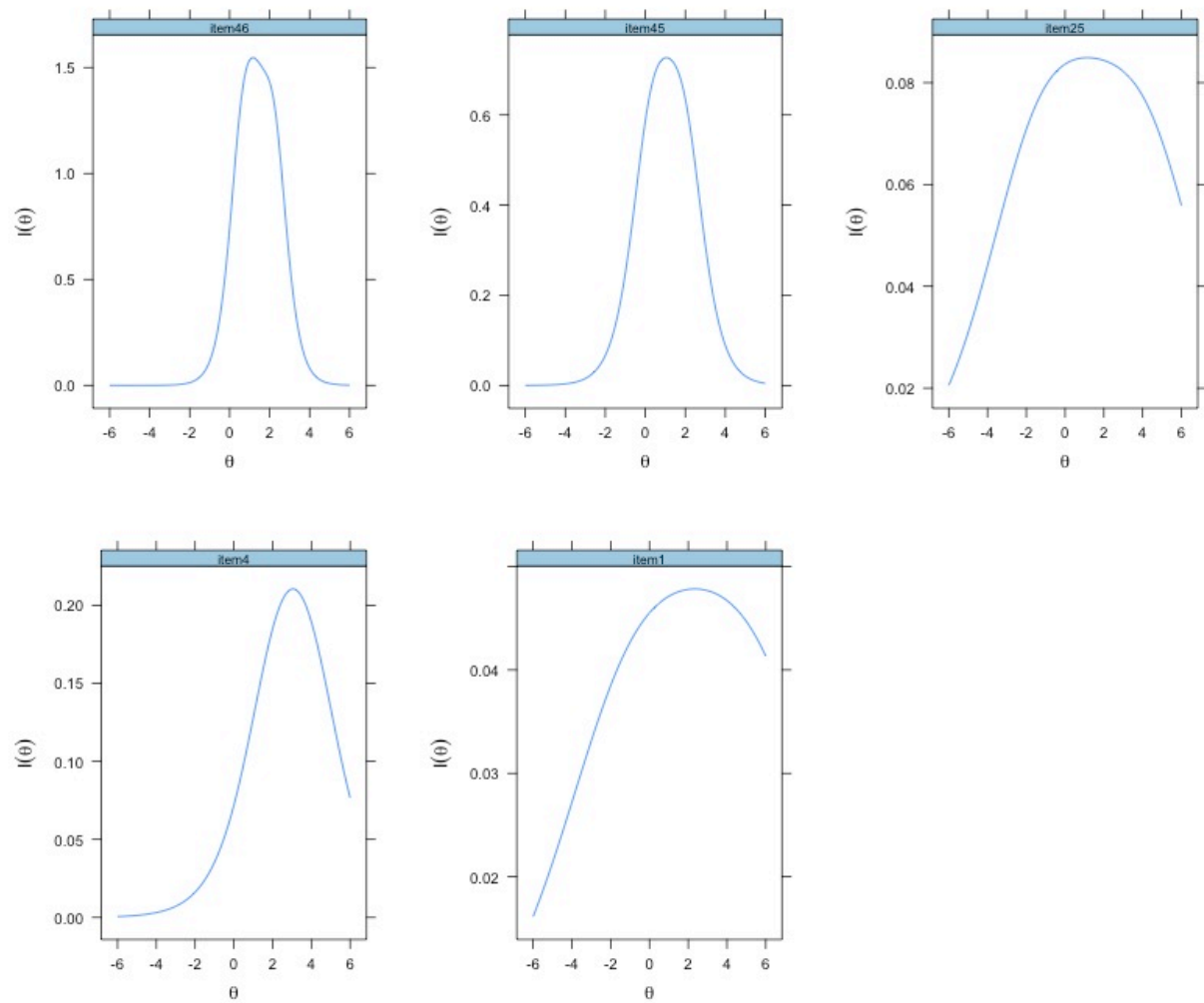
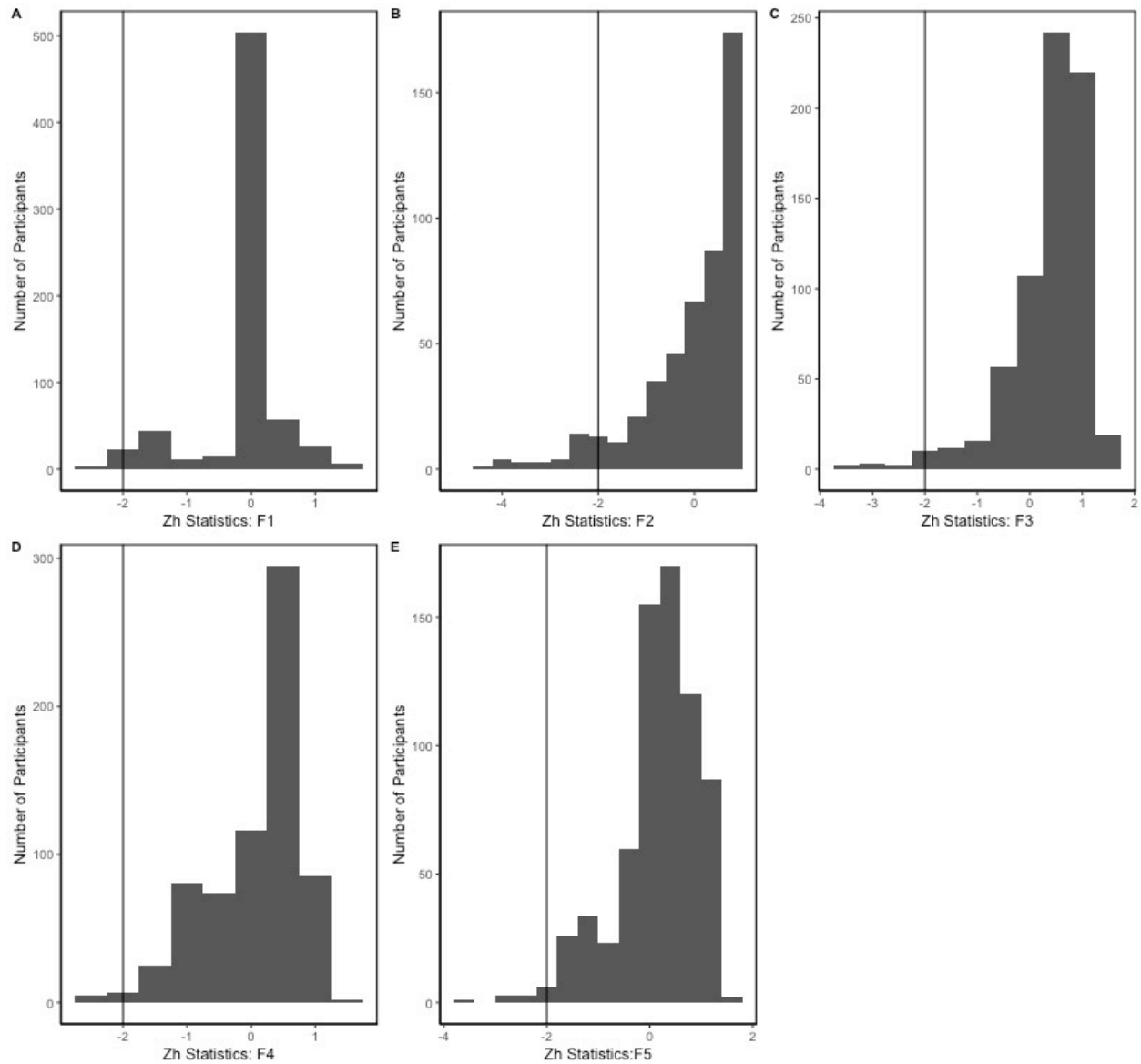


Figure A6. Item information curve of LEBA F1



*Figure A7.* Person fit of the five fitted IRT models (a) Wearing blue light filters (b) Spending time outdoors (c) Using phone and smartwatch in bed (d) Using light before bedtime (e) Using light in the morning and during daytime

Table A2

*Demographic Characteristics: Native English Speakers*

Variable	Overall, N = 262	Yes, N = 129	No, N = 133	p-value	q-value
Age	32.89 (13.66)	34.08 (15.32)	31.74 (11.77)	0.5	0.6
Sex				0.002	0.009
Female	136 (52%)	80 (62%)	56 (42%)		
Male	121 (46%)	48 (37%)	73 (55%)		
Other	5 (1.9%)	1 (0.8%)	4 (3.0%)		
Occupational Status				0.7	0.7
Work	161 (61%)	76 (59%)	85 (64%)		
School	52 (20%)	27 (21%)	25 (19%)		
Neither	49 (19%)	26 (20%)	23 (17%)		
Occupational setting				0.4	0.6
Home office/Home schooling	109 (42%)	50 (39%)	59 (44%)		
Face-to-face work/Face-to-face schooling	41 (16%)	22 (17%)	19 (14%)		
Combination of home- and face-to-face- work/schooling	53 (20%)	23 (18%)	30 (23%)		
Neither (no work or school, or in vacation)	59 (23%)	34 (26%)	25 (19%)		

<sup>1</sup> Mean (SD); n (%)<sup>2</sup> False discovery rate correction for multiple testing<sup>3</sup> Wilcoxon rank sum test<sup>4</sup> Fisher's exact test<sup>5</sup> Pearson's Chi-squared test



## Appendix B

Table B1

*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

Table B1 continued

item	MR1	MR2	MR3	MR4	MR5	Communality	Uniqueness
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 C

## Factor analysis with six factors

Table C1

*Factor loadings and communality of the retained items(six factor)*

item	PA1	PA2	PA3	PA4	PA5	PA6	Communality	Uniqueness
item16	0.99						0.987	0.013
item36	0.94						0.896	0.104
item17	0.8						0.674	0.326
item11		0.82					0.698	0.302
item10		0.81					0.656	0.344
item12		0.64					0.467	0.533
item8		-0.48					0.254	0.746
item7		0.47					0.257	0.743
item9		0.33					0.122	0.878
item33			0.97				0.978	0.022
item32			0.77				0.69	0.31
item35			0.54			0.3	0.408	0.592
item31			0.49				0.332	0.668
item3				0.84			0.728	0.272
item27				0.81			0.666	0.334
item40				0.69			0.535	0.465
item46					0.65		0.525	0.475
item45					0.57		0.355	0.645
item4					0.48		0.332	0.668
item25					0.4		0.238	0.762
item1					0.35		0.134	0.866
item26					0.35		0.161	0.839
item37						-0.8	0.682	0.318

Table C1 continued

item	PA1	PA2	PA3	PA4	PA5	PA6	Communality	Uniqueness
item38						0.39	0.245	0.755
% of Variance	0.11	0.1	0.09	0.09	0.06	0.05		

*Note.* Only loading higher than .30 is reported; Sixth factor has only two salient loadings

827

Table C2

*Factor loadings and communality of the retained items in five factor solution [Unmerged Responses]*

item	PA1	PA2	PA5	PA3	PA4	Communality	Uniqueness
item19	0.99					1.007	-0.007
item20	0.91					0.874	0.126
item18	0.82					0.711	0.289
item21	0.8					0.683	0.317
item4	0.47					0.25	0.75
item11		0.83				0.687	0.313
item10		0.81				0.67	0.33
item12		0.56				0.371	0.629
item8		-0.44				0.206	0.794
item7		0.42				0.226	0.774
item9		0.33				0.115	0.885
item16			0.95			0.946	0.054
item17			0.74			0.595	0.405
item36	0.3		0.73			0.653	0.347

Table C2 continued

item	PA1	PA2	PA5	PA3	PA4	Communality	Uniqueness
item3				0.85		0.746	0.254
item27				0.78		0.624	0.376
item40				0.71		0.512	0.488
item35					0.58	0.351	0.649
item48					0.57	0.354	0.646
item33					0.55	0.32	0.68
item47					0.52	0.294	0.706
item44					0.45	0.216	0.784
item31					0.41	0.206	0.794
item38					0.33	0.129	0.871
% of Variance	0.15	0.09	0.09	0.08	0.08		

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

828

Table C3

*Factor loadings and communality of the retained items in six factor solution  
[Unmerged Responses]*

item	PA1	PA2	PA3	PA4	PA6	PA5	Communality	Uniqueness
item19	0.98						0.995	0.005
item20	0.92						0.904	0.096
item21	0.79						0.666	0.334
item4	0.49						0.296	0.704
item43	0.32					0.31	0.282	0.718
item10		0.81					0.67	0.33

Table C3 continued

item	PA1	PA2	PA3	PA4	PA6	PA5	Communality	Uniqueness
item11		0.81					0.668	0.332
item12		0.58					0.408	0.592
item8		-0.45					0.218	0.782
item7		0.42					0.229	0.771
item9		0.33					0.115	0.885
item3			0.85				0.731	0.269
item27			0.77				0.606	0.394
item40			0.72				0.533	0.467
item35				0.64			0.426	0.574
item33				0.62			0.413	0.587
item48				0.52			0.305	0.695
item47				0.48			0.259	0.741
item31				0.39			0.206	0.794
item38				0.32			0.18	0.82
item17					0.85		0.786	0.214
item16					0.78		0.681	0.319
item13						0.57	0.336	0.664
item14						0.5	0.356	0.644
item15						0.48	0.277	0.723
item42						0.37	0.168	0.832
item26							0.064	0.936
% of Variance	0.11	0.08	0.07	0.06	0.06	0.05		

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

## 830 Items Retained in the Five Factor Solution [Unmerged Responses]

---

**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.

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.

---

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


---

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.

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

---

Table C5

*Geographical Location*

	<b>**N =</b>
	<b>690**</b>
<hr/>	
<b>__Time zone - Country__</b>	
United States - America/New_York (UTC -04:00)	63 (9.1%)
United Kingdom - Europe/London (UTC)	57 (8.3%)
Germany - Europe/Berlin (UTC +01:00)	53 (7.7%)
India - Asia/Kolkata (UTC +05:30)	38 (5.5%)
United States - America/Los_Angeles (UTC -07:00)	37 (5.4%)
United States - America/Chicago (UTC -05:00)	30 (4.3%)
France - Europe/Paris (UTC +01:00)	22 (3.2%)
Switzerland - Europe/Zurich (UTC +01:00)	21 (3.0%)
Brazil - America/Sao_Paulo (UTC -03:00)	19 (2.8%)
Netherlands - Europe/Amsterdam (UTC +01:00)	19 (2.8%)
Canada - America/Toronto (UTC -04:00)	16 (2.3%)



Table C5

*Geographical Location (continued)*

	**N =
	690**
Poland - Europe/Warsaw (UTC +01:00)	15 (2.2%)
Canada - America/Edmonton (UTC -06:00)	14 (2.0%)
Finland - Europe/Helsinki (UTC +02:00)	9 (1.3%)
Indonesia - Asia/Jakarta (UTC +07:00)	9 (1.3%)
Italy - Europe/Rome (UTC +01:00)	9 (1.3%)
Chile - America/Santiago (UTC -03:00)	8 (1.2%)
Russian Federation - Europe/Moscow (UTC +03:00)	8 (1.2%)
China - Asia/Shanghai (UTC +08:00)	7 (1.0%)
Malaysia - Asia/Kuala_Lumpur (UTC +08:00)	7 (1.0%)
Spain - Europe/Madrid (UTC +01:00)	7 (1.0%)
United States - America/Phoenix (UTC -07:00)	7 (1.0%)
Canada - America/Vancouver (UTC -07:00)	6 (0.9%)
New Zealand - Pacific/Auckland (UTC +13:00)	6 (0.9%)
Philippines - Asia/Manila (UTC +08:00)	6 (0.9%)
Turkey - Europe/Istanbul (UTC +03:00)	6 (0.9%)
United States - America/Denver (UTC -06:00)	6 (0.9%)
United States - America/Detroit (UTC -04:00)	6 (0.9%)
Argentina - America/Argentina/Buenos_Aires (UTC -03:00)	5 (0.7%)
Australia - Australia/Melbourne (UTC +11:00)	5 (0.7%)
Ireland - Europe/Dublin (UTC)	5 (0.7%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Lithuania - Europe/Vilnius (UTC +02:00)	5 (0.7%)
South Africa - Africa/Johannesburg (UTC +02:00)	5 (0.7%)
Australia - Australia/Brisbane (UTC +10:00)	4 (0.6%)
Belgium - Europe/Brussels (UTC +01:00)	4 (0.6%)
Israel - Asia/Jerusalem (UTC +02:00)	4 (0.6%)
Sweden - Europe/Stockholm (UTC +01:00)	4 (0.6%)
United States - America/Boise (UTC -06:00)	4 (0.6%)
Czech Republic - Europe/Prague (UTC +01:00)	3 (0.4%)
Denmark - Europe/Copenhagen (UTC +01:00)	3 (0.4%)
Germany - Europe/Busingen (UTC +01:00)	3 (0.4%)
Greece - Europe/Athens (UTC +02:00)	3 (0.4%)
Iran	3 (0.4%)
Japan - Asia/Tokyo (UTC +09:00)	3 (0.4%)
Norway - Europe/Oslo (UTC +01:00)	3 (0.4%)
Romania - Europe/Bucharest (UTC +02:00)	3 (0.4%)
Serbia - Europe/Belgrade (UTC +01:00)	3 (0.4%)
Slovenia - Europe/Ljubljana (UTC +01:00)	3 (0.4%)
Taiwan	3 (0.4%)
United States - America/Anchorage (UTC -08:00)	3 (0.4%)
United States - America/Indiana/Indianapolis (UTC -04:00)	3 (0.4%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
United States - America/Kentucky/Louisville (UTC -04:00)	3 (0.4%)
Argentina - America/Argentina/Cordoba (UTC -03:00)	2 (0.3%)
Australia - Australia/Adelaide (UTC +10:30)	2 (0.3%)
Australia - Australia/Perth (UTC +08:00)	2 (0.3%)
Australia - Australia/Sydney (UTC +11:00)	2 (0.3%)
Brazil - America/Araguaina (UTC -03:00)	2 (0.3%)
Brazil - America/Bahia (UTC -03:00)	2 (0.3%)
Canada - America/Moncton (UTC -03:00)	2 (0.3%)
Colombia - America/Bogota (UTC -05:00)	2 (0.3%)
Costa Rica - America/Costa_Rica (UTC -06:00)	2 (0.3%)
Croatia - Europe/Zagreb (UTC +01:00)	2 (0.3%)
Ecuador - America/Guayaquil (UTC -05:00)	2 (0.3%)
Estonia - Europe/Tallinn (UTC +02:00)	2 (0.3%)
Hong Kong - Asia/Hong_Kong (UTC +08:00)	2 (0.3%)
Hungary - Europe/Budapest (UTC +01:00)	2 (0.3%)
Jordan - Asia/Amman (UTC +03:00)	2 (0.3%)
Latvia - Europe/Riga (UTC +02:00)	2 (0.3%)
Malaysia - Asia/Kuching (UTC +08:00)	2 (0.3%)
Mexico - America/Mexico_City (UTC -06:00)	2 (0.3%)
Nepal - Asia/Kathmandu (UTC +05:45)	2 (0.3%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Portugal - Europe/Lisbon (UTC)	2 (0.3%)
Slovakia - Europe/Bratislava (UTC +01:00)	2 (0.3%)
Spain - Africa/Ceuta (UTC +01:00)	2 (0.3%)
Sudan - Africa/Khartoum (UTC +02:00)	2 (0.3%)
United States - America/Adak (UTC -09:00)	2 (0.3%)
United States - Pacific/Honolulu (UTC -10:00)	2 (0.3%)
Viet Nam - Asia/Ho_Chi_Minh (UTC +07:00), British - America/Tortola (UTC -04:00)	2 (0.3%)
Albania - Europe/Tirane (UTC +01:00)	1 (0.1%)
Argentina - America/Argentina/Jujuy (UTC -03:00)	1 (0.1%)
Australia - Antarctica/Macquarie (UTC +11:00)	1 (0.1%)
Australia - Australia/Darwin (UTC +09:30)	1 (0.1%)
Austria - Europe/Vienna (UTC +01:00)	1 (0.1%)
Bangladesh - Asia/Dhaka (UTC +06:00)	1 (0.1%)
Brazil - America/Cuiaba (UTC -04:00)	1 (0.1%)
Brazil - America/Fortaleza (UTC -03:00)	1 (0.1%)
Bulgaria - Europe/Sofia (UTC +02:00)	1 (0.1%)
Cameroon - Africa/Douala (UTC +01:00)	1 (0.1%)
Canada - America/Blanc-Sablon (UTC -04:00)	1 (0.1%)
Canada - America/Halifax (UTC -03:00)	1 (0.1%)
Canada - America/Resolute (UTC -05:00)	1 (0.1%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Cayman Islands - America/Cayman (UTC -05:00)	1 (0.1%)
Chile - Pacific/Easter (UTC -05:00)	1 (0.1%)
Cyprus - Asia/Famagusta (UTC +02:00)	1 (0.1%)
Guatemala - America/Guatemala (UTC -06:00)	1 (0.1%)
Korea, Republic of - Asia/Seoul (UTC +09:00)	1 (0.1%)
Macedonia	1 (0.1%)
Martinique - America/Martinique (UTC -04:00)	1 (0.1%)
Mexico - America/Monterrey (UTC -06:00)	1 (0.1%)
Mongolia - Asia/Ulaanbaatar (UTC +08:00)	1 (0.1%)
Myanmar - Asia/Yangon (UTC +06:30)	1 (0.1%)
New Zealand - Pacific/Chatham (UTC +13:45)	1 (0.1%)
Nigeria - Africa/Lagos (UTC +01:00)	1 (0.1%)
Pakistan - Asia/Karachi (UTC +05:00)	1 (0.1%)
Panama - America/Panama (UTC -05:00)	1 (0.1%)
Russian Federation - Asia/Barnaul (UTC +07:00)	1 (0.1%)
Russian Federation - Asia/Novosibirsk (UTC +07:00)	1 (0.1%)
Russian Federation - Asia/Tomsk (UTC +07:00)	1 (0.1%)
Russian Federation - Asia/Vladivostok (UTC +10:00)	1 (0.1%)
Russian Federation - Asia/Yekaterinburg (UTC +05:00)	1 (0.1%)
Saudi Arabia - Asia/Riyadh (UTC +03:00)	1 (0.1%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Singapore - Asia/Singapore (UTC +08:00)	1 (0.1%)
Spain - Atlantic/Canary (UTC)	1 (0.1%)
Tanzania	1 (0.1%)
Ukraine - Europe/Kiev (UTC +02:00)	1 (0.1%)
United States - America/Indiana/Tell_City (UTC -05:00)	1 (0.1%)
United States - America/North_Dakota/Center (UTC -05:00)	1 (0.1%)
United States - America/North_Dakota/New_Salem (UTC -05:00)	1 (0.1%)
Aland Islands - Europe/Mariehamn (UTC +02:00)	0 (0%)
Afghanistan - Asia/Kabul (UTC +04:30)	0 (0%)
Algeria - Africa/Algiers (UTC +01:00)	0 (0%)
American Samoa - Pacific/Pago_Pago (UTC -11:00)	0 (0%)
Andorra - Europe/Andorra (UTC +01:00)	0 (0%)
Angola - Africa/Luanda (UTC +01:00)	0 (0%)
Anguilla - America/Anguilla (UTC -04:00)	0 (0%)
Antarctica - Antarctica/Casey (UTC +11:00)	0 (0%)
Antarctica - Antarctica/Davis (UTC +07:00)	0 (0%)
Antarctica - Antarctica/DumontDUrville (UTC +10:00)	0 (0%)
Antarctica - Antarctica/Mawson (UTC +05:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Antarctica - Antarctica/McMurdo (UTC +13:00)	0 (0%)
Antarctica - Antarctica/Palmer (UTC -03:00)	0 (0%)
Antarctica - Antarctica/Rothera (UTC -03:00)	0 (0%)
Antarctica - Antarctica/Syowa (UTC +03:00)	0 (0%)
Antarctica - Antarctica/Troll (UTC)	0 (0%)
Antarctica - Antarctica/Vostok (UTC +06:00)	0 (0%)
Antigua and Barbuda - America/Antigua (UTC -04:00)	0 (0%)
Argentina - America/Argentina/Catamarca (UTC -03:00)	0 (0%)
Argentina - America/Argentina/La_Rioja (UTC -03:00)	0 (0%)
Argentina - America/Argentina/Mendoza (UTC -03:00)	0 (0%)
Argentina - America/Argentina/Rio_Gallegos (UTC -03:00)	0 (0%)
Argentina - America/Argentina/Salta (UTC -03:00)	0 (0%)
Argentina - America/Argentina/San_Juan (UTC -03:00)	0 (0%)
Argentina - America/Argentina/San_Luis (UTC -03:00)	0 (0%)
Argentina - America/Argentina/Tucuman (UTC -03:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Argentina - America/Argentina/Ushuaia (UTC -03:00)	0 (0%)
Armenia - Asia/Yerevan (UTC +04:00)	0 (0%)
Aruba - America/Aruba (UTC -04:00)	0 (0%)
Australia - Australia/Broken_Hill (UTC +10:30)	0 (0%)
Australia - Australia/Currie (UTC +11:00)	0 (0%)
Australia - Australia/Eucla (UTC +08:45)	0 (0%)
Australia - Australia/Hobart (UTC +11:00)	0 (0%)
Australia - Australia/Lindeman (UTC +10:00)	0 (0%)
Australia - Australia/Lord_Howe (UTC +11:00)	0 (0%)
Azerbaijan - Asia/Baku (UTC +04:00)	0 (0%)
Bahamas - America/Nassau (UTC -04:00)	0 (0%)
Bahrain - Asia/Bahrain (UTC +03:00)	0 (0%)
Barbados - America/Barbados (UTC -04:00)	0 (0%)
Belarus - Europe/Minsk (UTC +03:00)	0 (0%)
Belize - America/Belize (UTC -06:00)	0 (0%)
Benin - Africa/Porto-Novo (UTC +01:00)	0 (0%)
Bermuda - Atlantic/Bermuda (UTC -03:00)	0 (0%)
Bhutan - Asia/Thimphu (UTC +06:00),Plurinational	0 (0%)
State of - America/La_Paz (UTC -04:00)	
Bolivia,Sint Eustatius and Saba - America/Kralendijk	0 (0%)
(UTC -04:00)	



Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Bonaire	0 (0%)
Bosnia and Herzegovina - Europe/Sarajevo (UTC +01:00)	0 (0%)
Botswana - Africa/Gaborone (UTC +02:00)	0 (0%)
Brazil - America/Belem (UTC -03:00)	0 (0%)
Brazil - America/Boa_Vista (UTC -04:00)	0 (0%)
Brazil - America/Campo_Grande (UTC -04:00)	0 (0%)
Brazil - America/Eirunepe (UTC -05:00)	0 (0%)
Brazil - America/Maceio (UTC -03:00)	0 (0%)
Brazil - America/Manaus (UTC -04:00)	0 (0%)
Brazil - America/Noronha (UTC -02:00)	0 (0%)
Brazil - America/Porto_Velho (UTC -04:00)	0 (0%)
Brazil - America/Recife (UTC -03:00)	0 (0%)
Brazil - America/Rio_Branco (UTC -05:00)	0 (0%)
Brazil - America/Santarem (UTC -03:00)	0 (0%)
British Indian Ocean Territory - Indian/Chagos (UTC +06:00)	0 (0%)
Brunei Darussalam - Asia/Brunei (UTC +08:00)	0 (0%)
Burkina Faso - Africa/Ouagadougou (UTC)	0 (0%)
Burundi - Africa/Bujumbura (UTC +02:00)	0 (0%)
Cambodia - Asia/Phnom_Penh (UTC +07:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Canada - America/Atikokan (UTC -05:00)	0 (0%)
Canada - America/Cambridge_Bay (UTC -06:00)	0 (0%)
Canada - America/Creston (UTC -07:00)	0 (0%)
Canada - America/Dawson (UTC -07:00)	0 (0%)
Canada - America/Dawson_Creek (UTC -07:00)	0 (0%)
Canada - America/Fort_Nelson (UTC -07:00)	0 (0%)
Canada - America/Glace_Bay (UTC -03:00)	0 (0%)
Canada - America/Goose_Bay (UTC -03:00)	0 (0%)
Canada - America/Inuvik (UTC -06:00)	0 (0%)
Canada - America/Iqaluit (UTC -04:00)	0 (0%)
Canada - America/Nipigon (UTC -04:00)	0 (0%)
Canada - America/Pangnirtung (UTC -04:00)	0 (0%)
Canada - America/Rainy_River (UTC -05:00)	0 (0%)
Canada - America/Rankin_Inlet (UTC -05:00)	0 (0%)
Canada - America/Regina (UTC -06:00)	0 (0%)
Canada - America/St_Johns (UTC -02:30)	0 (0%)
Canada - America/Swift_Current (UTC -06:00)	0 (0%)
Canada - America/Thunder_Bay (UTC -04:00)	0 (0%)
Canada - America/Whitehorse (UTC -07:00)	0 (0%)
Canada - America/Winnipeg (UTC -05:00)	0 (0%)
Canada - America/Yellowknife (UTC -06:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Cape Verde - Atlantic/Cape_Verde (UTC -01:00)	0 (0%)
Central African Republic - Africa/Bangui (UTC +01:00)	0 (0%)
Chad - Africa/Ndjamena (UTC +01:00)	0 (0%)
Chile - America/Punta_Arenas (UTC -03:00)	0 (0%)
China - Asia/Urumqi (UTC +06:00)	0 (0%)
Christmas Island - Indian/Christmas (UTC +07:00)	0 (0%)
Cocos (Keeling) Islands - Indian/Cocos (UTC +06:30)	0 (0%)
Comoros - Indian/Comoro (UTC +03:00)	0 (0%)
Congo - Africa/Brazzaville (UTC +01:00),the	0 (0%)
Democratic Republic of the - Africa/Kinshasa (UTC +01:00)	
Congo,the Democratic Republic of the -	0 (0%)
Africa/Lubumbashi (UTC +02:00)	
Congo	0 (0%)
Cook Islands - Pacific/Rarotonga (UTC -10:00)	0 (0%)
Cuba - America/Havana (UTC -04:00)	0 (0%)
Curaçao - America/Curacao (UTC -04:00)	0 (0%)
Cyprus - Asia/Nicosia (UTC +02:00)	0 (0%)
Côte d'Ivoire - Africa/Abidjan (UTC)	0 (0%)
Djibouti - Africa/Djibouti (UTC +03:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Dominica - America/Dominica (UTC -04:00)	0 (0%)
Dominican Republic - America/Santo_Domingo (UTC -04:00)	0 (0%)
Ecuador - Pacific/Galapagos (UTC -06:00)	0 (0%)
Egypt - Africa/Cairo (UTC +02:00)	0 (0%)
El Salvador - America/El_Salvador (UTC -06:00)	0 (0%)
Equatorial Guinea - Africa/Malabo (UTC +01:00)	0 (0%)
Eritrea - Africa/Asmara (UTC +03:00)	0 (0%)
Ethiopia - Africa/Addis_Ababa (UTC +03:00)	0 (0%)
Falkland Islands (Malvinas) - Atlantic/Stanley (UTC -03:00)	0 (0%)
Faroe Islands - Atlantic/Faroe (UTC)	0 (0%)
Fiji - Pacific/Fiji (UTC +12:00)	0 (0%)
French Guiana - America/Cayenne (UTC -03:00)	0 (0%)
French Polynesia - Pacific/Gambier (UTC -09:00)	0 (0%)
French Polynesia - Pacific/Marquesas (UTC -09:30)	0 (0%)
French Polynesia - Pacific/Tahiti (UTC -10:00)	0 (0%)
French Southern Territories - Indian/Kerguelen (UTC +05:00)	0 (0%)
Gabon - Africa/Libreville (UTC +01:00)	0 (0%)
Gambia - Africa/Banjul (UTC)	0 (0%)
Georgia - Asia/Tbilisi (UTC +04:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	**N =
	690**
Ghana - Africa/Accra (UTC)	0 (0%)
Gibraltar - Europe/Gibraltar (UTC +01:00)	0 (0%)
Greenland - America/Danmarkshavn (UTC)	0 (0%)
Greenland - America/Nuuk (UTC -03:00)	0 (0%)
Greenland - America/Scoresbysund (UTC -01:00)	0 (0%)
Greenland - America/Thule (UTC -03:00)	0 (0%)
Grenada - America/Grenada (UTC -04:00)	0 (0%)
Guadeloupe - America/Guadeloupe (UTC -04:00)	0 (0%)
Guam - Pacific/Guam (UTC +10:00)	0 (0%)
Guernsey - Europe/Guernsey (UTC)	0 (0%)
Guinea - Africa/Conakry (UTC)	0 (0%)
Guinea-Bissau - Africa/Bissau (UTC)	0 (0%)
Guyana - America/Guyana (UTC -04:00)	0 (0%)
Haiti - America/Port-au-Prince (UTC -04:00)	0 (0%)
Holy See (Vatican City State) - Europe/Vatican (UTC +01:00)	0 (0%)
Honduras - America/Tegucigalpa (UTC -06:00)	0 (0%)
Iceland - Atlantic/Reykjavik (UTC)	0 (0%)
Indonesia - Asia/Jayapura (UTC +09:00)	0 (0%)
Indonesia - Asia/Makassar (UTC +08:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Indonesia - Asia/Pontianak (UTC +07:00),Islamic	0 (0%)
Republic of - Asia/Tehran (UTC +03:30)	
Iraq - Asia/Baghdad (UTC +03:00)	0 (0%)
Isle of Man - Europe/Isle_of_Man (UTC)	0 (0%)
Jamaica - America/Jamaica (UTC -05:00)	0 (0%)
Jersey - Europe/Jersey (UTC)	0 (0%)
Kazakhstan - Asia/Almaty (UTC +06:00)	0 (0%)
Kazakhstan - Asia/Aqtau (UTC +05:00)	0 (0%)
Kazakhstan - Asia/Aqtobe (UTC +05:00)	0 (0%)
Kazakhstan - Asia/Atyrau (UTC +05:00)	0 (0%)
Kazakhstan - Asia/Oral (UTC +05:00)	0 (0%)
Kazakhstan - Asia/Qostanay (UTC +06:00)	0 (0%)
Kazakhstan - Asia/Qyzylorda (UTC +05:00)	0 (0%)
Kenya - Africa/Nairobi (UTC +03:00)	0 (0%)
Kiribati - Pacific/Enderbury (UTC +13:00)	0 (0%)
Kiribati - Pacific/Kiritimati (UTC +14:00)	0 (0%)
Kiribati - Pacific/Tarawa (UTC +12:00),Democratic	0 (0%)
Peoples Republic of - Asia/Pyongyang (UTC +09:00)	
Korea	0 (0%)
Kuwait - Asia/Kuwait (UTC +03:00)	0 (0%)
Kyrgyzstan - Asia/Bishkek (UTC +06:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Lao Peoples Democratic Republic - Asia/Vientiane (UTC +07:00)	0 (0%)
Lebanon - Asia/Beirut (UTC +02:00)	0 (0%)
Lesotho - Africa/Maseru (UTC +02:00)	0 (0%)
Liberia - Africa/Monrovia (UTC)	0 (0%)
Libya - Africa/Tripoli (UTC +02:00)	0 (0%)
Liechtenstein - Europe/Vaduz (UTC +01:00)	0 (0%)
Luxembourg - Europe/Luxembourg (UTC +01:00)	0 (0%)
Macao - Asia/Macau (UTC +08:00),the Former Yugoslav Republic of - Europe/Skopje (UTC +01:00)	0 (0%)
Madagascar - Indian/Antananarivo (UTC +03:00)	0 (0%)
Malawi - Africa/Blantyre (UTC +02:00)	0 (0%)
Maldives - Indian/Maldives (UTC +05:00)	0 (0%)
Mali - Africa/Bamako (UTC)	0 (0%)
Malta - Europe/Malta (UTC +01:00)	0 (0%)
Marshall Islands - Pacific/Kwajalein (UTC +12:00)	0 (0%)
Marshall Islands - Pacific/Majuro (UTC +12:00)	0 (0%)
Mauritania - Africa/Nouakchott (UTC)	0 (0%)
Mauritius - Indian/Mauritius (UTC +04:00)	0 (0%)
Mayotte - Indian/Mayotte (UTC +03:00)	0 (0%)
Mexico - America/Bahia_Banderas (UTC -06:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Mexico - America/Cancun (UTC -05:00)	0 (0%)
Mexico - America/Chihuahua (UTC -07:00)	0 (0%)
Mexico - America/Hermosillo (UTC -07:00)	0 (0%)
Mexico - America/Matamoros (UTC -05:00)	0 (0%)
Mexico - America/Mazatlan (UTC -07:00)	0 (0%)
Mexico - America/Merida (UTC -06:00)	0 (0%)
Mexico - America/Ojinaga (UTC -06:00)	0 (0%)
Mexico - America/Tijuana (UTC -07:00), Federated States of - Pacific/Chuuk (UTC +10:00)	0 (0%)
Micronesia, Federated States of - Pacific/Kosrae (UTC +11:00)	0 (0%)
Micronesia, Federated States of - Pacific/Pohnpei (UTC +11:00)	0 (0%)
Micronesia, Republic of - Europe/Chisinau (UTC +02:00)	0 (0%)
Moldova	0 (0%)
Monaco - Europe/Monaco (UTC +01:00)	0 (0%)
Mongolia - Asia/Choibalsan (UTC +08:00)	0 (0%)
Mongolia - Asia/Hovd (UTC +07:00)	0 (0%)
Montenegro - Europe/Podgorica (UTC +01:00)	0 (0%)
Montserrat - America/Montserrat (UTC -04:00)	0 (0%)
Morocco - Africa/Casablanca (UTC +01:00)	0 (0%)



Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Mozambique - Africa/Maputo (UTC +02:00)	0 (0%)
Namibia - Africa/Windhoek (UTC +02:00)	0 (0%)
Nauru - Pacific/Nauru (UTC +12:00)	0 (0%)
New Caledonia - Pacific/Noumea (UTC +11:00)	0 (0%)
Nicaragua - America/Managua (UTC -06:00)	0 (0%)
Niger - Africa/Niamey (UTC +01:00)	0 (0%)
Niue - Pacific/Niue (UTC -11:00)	0 (0%)
Norfolk Island - Pacific/Norfolk (UTC +12:00)	0 (0%)
Northern Mariana Islands - Pacific/Saipan (UTC +10:00)	0 (0%)
Oman - Asia/Muscat (UTC +04:00)	0 (0%)
Palau - Pacific/Palau (UTC +09:00), State of - Asia/Gaza (UTC +02:00)	0 (0%)
Palestine, State of - Asia/Hebron (UTC +02:00)	0 (0%)
Palestine	0 (0%)
Papua New Guinea - Pacific/Bougainville (UTC +11:00)	0 (0%)
Papua New Guinea - Pacific/Port_Moresby (UTC +10:00)	0 (0%)
Paraguay - America/Asuncion (UTC -03:00)	0 (0%)
Peru - America/Lima (UTC -05:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Pitcairn - Pacific/Pitcairn (UTC -08:00)	0 (0%)
Portugal - Atlantic/Azores (UTC -01:00)	0 (0%)
Portugal - Atlantic/Madeira (UTC)	0 (0%)
Puerto Rico - America/Puerto_Rico (UTC -04:00)	0 (0%)
Qatar - Asia/Qatar (UTC +03:00)	0 (0%)
Russian Federation - Asia/Anadyr (UTC +12:00)	0 (0%)
Russian Federation - Asia/Chita (UTC +09:00)	0 (0%)
Russian Federation - Asia/Irkutsk (UTC +08:00)	0 (0%)
Russian Federation - Asia/Kamchatka (UTC +12:00)	0 (0%)
Russian Federation - Asia/Khandyga (UTC +09:00)	0 (0%)
Russian Federation - Asia/Krasnoyarsk (UTC +07:00)	0 (0%)
Russian Federation - Asia/Magadan (UTC +11:00)	0 (0%)
Russian Federation - Asia/Novokuznetsk (UTC +07:00)	0 (0%)
Russian Federation - Asia/Omsk (UTC +06:00)	0 (0%)
Russian Federation - Asia/Sakhalin (UTC +11:00)	0 (0%)
Russian Federation - Asia/Srednekolymsk (UTC +11:00)	0 (0%)
Russian Federation - Asia/Ust-Nera (UTC +10:00)	0 (0%)
Russian Federation - Asia/Yakutsk (UTC +09:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Russian Federation - Europe/Astrakhan (UTC +04:00)	0 (0%)
Russian Federation - Europe/Kaliningrad (UTC +02:00)	0 (0%)
Russian Federation - Europe/Kirov (UTC +03:00)	0 (0%)
Russian Federation - Europe/Samara (UTC +04:00)	0 (0%)
Russian Federation - Europe/Saratov (UTC +04:00)	0 (0%)
Russian Federation - Europe/Ulyanovsk (UTC +04:00)	0 (0%)
Russian Federation - Europe/Volgograd (UTC +04:00)	0 (0%)
Rwanda - Africa/Kigali (UTC +02:00)	0 (0%)
Réunion - Indian/Reunion (UTC +04:00)	0 (0%)
Saint Barthélemy - America/St_Barthelemy (UTC -04:00), Ascension and Tristan da Cunha - Atlantic/St_Helena (UTC)	0 (0%)
Saint Helena	0 (0%)
Saint Kitts and Nevis - America/St_Kitts (UTC -04:00)	0 (0%)
Saint Lucia - America/St_Lucia (UTC -04:00)	0 (0%)
Saint Martin (French part) - America/Marigot (UTC -04:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Saint Pierre and Miquelon - America/Miquelon (UTC -02:00)	0 (0%)
Saint Vincent and the Grenadines - America/St_Vincent (UTC -04:00)	0 (0%)
Samoa - Pacific/Apia (UTC +14:00)	0 (0%)
San Marino - Europe/San_Marino (UTC +01:00)	0 (0%)
Sao Tome and Principe - Africa/Sao_Tome (UTC)	0 (0%)
Senegal - Africa/Dakar (UTC)	0 (0%)
Seychelles - Indian/Mahe (UTC +04:00)	0 (0%)
Sierra Leone - Africa/Freetown (UTC)	0 (0%)
Sint Maarten (Dutch part) - America/Lower_Princes (UTC -04:00)	0 (0%)
Solomon Islands - Pacific/Guadalcanal (UTC +11:00)	0 (0%)
Somalia - Africa/Mogadishu (UTC +03:00)	0 (0%)
South Georgia and the South Sandwich Islands - Atlantic/South_Georgia (UTC -02:00)	0 (0%)
South Sudan - Africa/Juba (UTC +03:00)	0 (0%)
Sri Lanka - Asia/Colombo (UTC +05:30)	0 (0%)
Suriname - America/Paramaribo (UTC -03:00)	0 (0%)
Svalbard and Jan Mayen - Arctic/Longyearbyen (UTC +01:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
Swaziland - Africa/Mbabane (UTC +02:00)	0 (0%)
Syrian Arab Republic - Asia/Damascus (UTC +03:00),Province of China - Asia/Taipei (UTC +08:00)	0 (0%)
Tajikistan - Asia/Dushanbe (UTC +05:00),United Republic of - Africa/Dar_es_Salaam (UTC +03:00)	0 (0%)
Thailand - Asia/Bangkok (UTC +07:00)	0 (0%)
Timor-Leste - Asia/Dili (UTC +09:00)	0 (0%)
Togo - Africa/Lome (UTC)	0 (0%)
Tokelau - Pacific/Fakaofo (UTC +13:00)	0 (0%)
Tonga - Pacific/Tongatapu (UTC +13:00)	0 (0%)
Trinidad and Tobago - America/Port_of_Spain (UTC -04:00)	0 (0%)
Tunisia - Africa/Tunis (UTC +01:00)	0 (0%)
Turkmenistan - Asia/Ashgabat (UTC +05:00)	0 (0%)
Turks and Caicos Islands - America/Grand_Turk (UTC -04:00)	0 (0%)
Tuvalu - Pacific/Funafuti (UTC +12:00)	0 (0%)
Uganda - Africa/Kampala (UTC +03:00)	0 (0%)
Ukraine - Europe/Simferopol (UTC +03:00)	0 (0%)
Ukraine - Europe/Uzhgorod (UTC +02:00)	0 (0%)
Ukraine - Europe/Zaporozhye (UTC +02:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
United Arab Emirates - Asia/Dubai (UTC +04:00)	0 (0%)
United States - America/Indiana/Knox (UTC -05:00)	0 (0%)
United States - America/Indiana/Marengo (UTC -04:00)	0 (0%)
United States - America/Indiana/Petersburg (UTC -04:00)	0 (0%)
United States - America/Indiana/Vevay (UTC -04:00)	0 (0%)
United States - America/Indiana/Vincennes (UTC -04:00)	0 (0%)
United States - America/Indiana/Winamac (UTC -04:00)	0 (0%)
United States - America/Juneau (UTC -08:00)	0 (0%)
United States - America/Kentucky/Monticello (UTC -04:00)	0 (0%)
United States - America/Menominee (UTC -05:00)	0 (0%)
United States - America/Metlakatla (UTC -08:00)	0 (0%)
United States - America/Nome (UTC -08:00)	0 (0%)
United States - America/North_Dakota/Beulah (UTC -05:00)	0 (0%)
United States - America/Sitka (UTC -08:00)	0 (0%)
United States - America/Yakutat (UTC -08:00)	0 (0%)

Table C5

*Geographical Location (continued)*

	<b>**N =</b>
	<b>690**</b>
United States Minor Outlying Islands - Pacific/Midway (UTC -11:00)	0 (0%)
United States Minor Outlying Islands - Pacific/Wake (UTC +12:00)	0 (0%)
Uruguay - America/Montevideo (UTC -03:00)	0 (0%)
Uzbekistan - Asia/Samarkand (UTC +05:00)	0 (0%)
Uzbekistan - Asia/Tashkent (UTC +05:00)	0 (0%)
Vanuatu - Pacific/Efate (UTC +11:00), Bolivarian Republic of - America/Caracas (UTC -04:00)	0 (0%)
Venezuela	0 (0%)
Virgin Islands, U.S. - America/St_Thomas (UTC -04:00)	0 (0%)
Virgin Islands	0 (0%)
Wallis and Futuna - Pacific/Wallis (UTC +12:00)	0 (0%)
Western Sahara - Africa/El_Aaiun (UTC +01:00)	0 (0%)
Yemen - Asia/Aden (UTC +03:00)	0 (0%)
Zambia - Africa/Lusaka (UTC +02:00)	0 (0%)
Zimbabwe - Africa/Harare (UTC +02:00)	0 (0%)

## Appendix D

**Disclaimer:** This is a non-public version of LEBA (dated January 3, 2022) 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 E  
LEBA Long Form (23 Items)

	Items	Never	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.					
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.					

	Items	Never	Rarely	Sometimes	Often	Always
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.					
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.					

	Items	Never	Rarely	Sometimes	Often	Always
16	I use a blue-filter app on my computer screen within 1 hour before attempting to fall asleep.					
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.					

### 839 Latent Structure, Reliability and Structural Validity

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

Factor names	Items	Reliability Coefficients: ordinal alpha
F1: Wearing blue light filters	1-3	.96
F2: Spending time outdoors	4-9 (Item 4 is reversed)	.83
F3: Using phone and smartwatch in bed	10-14	.70
F4: Using light before bedtime	15-18	.69
F5: Using light in the morning and during daytime	19-23	.52
McDonald's Omega coefficient for the total scale		.73 (Total scale)

841 LEBA -long form showed satisfactory structural validity (CFI =.97; TLI = .96; RMSEA = .05[.04-.06, 90% CI]; SRMR =  
842 .09).

843 How to cite:

Appendix F  
LEBA Short Form (18 Items)

	Short Form (18 Items)	Never	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.					
05	I spend between 30 minutes and 1 hour per day (in total) outside.					
06	I spend between 1 and 3 hours per day (in total) outside.					
07	I spend more than 3 hours per day (in total) outside.					

	Short Form (18 Items)	Never	Rarely	Sometimes	Often	Always
08	I spend as much time outside as possible.					
09	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.					
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 dim my mobile phone screen within 1 hour before attempting to fall asleep.					
14	I use a blue-filter app on my computer screen within 1 hour before attempting to fall asleep.					
15	I dim my computer screen within 1 hour before attempting to fall asleep.					

	Short Form (18 Items)	Never	Rarely	Sometimes	Often	Always
16	I use tunable lights to create a healthy light environment.					
17	I use LEDs to create a healthy light environment.					
18	I use an alarm with a dawn simulation light.					

844 Latent Structure, Reliability and Structural Validity

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

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

846 How to cite: