

¹ *Light Exposure Behavior Assessment (LEBA): Development of a novel instrument to capture light exposure-related behaviours*

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

37 The authors made the following contributions. Mushfiqul Anwar Siraji: Formal
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40 original draft, Writing – review & editing;; Juliëtte van Duijnhoven: Conceptualization,
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47 Manuel Spitschan: Conceptualization, Data curation, Investigation, Project
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49 editing.

50

Abstract

51 Light exposure is an important driver of health and well-being. Many aspects of light
52 exposure are modulated by our behaviour. How these light-related behaviours can be
53 shaped to optimise personal light exposure is currently unknown. Here, we present a
54 novel, self-reported and psychometrically validated instrument to capture light
55 exposure-related behaviour, the Light Exposure Behavior Assessment (LEBA).

56 An expert panel prepared the initial 48 item pool. Responses to these items were
57 then collected in an online survey producing responses from an international sample
58 (690 completed responses, 74 countries, 28 time zones). Exploratory factor analysis
59 (EFA) on an initial subset of our sample ($n=428$) rendered a five-factor solution with 25
60 items (Wearing blue light filters, spending time outdoors, using phone and smart-watch
61 in bed, using light before bedtime, using light in the morning and during daytime). In a
62 confirmatory factor analysis (CFA) performed on an independent subset of participants
63 ($n=262$), we removed two further items to attain the best fit for the five-factor solution
64 ($CFI=0.97$, $TLI=0.96$, $RMSEA=0.05$, $SRMR=0.09$). The internal consistency reliability
65 coefficient for the total instrument was McDonald's $\omega_{total}=0.73$. Measurement
66 model invariance analysis between native and non-native English speakers showed our
67 model attained the highest level of invariance (residual invariance; $CFI=0.95$, $TLI =0.95$,
68 $RMSEA=0.05$). Lastly, a short form of LEBA ($n=18$) was developed using Item Response
69 Theory on the complete sample ($n=690$).

70 The psychometric properties of the LEBA instrument indicate the usability to
71 measure the light exposure-related behaviours across a variety of settings and may offer
72 a scalable solution to characterise light exposure-related behaviours in remote samples.
73 The LEBA instrument will be available under the open-access CC-BY-NC-ND license.

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

76

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

87 Ethical approval

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

94 Data Availability

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

97 Survey characteristics

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

104 Completing the online survey took approx. 15 to 20 minutes and was not
105 compensated. The first page of the survey comprised a participant information sheet,
106 where participants' informed consent to participate was obtained before any of the
107 questions were displayed. Underaged participants (<18 years) were urged to obtain
108 assent from their parents/legal guardians, before filling in the survey. Information on the
109 first page included the objectives of the study, inclusion criteria, estimated duration, the
110 use, storage and sharing of the data, compensation (none), and information about the
111 type of questions in the survey. Moreover, participants needed to confirm that they were
112 participating the survey for the first time. To ensure high data quality, five attention check
113 items were included in the survey (e.g., "We want to make sure you are paying attention.
114 What is 4+5?"). The data analysed in this study was collected between 17.05.2021 and
115 03.09.2021. Questions incorporating retrospective recall were all aligned to the period of
116 "past four weeks," matching the presented LEBA instrument.

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

- 119 • Sleep disturbance and sleep-related impairment (adult and pediatric versions)
120 (Bevans et al., 2019; Daniel J. Buysse et al., 2010; Forrest et al., 2018; Harb,
121 Hidalgo, & Martau, 2015; L. Yu et al., 2011)

- 122 • Sleep duration, timing, and latency, chronotype, social jetlag, time in bed,
- 123 work/sleep schedule and outdoor light exposure duration (version for adults and
- 124 adolescents) (Roenneberg, Wirz-Justice, & Merrow, 2003)
- 125 • Sleep environment (Olivier et al., 2016)
- 126 • Meal timing & caffeine consumption [custom items]
- 127 • Light sensitivity (photophobia vs. photophilia) (Wu & Hallett, 2017)
- 128 • Self-reported pubertal stage (only if younger than 18 years old) (Petersen,
- 129 Crockett, Richards, & Boxer, 1988)

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

- 131 • Age
- 132 • Sex
- 133 • Gender identity
- 134 • Occupational Status
- 135 • COVID-19 related Occupational setting during the past four weeks
- 136 • Time zone & country of residence
- 137 • English as native language

138 Participants

139 Table 2 summarizes the survey participants' demographic characteristics. Only
140 participants completing the full LEBA questionnaire were included, thus there are no
141 missing values in the item analyses. XX participants were excluded from analysis due to
142 not passing at least one of the "attention check" items. For exploring initial factor
143 structure (EFA), a sample of 250-300 is recommended (Comrey & Lee, 1992;
144 Schönbrodt & Perugini, 2013). For estimating the sample size for the confirmatory factor
145 analysis (CFA) we followed the N:q rule (Bentler & Chou, 1987; Jackson, 2003; Kline,
146 2015; Worthington & Whittaker, 2006), where ten participants per parameter is required

147 to earn trustworthiness of the result. Our sample size exceeds these requirements:
148 Anonymous responses from a total of $n = 690$ participants were included in the analysis
149 of the current study, split into samples for Exploratory Factor Analysis (EFA: $n = 428$) and
150 Confirmatory Factor Analysis (CFA: $n = 262$). The EFA sample included participants
151 filling out the questionnaire from 17.05.2021 to XX.XX.XXXX, whereas participants who
152 filled out the questionnaire from YY.YY.YYYY to 03.09.2021 were included in the CFA
153 analysis. Participants indicated filling out the online survey from a diverse range of
154 geographic locations. The ten most common country + timezone combinations included:

- 155 • United States - America/New_York (UTC -04:00): 63 (9.1%)
- 156 • United Kingdom - Europe/London (UTC): 57 (8.3%)
- 157 • Germany - Europe/Berlin (UTC +01:00): 53 (7.7%)
- 158 • India - Asia/Kolkata (UTC +05:30): 38 (5.5%)
- 159 • United States - America/Los_Angeles (UTC -07:00): 37 (5.4%)
- 160 • United States - America/Chicago (UTC -05:00): 30 (4.3%)
- 161 • France - Europe/Paris (UTC +01:00): 22 (3.2%)
- 162 • Switzerland - Europe/Zurich (UTC +01:00): 21 (3.0%)
- 163 • Brazil - America/Sao_Paulo (UTC -03:00): 19 (2.8%)
- 164 • Netherlands - Europe/Amsterdam (UTC +01:00): 19 (2.8%)

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

166 Age among all participants ranged from 11 years to 84 years [EFA: $\min = 11$, $\max =$
167 84; CFA: $\min = 12$, $\max = 74$], with an overall mean of ~ 33 years of age [Overall: $M =$
168 32.95, $SD = 14.57$; EFA: $M = 32.99$, $SD = 15.11$; CFA: $M = 32.89$, $SD = 13.66$]. In total
169 325 (47%) of the participants indicated female sex [EFA: 189 (44%); CFA: 136 (52%)],
170 351 (51%) indicated male [EFA: 230 (54%); CFA: 121 (46%)] and 14 (2.0%) indicated
171 other sex [EFA: 9 (2.1%), CFA: 5 (1.9%)]. Overall, 49 (7.2%) [EFA: 33 (7.8%); CFA: 16
172 (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 χ^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 ??). 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

199 chronobiology, light research, neuroscience and psychology (including seven of the
200 authors, see authors roles) generated and collected preliminary item ideas. Special
201 attention was paid to design items circumscribed to assess light exposure *behavior* as
202 opposed to subjective measurements of the light environment (cf.(Eklund & Boyce,
203 1996) & (Dianat, Sedghi, Bagherzade, Jafarabadi, & Stedmon, 2013)) and
204 semi-quantitative assesments of light sources' illuminance (cf. (Bajaj, Rosner, Lockley, &
205 Schernhammer, 2011)) in order to maintain content validity. In a collective effort the
206 generated items were then peer-reviewed, amended, unified, and complemented with a
207 suitable response scale (5 point Likert-scale ranging from 1 "Never/Does not apply/
208 don't know" to 5 "Always"). This process was finalized when all experts were in
209 agreement, resulting in 48 items to implement in the data collection.

210 Analytic Strategies

211 Figure 1 summarizes the steps of our psychometric analysis. In our analysis we
212 used R (version 4.1.0), with several R packages. Initially, our tool had six point Likert
213 type response format (0:Does not apply/I don't know; 1:Never, 2:Rarely; 3:Sometimes;
214 4:Often; 5:Always). Our purpose was to capture light exposure related behavior and
215 these two response options: "Does not apply/I don't know" and "Never" were providing
216 similar information. As such we decided to collapse them into one, making it a 5 point
217 Likert type response format. Necessary assumptions of EFA, including sample
218 adequacy, normality assumptions, quality of correlation matrix, were assessed. Our data
219 violated both the univariate and multivariate normality assumptions. Due to these
220 violations and the ordinal nature of our response data, we used polychoric correlation
221 matrix (Desjardins & Bulut, 2018) for the EFA. We employed principal axis (PA) as factor
222 extraction method with varimax rotation. PA is robust to the normality assumption
223 violations (Watkins, 2020). The obtained latent structure was confirmed by another factor
224 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 A9) 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

250 chosen (Desjardins & Bulut, 2018). The corrected item-total correlation ranges between
251 .03 -.48. However, no item was discarded based on descriptive statistics or item analysis.

252 **Exploratory Factor Analysis**

253 Sampling adequacy was checked using Kaiser-Meyer-Olkin (KMO) measures of
254 sampling adequacy (Kaiser, 1974) . The overall KMO vale for 48 items was 0.63 which
255 was above the cutoff value (.50) indicating a mediocre sample (Hutcheson, 1999).
256 Bartlett's test of sphericity (Bartlett, 1954), χ^2 (1128) = 5042.86, p < .001 indicated the
257 correlations between items are adequate for the EFA. However only 4.96% of the
258 inter-item correlation coefficients were greater than .30. The absolute value of inter-item
259 correlation ranged between .00 to .91. Figure 2 depicts the correlation matrix.

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

265 With the initial 48 items we conducted three rounds of EFA and gradually discarded
266 problematic items. (cross-loading items and poor factor loading (<.30) items). Finally, a
267 five-factor EFA solution with 25 items was accepted with low RMSR = 0.08 (Brown,
268 2015), all factor-loading higher than .30 and no cross-loading greater than .30. We
269 further confirmed this five-factor latent structure by another EFA using varimax rotation
270 with a minimum residual extraction method (Sup.Table A7). Table 4 displays the
271 factor-loading (structural coefficients) and communality of the items. The absolute value
272 of the factor-loading ranged from .49 to .99 indicating strong coefficients. The
273 commonalities ranged between .11 to .99. Figure 4(A) depicts the obtained five factor
274 structure. However, the histogram of the absolute values of non-redundant

275 residual-correlations (Figure 4(B)) showed 26% correlations were greater than the
276 absolute value of .05, indicating a possible under-factoring. (Desjardins & Bulut, 2018).
277 Subsequently, we fitted a six-factor solution. However, a factor emerged with only two
278 salient variables, thus disqualifying the six-factor solution (Sup.Table A8). Internal
279 consistency reliability coefficient Cronbach's alpha assumes all the factor-loadings of the
280 items under a factor are equal (Graham, 2006; Novick & Lewis, 1967) which is not the
281 case in our sample. Additionally Cronbach's alpha coefficient has a tendency to deflate
282 the estimates for Likert type data as the calculation is based on pearson-correlation
283 matrix which requires that response data should be in continuous of nature (Gadermann,
284 Guhn, & Zumbo, 2012; Zumbo, Gadermann, & Zeisser, 2007). Subsequently to get
285 better estimates of reliability we reported ordinal alpha which used polychoric-correlation
286 matrix and assumed that the responses data were ordered in nature instead of
287 continuous (Zumbo et al., 2007). Ordinal alpha coefficient value ranges from 0 to 1 and
288 higher value represents better reliability. In the five-factor solution, the first factor
289 contained three items and explained 10.25% of the total variance with a internal reliability
290 coefficient ordinal $\alpha = .94$. All the items in this factor stemmed from the individual's
291 preference to use blue light filters in different light environments. The second factor
292 contained six items and explained 9.93% of the total variance with a internal reliability
293 coefficient ordinal $\alpha = .76$. Items under this factor commonly investigated an individual's
294 hours spent outdoor. The third factor contained five items and explained 8.83% of the
295 total variance. Items under this factor dealt with the specific behaviors pertaining to using
296 phone and smart-watch in bed. The internal consistency reliability coefficient was,
297 ordinal $\alpha = .75$. The fourth factor contained five items and explained 8.44% of the total
298 variance with an internal consistency coefficient, ordinal $\alpha = .72$. These five items
299 investigated the behaviors related to individual's light exposure before bedtime. Lastly,
300 the fifth factor contained six items and explained 6.14% of the total variance. This factor
301 captured individual's morning and daytime light exposure related behavior. The internal

302 consistency reliability was, ordinal $\alpha = .62$. It is essential to attain a balance between
303 psychometric properties and interpretability of the common themes when exploring the
304 latent structure. As all of the emerged factors are highly interpretable and relevant
305 towards our aim to capture light exposure related behavior, regardless of the apparent
306 low reliability of the fifth factor, we retain all the five-factors with 23 items for our
307 confirmatory factor analysis (CFA). Two items showed negative factor-loading (items 44
308 and 21). Upon inspection, it was understood that these items are negatively correlated to
309 the common theme, and thus in the CFA analysis, we reversed the response code for
310 these two items. Figure ?? depicts the data distribution and endorsement pattern for the
311 included items in our LEBA tool for both the EFA and CFA sample.

312 **Confirmatory Factor Analysis**

313 We conducted categorical confirmatory factor analysis with robust weighted least
314 square (WLSMV) estimator since our response data was of ordinary nature (Desjardins
315 & Bulut, 2018). Several indices are suggested to measure model fit which can be
316 categorized as absolute, comparative and parsimony fit indices (Brown, 2015). Absolute
317 fit assess the model fit at an absolute level using indices including χ^2 test statistics and
318 the standardized root mean square (SRMR). Parsimony fit indices including the root
319 mean square error of approximation (RMSEA) considers the number of free parameters
320 in the model to assesses the parsimony of the model. Comparative fit indices evaluate
321 the fit of the specified model solution in relation to a more restricted baseline model
322 restricting all covariances among the indicators as zero. Comparative fit index (CFI) and
323 the Tucker Lewis index (TLI) are such two comparative fit indices. Commonly used
324 Model fit guidelines (Hu & Bentle, 1999; Schumacker & Lomax, 2004) includes (i)
325 Reporting of χ^2 test statistics (A non-significant test statistics is required to reflect model
326 fit) (ii) CFI and TLI (CFI/TLI close to .95 or above/ranging between 90-95 and above) (iii)
327 RMSEA (close to .06 or below), (iv) SRMR (close to .08 or below) to estimate the model

328 fit. Table 5 summarizes the fit indices of our fitted model. Our fitted model failed to attain
329 an absolute fit estimated by the χ^2 test. However, the χ^2 test is sensitive to sample size
330 and not recommended to be used as the sole index of absolute model fit (Brown, 2015).
331 Another absolute fit index we obtained in our analysis was SRMR which does not work
332 well with categorical data (C. Yu, 2002). We judged the model fit based on the
333 comparative fit indices: CFI, TLI and parsimony fit index:RMSEA. Our fitted model
334 attained acceptable fit (CFI = .94; TLI = .93); RMSEA = .06,[.05-.07, 90% CI]) with two
335 imposed equity constrain on item pairs 32-33 [I dim my mobile phone screen within 1
336 hour before attempting to fall asleep.;I dim my computer screen within 1 hour before
337 attempting to fall asleep.] and 16-17 [I wear blue-filtering, orange-tinted, and/or
338 red-tinted glasses indoors during the day.;I wear blue-filtering, orange-tinted, and/or
339 red-tinted glasses outdoors during the day.]. Items pair 32-33 stemmed from the
340 preference of dimming electric device's brightness before bed time and items pair 16 and
341 19 stemmed from the preference of using blue filtering or colored glasses during the
342 daytime. Nevertheless, SRMR value was higher than the guideline (SRMR = .12).
343 Further by allowing one pair of items (30-41) [I look at my smartwatch within 1 hour
344 before attempting to fall asleep.;I look at my smartwatch when I wake up at night.] to
345 covary their error variance and discarding two item (item 37 & 26) for very low r-square
346 value, our model attained best fit (CFI = .95; TLI = .95); RMSEA = .06[.05-.06, 90% CI])
347 and SRMR value (SRMR = .11) was also close to the suggestions of Hu and Bentle
348 (1999). Internal consistency ordinal α for the five factors of LEBA were .96, .83, .70, .69,
349 .52 respectively. We also estimated the internal consistency reliability of the total scale
350 using McDonald's ω_t coefficient which is a better reliability estimate for multidimensional
351 constructs (Dunn, Baguley, & Brunsden, 2014; Sijtsma, 2009). McDonald's ω_t coefficient
352 for the total scale was .68. Figure 7 depicts the obtained CFA structure.

353 **Measurement Invariance**

354 Measurement invariance (MI) evaluates whether a construct has the psychometric
355 equivalence and same meaning across groups or measurement occasions (Kline, 2015;
356 Putnick & Bornstein, 2016). We used structural equation modeling framework to assess
357 the measurement invariance of our developed tool across two groups: native English
358 speakers(n= 129) and non-native English speakers (n = 133). For a detailed description
359 these two groups please see Sup. Table ???. Our measurement invariance testing
360 involved successively comparing the nested models: configural, metric, scalar, and
361 residual invariance models with each others (Widaman & Reise, 1997). Among these
362 nested models configural model is the first and least restrictive model. The configural
363 model assumes that the number of factors and item number under each factor will be
364 equal across two groups. The metric invariance model assumes configural invariance of
365 the fitted model and requires the factor-loadings of the items across the two groups to be
366 equal. Having the factor-loadings equal across groups indicates each item contributes to
367 the measured construct equivalently. Scalar invariance assumes the metric invariance of
368 the fitted model demands the item intercepts to be equivalent across groups. This equity
369 of item intercepts indicates the equivalence of response scale across the groups, i.e.,
370 persons with the same level of the underlying construct will score the same across the
371 groups. The residual invariance model assumes metric invariance for the fitted model
372 and adds the assumption of equality in error variances and covariances across the
373 groups. This model is the highest level of MI and assures the equivalence of precision of
374 items across the groups in measuring the underlying constructs. The invariance model fit
375 of our tool was assessed using the fit indices including χ^2 test, CFI and TLI (close to .95
376 or above), RMSEA (close to .06 or below) (Hu & Bentle, 1999). We excluded SRMR
377 from our consideration as it does not behave optimally for categorical variables (C. Yu,
378 2002). Table 6 summarized the fit indices. The comparison among different
379 measurement invariance models was made using the χ^2 difference test ($\Delta\chi^2$) to

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

390 **Semantic Analysis**

391 To find out if our developed tool (23 items) is overlapping with existing instruments,
392 we subjected the items of LEBA to the “Semantic Scale Network”(SSN) analysis
393 (Rosenbusch et al., 2020). The SSN detects semantically related scales and provides
394 cosine similarity index ranging between -.66 to 1 (Rosenbusch et al., 2020). Pair of
395 scales with a cosine similarity index value of 1 indicates they are perfectly semantically
396 similar scales indicating redundancy. LEBA appeared most strongly related to scales
397 about sleep: “Sleep Disturbance Scale For Children” (Bruni et al., 1996) and
398 “WHO-Composite International Diagnostic Interview (CIDI): Insomnia”(WHO, 1990).The
399 cosine similarities lie between .47 to .51. Flesch-Kincaid Grade Level (Flesch, 1948)
400 analysis on the the 23 items of our scale indicated required educational grade level was
401 3.33 and with a age above 8.33.

403 **Developing Short form of LEBA**

404 We sought the Item Response Theory (IRT) to develop the short form of LEBA. IRT
405 the conventional classical test theory-based analysis by gathering information on item
406 quality by indices like item difficulty, item discrimination, and item information (Baker,
407 2017). Item is judged based on item information in relation to participants' latent trait
408 level (θ). We fitted each factor of LEBA with the graded response model (Samejima,
409 Liden, & Hambleton, 1997) to the combined EFA and CFA sample (n =690). Item
410 discrimination indicates the pattern of variation in the categorical responses with the
411 changes in latent trait level (θ), and item information curve (IIC) indicates the amount of
412 information an item carries along the latent trait continuum. Here, we reported the item
413 discrimination parameter and only discarded the items with relatively flat item information
414 curve (information <.2) to develop the short form of LEBA. Baker (2017) categorized the
415 item discrimination in as none = 0; very low =0.01 to 0.34; low = 0.35 to 0.64; moderate =
416 0.65 to 1.34 ; high = 1.35 to 1.69; very high >1.70. Table 7 summarizes the IRT
417 parameters of our tool. Item discrimination parameters of our tool fell in very high (10
418 items), high (4 items), moderate (4 items), and low (5 items) categorizes indicating a
419 good range of discrimination along the latent trait level (θ). Examination of the item
420 information curve (Sup.fig A2-A5) indicated 5 items (1, 25, 38, 30, & 41) had relatively
421 flat information curves ($I(\theta) < .20$) thus discarded creating a short form of LEBA with 5
422 factors and 18 items.

423 Test information curve (TIC) (Figure 8) indicate the amount of information an the
424 full-scale carry along the latent trait continuum. As we treated each factor of short-LEBA
425 as an unidimensional construct we obtain 5 TICs (Figure 8). These information curves
426 indicated except the first and fifth factors, the other three factor's TICs are roughly
427 centered on the center of the trait continuum (θ).The first and fifth factor had a peak to
428 the right side of the center of latent trait.Thus we conferred the LEBA tool estimated the

429 light exposure related behavior with precision near the center of trait continuum for 2nd,
430 3rd and 4th factors and near the right side of the center of trait continuum for 1st and 5th
431 factors (Baker, 2017).

432 Table 8 summarizes the item fit indexes of the items. All the items fitted well to the
433 respective models as assessed by RMSEA value obtained from Signed- χ^2 index
434 implementation. All of the items had RMSEA value $\leq .06$ indicating adequate fit.
435 Sup. Figure A6 depicts the person fit of out fitted models. Person fit indicates the validity
436 and meaningfulness of the fitted model at the participants latent trait level (Desjardins &
437 Bulut, 2018). We estimated the person fit statistics using standardized fit index Zh
438 statistics (Drasgow, Levine, & Williams, 1985). Zh < -2 should be considered as a misfit.
439 Fig indicates that Zh is larger than -2 for most participants, suggesting a good fit of the
440 selected IRT models.

441 Discussion

442 We developed a self-reported tool to capture different light exposure related
443 behavior and evaluated its psychometric properties using classical test theory and Item
444 Response Theory based analysis.

445 48 items were generated by an expert panel and administered to a large sample (n
446 = 428 to explore the latent structure. Exploratory Factor Analysis revealed a five factor
447 solution with 25 items. ("Wearing blue light filters," "Spending time outdoors," "Using
448 phone and smart-watch in bed," "Using light before bedtime," and "Using light in the
449 morning and during daytime"). The internal consistency reliability coefficient ordinal
450 alpha ranged between .62-.94. As all the retained factors were meaningful and
451 contributed essentially towards our aim we retained all five factors.

452 A CFA on a separate sample (($n = 262$ gave a five-factor solution (CFI = .95; TLI =
453 .95); RMSEA = .06[.05-.06, 90% CI]) and SRMR = .11) after discarding two item. The

454 internal consistency McDonald's ω_t of the five factors were satisfactory (.96, .83, .70,
455 .69, .52) Internal consistency reliability of the total scale (23 items) was also satisfactory,
456 McDonald's $\omega_t = .68$. In the same sample, our measurement invariance analysis
457 revealed that the latent structure attained the residual measurement invariance across
458 subgroups: male and female (CFI: .98, TLI: .98, SRMR: .98).

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

468 Lastly, we developed a short-LEBA (n=23) using IRT analysis. We fitted a graded
469 response model model to the combined EFA and CFA sample (n =690). We discarded 5
470 items with relatively flat item information curve [$I(\theta) < .20$]. IRT analysis indicated short
471 form of LEBA is a psychometrically sound measure. Item fit indexes and person fit index
472 for all five fitted model were acceptable. Items had diverse slope parameters indicating a
473 good range of discrimination- the ability to differentiate respondents with different levels
474 of the light exposure related behavior. Test information curve also indicated a good
475 coverage of underlying trait continuum with precision.

476 Conclusion

477 "The Light exposure behavior assessment"(LEBA) gave a five solution with 25
478 items in an EFA. A CFA with this 25-item scale again offered a five-factor solution, but

479 this time two more item was discarded. The 23-item “LEBA” was found reliable and valid.

480 A short-form of LEBA was developed using IRT analysis. IRT analysis gave a 18-item

481 scale with a good range of coverage across the underlying trait continuum. All-in-all, we

482 can recommend both forms to be used to capture individual’s light exposure related

483 behavior

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Table 1

List of instruments measuring related constructs to LEBA

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 photosensitivity symptoms	None	5-point Likert scale	Not available
Office Light Survey	Eklundet al., 1996	30 items survey to assess electrical lighting environment in office	Item 29	Mixed response format	Not available
Harvard Exposure Assessment Questionnaire	Bajaj et al., 2011	1 item semi-quantitative questionnaire	None light	Semi-quantitative	Correlation with physical measurement
Hospital Lighting Survey	Dianat et el., 2013	23 items questionnaire to assess light environment in a hospital	Item 16,17	5-point Likert scale	Face and Content validity
Morningness-Eveningness Questionnaire	Horne et al., 1976	19 items questionnaire to understand your body clock	item 1,2,8,13,14	Mixed response format	Correlation with oral temperature

Table 1 continued

Name	Author	Description	Relevant Items	Scale type	Validity evidences
Munich Chrono-type Questionnaire (MCTQ)	Roenneberg et al., 2003	17 items questionnaire to understand individuals phase of entrainment	Time spent outdoors	Mixed response format	Correlation with sleep-logs, actimetry, and physiological parameters
Sleep Practices and Attitudes Questionnaire (SPAQ)	Olivier et.al., 2016	16 Factor questionnaire measuring practice, behavior and attitude related sleep	Subscale 8&9	5-point Likert scale	Face and Construct validity
The Pittsburgh Sleep Quality Index (PSQI)	Buysse et al., 1989	9 items inventory to measure sleep quality and sleeping pattern	item 1-4	Mixed response format	Correlation with clinical measurements
Self-Rating of Biological Rhythm Disorder for Adolescents (SBRDA)	Xie et al., 2021	29 Items questionnaire assessing four dimensions of biological rhythm disorder in adolescents	Item 3,6,22-25 and 29	5-point Likert scale	Construct validity
Photosensitivity Assessment Questionnaire (PAQ)	Bossini et al.,2006	16 dichotomous items questionnaire to assess "photophobia" and "photophilia"	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%)		

¹ Mean (SD); n (%)² False discovery rate correction for multiple testing³ Wilcoxon rank sum test⁴ Pearson's Chi-squared test

Table 3

Descriptive Statistics

Item	Stem	Mean	SD	Skew	Kurtosis	SW Statistics	Item-total corelation
item01	I turn on the lights immediately after waking up.	2.27	1.39	0.74	-0.81	0.81*	0.19
item02	I open the curtains or blinds immediately after waking up.	2.87	1.59	0.08	-1.60	0.83*	0.28
item03	I look at my mobile phone screen immediately after waking up.	3.36	1.38	-0.48	-1.03	0.87*	0.23
item04	I use an alarm with a dawn simulation light.	1.47	1.18	2.38	4.00	0.43*	0.24
item05	I have breakfast within 3 meters from a window.	4.01	1.40	-1.22	0.07	0.70*	0.17
item06	I have breakfast in a brightly lit room (illuminated by electric light).	2.79	1.55	0.19	-1.48	0.85*	0.13
item07	I go for a walk or exercise outside within 2 hours after waking up.	2.26	1.25	0.70	-0.60	0.85*	0.32
item08	I spend 30 minutes or less per day (in total) outside.	2.97	1.20	-0.06	-0.94	0.91*	0.25
item09	I spend between 30 minutes and 1 hour per day (in total) outside.	2.94	1.03	-0.12	-0.40	0.91*	0.08
item10	I spend between 1 and 3 hours per day (in total) outside.	2.74	1.04	0.09	-0.74	0.91*	0.42
item11	I spend more than 3 hours per day (in total) outside.	2.18	0.90	0.60	0.12	0.86*	0.41
item12	I spend as much time outside as possible.	2.36	1.22	0.59	-0.62	0.87*	0.48
item13	I use sunglasses when I go outside in bright daylight.	2.73	1.46	0.20	-1.36	0.87*	0.25
item14	I wear a visor or cap when I go outside in bright daylight.	2.14	1.31	0.77	-0.78	0.80*	0.28
item15	I seek shade when I am outside in bright daylight.	3.26	1.09	-0.26	-0.45	0.91*	0.03
item16	I wear blue-filtering, orange-tinted, and/or red-tinted glasses indoors during the day.	1.56	1.23	2.00	2.45	0.50*	0.28
item17	I wear blue-filtering, orange-tinted, and/or red-tinted glasses outdoors during the day.	1.54	1.21	2.07	2.75	0.49*	0.21
item18	I use light therapy applying a white light box.	1.12	0.49	5.02	27.80	0.25*	0.18

Table 3 continued

Item	Stem	Mean	SD	Skew	Kurtosis	SW Statistics	Item-total corelation
item19	I use light therapy applying a blue light box.	1.05	0.36	7.23	52.98	0.13*	0.17
item20	I use light therapy applying a light visor.	1.04	0.33	8.99	85.28	0.10*	0.16
item21	I use light therapy applying another form of light device.	1.14	0.59	4.79	24.05	0.25*	0.21
item22	I spend most of my daytime in a brightly lit environment.	3.57	1.07	-0.65	-0.17	0.88*	0.20
item23	I close the curtains or blinds during the day if the light from outside is bright.	2.56	1.27	0.33	-1.00	0.89*	0.08
item24	I spend most of my indoor time within 3 meters from a window.	4.14	0.99	-1.23	1.14	0.79*	0.22
item25	I use a desk lamp when I do focused work.	2.59	1.41	0.27	-1.27	0.86*	0.15
item26	I turn on my ceiling room light when it is light outside.	2.25	1.27	0.69	-0.64	0.84*	0.08
item27	I use my mobile phone within 1 hour before attempting to fall asleep.	3.80	1.29	-0.87	-0.42	0.82*	0.17
item28	I use my computer/laptop/tablet within 1 hour before attempting to fall asleep.	3.76	1.14	-0.68	-0.45	0.86*	0.18
item29	I watch television within 1 hour before attempting to fall asleep.	2.44	1.31	0.38	-1.14	0.86*	0.13
item30	I look at my smartwatch within 1 hour before attempting to fall asleep.	1.48	1.11	2.18	3.35	0.48*	0.13
item31	I dim my room light within 1 hour before attempting to fall asleep.	3.00	1.62	-0.08	-1.61	0.83*	0.39
item32	I dim my mobile phone screen within 1 hour before attempting to fall asleep.	3.55	1.65	-0.60	-1.34	0.76*	0.33

Table 3 continued

Item	Stem	Mean	SD	Skew	Kurtosis	SW Statistics	Item-total corelation
item33	I dim my computer screen within 1 hour before attempting to fall asleep.	3.62	1.64	-0.68	-1.25	0.74*	0.37
item34	I use a blue-filter app on my mobile phone screen within 1 hour before attempting to fall asleep.	3.42	1.83	-0.45	-1.69	0.69*	0.20
item35	I use a blue-filter app on my computer screen within 1 hour before attempting to fall asleep.	3.86	1.67	-0.99	-0.85	0.65*	0.20
item36	I wear blue-filtering, orange-tinted, and/or red-tinted glasses within 1 hour before attempting to fall asleep.	1.54	1.25	2.13	2.86	0.46*	0.35
item37	I purposely leave a light on in my sleep environment while sleeping.	1.33	0.91	3.03	8.43	0.41*	0.09
item38	I use as little light as possible when I get up during the night.	4.30	1.08	-1.79	2.53	0.67*	0.32
item39	I turn on the lights when I get up during the night.	1.96	0.98	1.02	0.69	0.82*	0.07
item40	I check my phone when I wake up at night.	2.16	1.19	0.71	-0.54	0.84*	0.25
item41	I look at my smartwatch when I wake up at night.	1.31	0.81	2.75	6.92	0.43*	0.14
item42	I close curtains or blinds to prevent light from entering the bedroom if I want to sleep.	3.93	1.48	-1.06	-0.44	0.71*	0.15
item43	I use a sleep mask that covers my eyes.	1.64	1.18	1.79	2.02	0.60*	0.22
item44	I modify my light environment to match my current needs.	3.51	1.30	-0.70	-0.59	0.85*	0.40
item45	I use LEDs to create a healthy light environment.	2.22	1.48	0.71	-1.02	0.76*	0.29
item46	I use tunable lights to create a healthy light environment.	1.76	1.23	1.35	0.44	0.66*	0.39

Table 3 continued

Item	Stem	Mean	SD	Skew	Kurtosis	SW Statistics	Item-total corelation
item47	I discuss the effects of light on my body with other people.	2.11	1.17	0.77	-0.39	0.83*	0.37
item48	I seek out knowledge on how to improve my light exposure.	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
item07		0.5				0.267	0.733
item08		-0.49				0.252	0.748
item09		0.32				0.113	0.887
item27			0.8			0.658	0.342
item03			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
item04					0.41	0.219	0.781
item01					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

Confirmatory Factor Analysis model fit indices of the two model: five factor model with 25 items and five factor model with 23 items. The second model attained the best fit.

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	415.45	231.00	.95	0.95	0.06	0.05	0.06	0.11

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

Measurement Invariance analysis on CFA sample (n=262) across native and non-native English speakers.

	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

*Items discrimination and response category
difficulty thresholds of 23 items in LEBA (n =690)*

	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
item07	1.09	-0.50	0.73	1.63	2.97
Ritem08	1.19	-2.26	-0.48	0.64	1.91
item09	0.91	-2.63	-0.96	1.11	3.49
item27	2.21	-1.88	-1.19	-0.73	0.30
item03	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
item04	0.84	2.44	2.80	3.18	3.67
item01	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 discrimination, response category difficulty thresholds and fit statistics of the 18 items in short LEBA (n=690)

Items	a	b1	b2	b3	b4	Signed Chi-square	df	RMSEA	p
item16	28.13	0.78	0.90	1.06	1.40	2.02	6.00	0.00	0.92
item36	4.49	0.94	1.08	1.23	1.40	39.07	13.00	0.05	0.00
item17	2.81	0.97	1.11	1.38	1.62	25.58	13.00	0.04	0.02
item11	3.27	-0.79	0.65	1.54	2.31	55.03	27.00	0.04	0.00
item10	3.07	-1.27	-0.09	0.82	2.00	53.19	30.00	0.03	0.01
item12	1.72	-0.67	0.44	1.28	2.11	34.39	42.00	0.00	0.79
item07	1.09	-0.50	0.73	1.63	2.97	67.45	46.00	0.03	0.02
Ritem08	1.19	-2.26	-0.48	0.64	1.91	140.90	46.00	0.05	0.00
item09	0.91	-2.63	-0.96	1.11	3.49	131.19	45.00	0.05	0.00
item27	2.12	-1.91	-1.21	-0.74	0.31	16.41	11.00	0.03	0.13
item03	3.24	-1.22	-0.76	-0.20	0.65	15.09	11.00	0.02	0.18
item40	1.57	-0.50	0.45	1.30	2.20	9.92	9.00	0.01	0.36
item32	1.60	-1.04	-0.79	-0.42	0.16	41.33	15.00	0.05	0.00
item35	1.34	-1.10	-0.99	-0.76	-0.41	41.71	14.00	0.05	0.00
item33	15.66	-0.66	-0.48	-0.24	0.13	46.89	14.00	0.06	0.00
item46	2.34	0.66	0.88	1.36	2.12	19.00	15.00	0.02	0.21
item45	1.51	0.30	0.55	1.17	1.91	15.05	15.00	0.00	0.45
item25	0.49	-1.45	-0.04	1.99	4.46	31.60	15.00	0.04	0.01

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

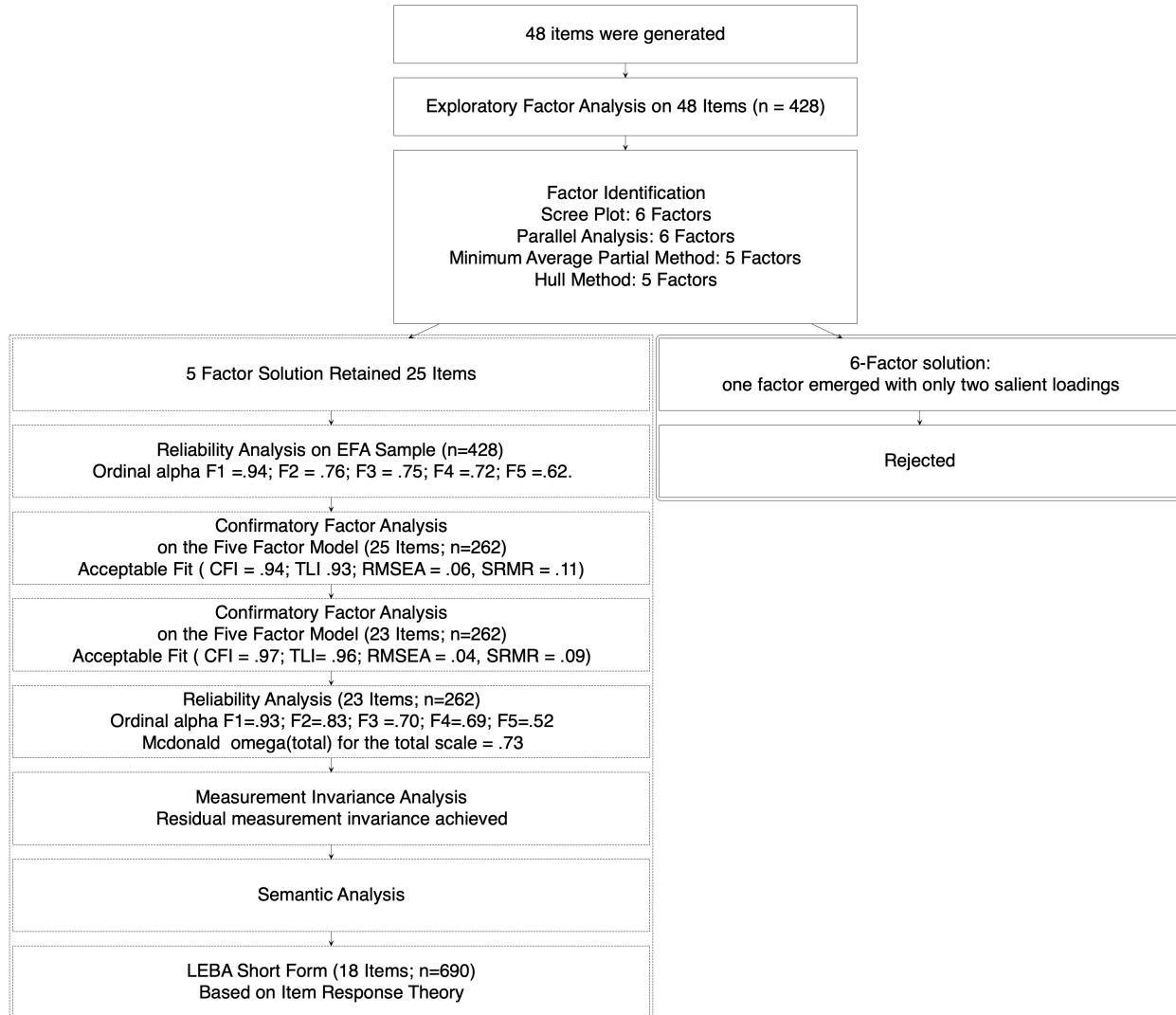


Figure 1. Development of long and short form of LEBA

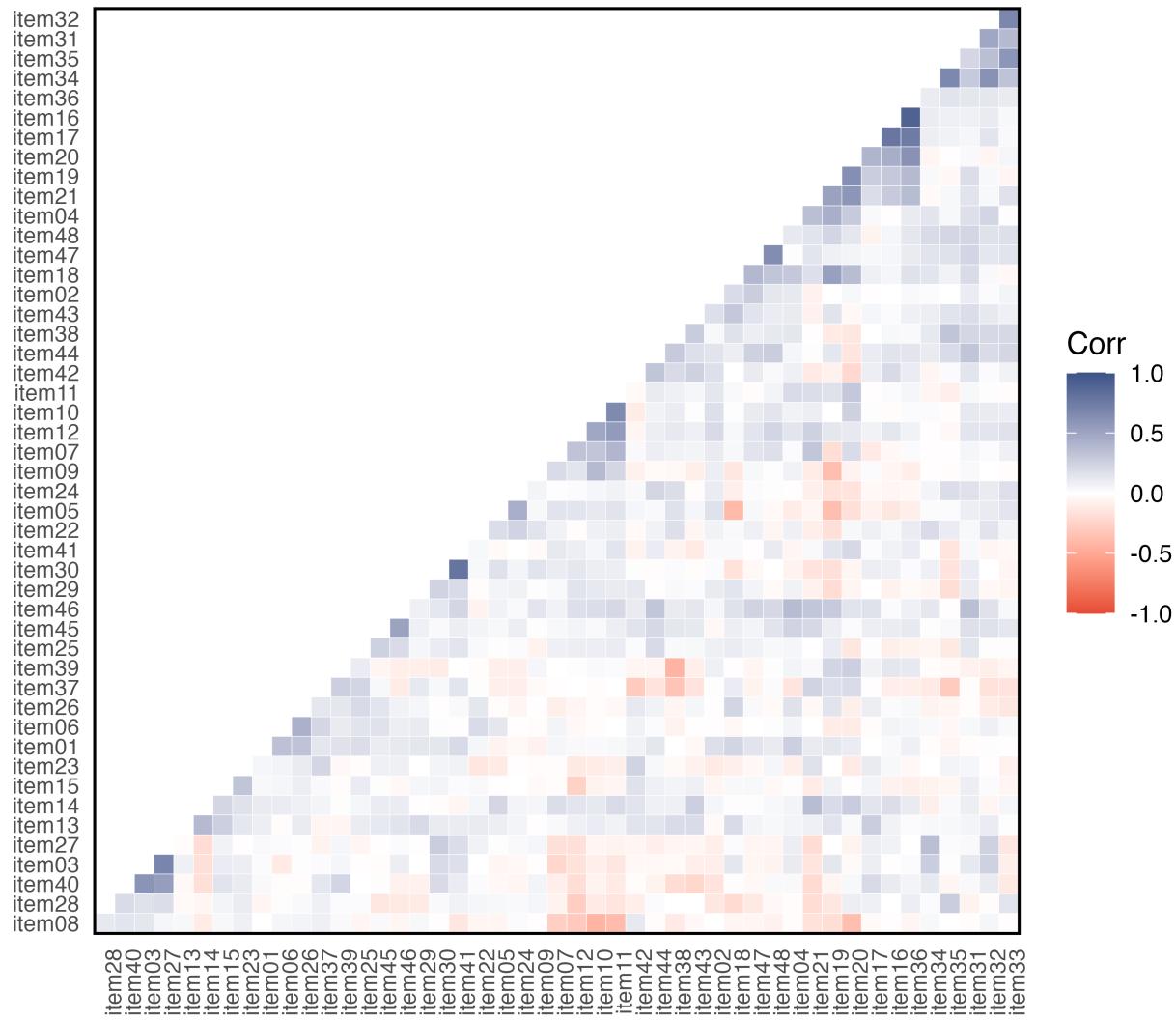


Figure 2. Inter item polychoric correlation coefficients for the 48 items. 4.9 % inter-item correlation coefficients were higher than .30

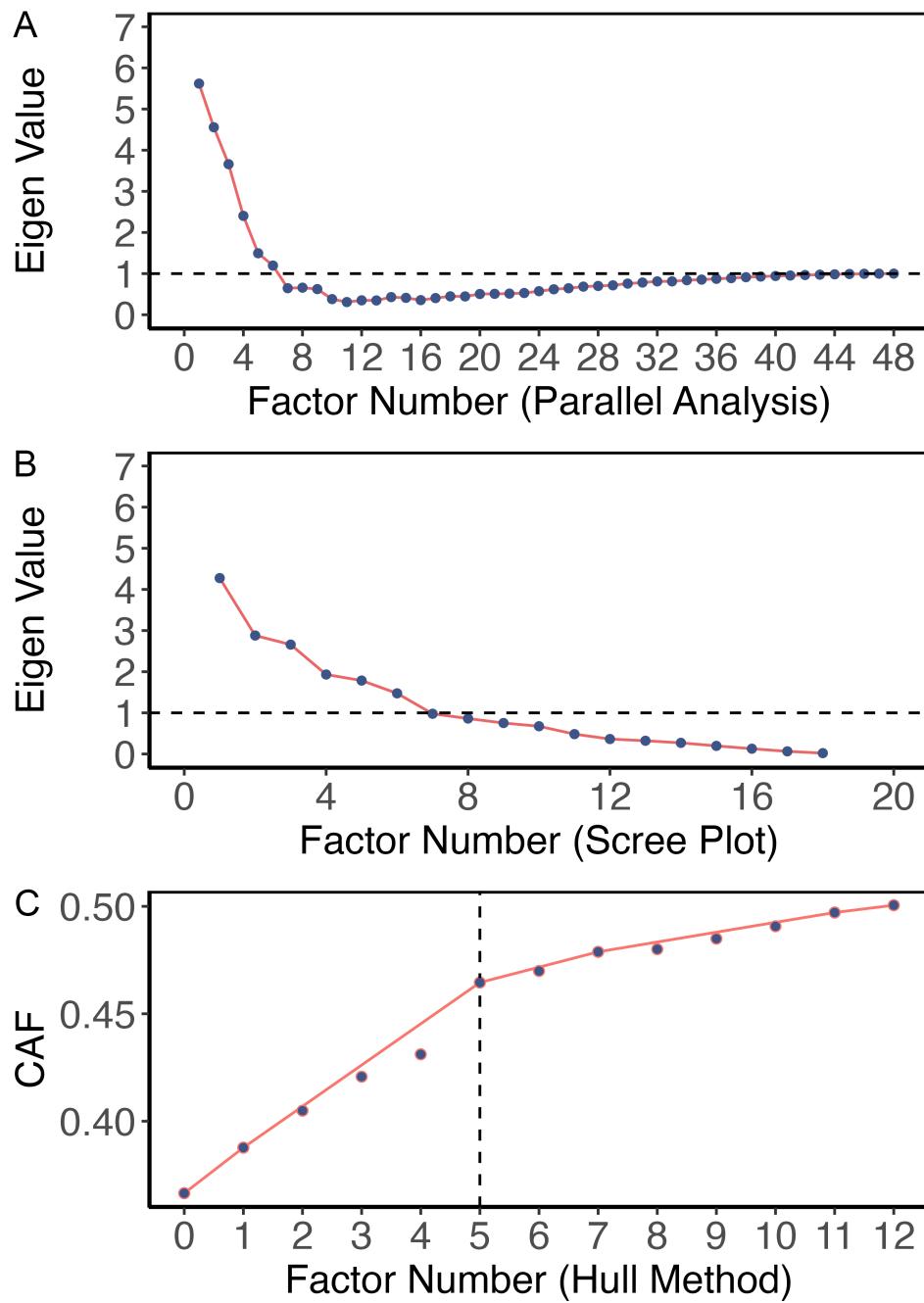


Figure 3. Factor Identification Methods (A) Parallel analysis indicated the optimal number of factors were six. (B) Scree plot suggested six factors. (C) Hull method indicated 5 factors were required to balance the model fit and number of parameters.

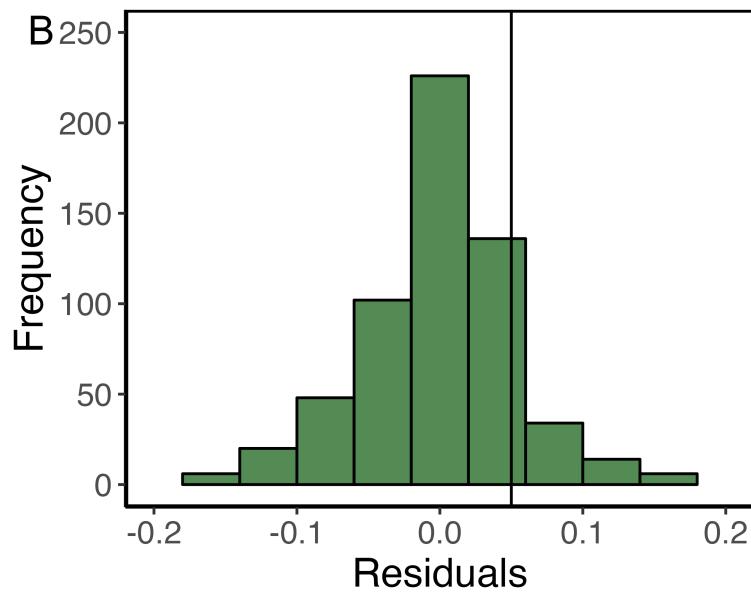
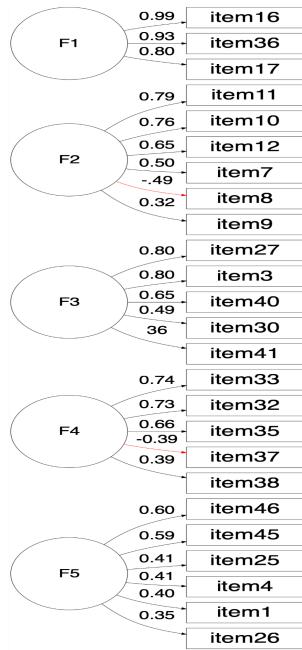
A

Figure 4. (A) Five Factor Solution obtained in Exploratory Factor Analysis (B) Histogram of nonredundant residula correlations

LEBA

Summary Descriptives EFA Sample (n =428)

Items	Summary Statistics				Histogram ¹	Density ²	Response Pattern				
	n	Mean	Median	SD			Never	Rarely	Sometimes	Often	Always
LEBA Items											
F1:Wearing blue light filters											
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)
item36	428	1.5	1.0	1.3			82.24% (352)	3.04% (13)	3.04% (13)	2.34% (10)	9.35% (40)
F2:Spending time outdoors											
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)
F3:Using phone and smart-watch in bed											
item03	428	3.4	4.0	1.4			15.89% (68)	11.45% (49)	17.29% (74)	31.07% (133)	24.30% (104)
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)
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)
F4:Using light before bedtime											
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)
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)
F5:Using light in the morning and during daytime											
item01	428	2.3	2.0	1.4			42.29% (181)	22.20% (95)	12.62% (54)	12.38% (53)	10.51% (45)
item04	428	1.5	1.0	1.2			84.11% (360)	3.50% (15)	2.10% (9)	2.10% (9)	8.18% (35)
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)
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)

¹Histogram²Density

Figure 5. Summary Descriptives EFA Sample

LEBA

Summary Descriptives CFA Sample (Nn=262)

Items	Summary Statistics				Graphics		Response Pattern					
	LEBA Items	n	Mean	Median	SD	Histogram ¹	Density ²	Never	Rarely	Sometimes	Often	Always
F1:Wearing blue light filters												
item16	262	1.6	1.0	1.3	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	1.2			80.15% (210)	3.44% (9)	5.34% (14)	2.67% (7)	8.40% (22)
item36	262	1.6	1.0	1.3	1.3			80.53% (211)	3.44% (9)	3.05% (8)	3.44% (9)	9.54% (25)
F2:Spending time outdoors												
item07	262	2.1	2.0	1.2	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	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	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	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	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	1.2			32.06% (84)	30.92% (81)	19.08% (50)	11.45% (30)	6.49% (17)
F3:Using phone and smart-watch in bed												
item03	262	3.7	4.0	1.3	1.3			11.83% (31)	7.25% (19)	17.56% (46)	28.24% (74)	35.11% (92)
item27	262	4.0	4.0	1.2	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	1.1			83.59% (219)	2.67% (7)	4.20% (11)	6.11% (16)	3.44% (9)
item40	262	2.5	2.0	1.3	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	0.7			90.08% (236)	3.82% (10)	2.29% (6)	2.67% (7)	1.15% (3)
F4:Using light before bedtime												
item32	262	3.4	4.0	1.7	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	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	1.8			27.48% (72)	2.67% (7)	7.25% (19)	6.49% (17)	56.11% (147)
item38	262	4.3	5.0	1.1	1.1			4.20% (11)	7.63% (20)	6.49% (17)	21.37% (56)	60.31% (158)
F5:Using light in the morning and during daytime												
item01	262	2.3	2.0	1.4	1.4			40.46% (106)	22.52% (59)	14.50% (38)	10.69% (28)	11.83% (31)
item04	262	1.3	1.0	0.8	0.8			89.31% (234)	2.29% (6)	3.44% (9)	3.05% (8)	1.91% (5)
item25	262	2.5	2.0	1.4	1.4			32.82% (86)	18.32% (48)	21.76% (57)	16.79% (44)	10.31% (27)
item45	262	2.0	1.0	1.4	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	1.2			75.57% (198)	2.67% (7)	8.02% (21)	9.54% (25)	4.20% (11)

¹ Histogram² Density

Figure 6. Summary Descriptives of CFA Sample

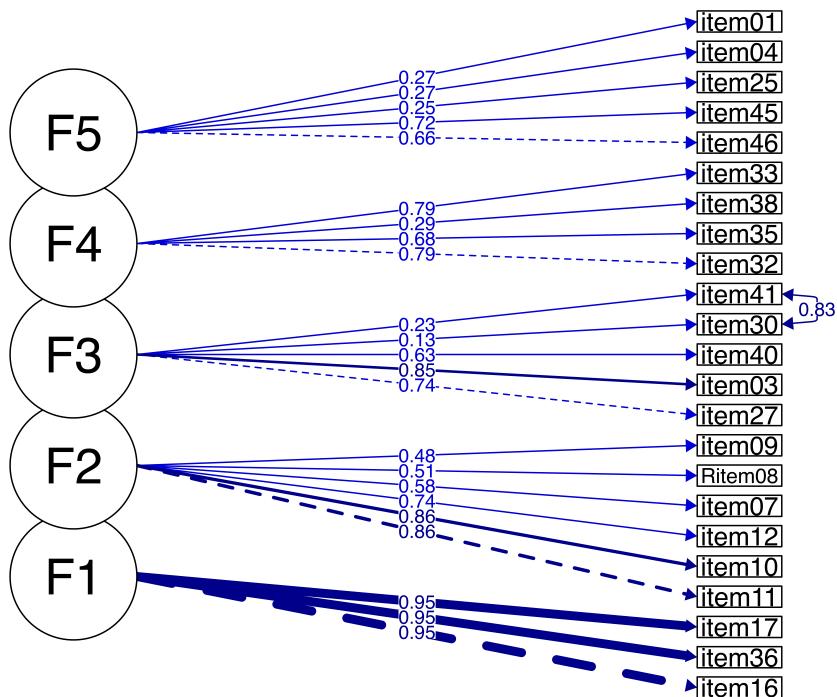


Figure 7. Five Factor Model of LEBA obtained by Confirmatory Factor Analysis. By allowing item pair 41 and 30 to covary their error variance our model attained the best fit.

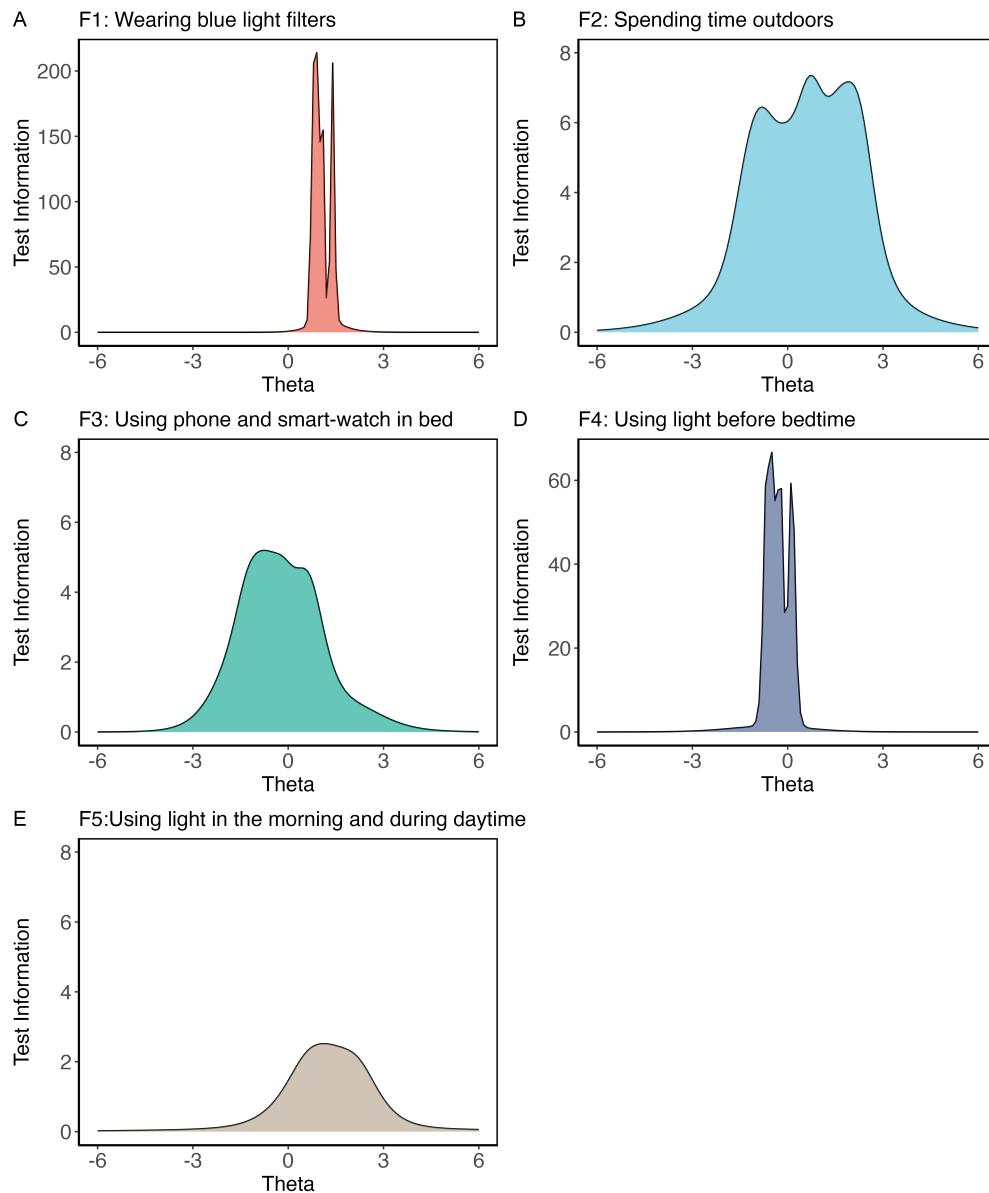


Figure 8. Test information curves (a) Wearing blue light filters (b) Spending time outdoors (c) Using phone and smartwatchin bed (d) Using light before bedtime (e) Using light in the morning andduring daytime

Appendix

839 **Disclaimer:** This is a non-public version of LEBA (dated January 11, 2022) and still a
840 work in progress. Please do not distribute!

841 LEBA captures light exposure-related behaviours on a 5 point Likert type scale
842 ranging from 1 to 5 (Never = 1; Rarely = 2; Sometimes = 3; Often = 4; Always = 5). The
843 score of each factor is calculated by the summation of scores of items belonging to the
844 corresponding factor.

845 **Instruction:**

846 "Please indicate how often you performed the following behaviours in the **past 4**
847 **weeks.**"

Table A1

LEBA Long Form (23 Items)

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 1 and 3 hours per day (in total) outside.					
06. I spend between 30 minutes and 1 hour per day (in total) outside.					
07.I spend more than 3 hours per day (in total) outside.					
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 look at my smartwatch within 1 hour before attempting to fall asleep					
14.I look at my smartwatch when I wake up at night.					
15.I dim my mobile phone screen within 1 hour before attempting to fall asleep.					
16.I use a blue-filter app on my computer screen within 1 hour before attempting to fall asleep.					
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.					

Table A1 continued

Items	Never	Rarely	Sometimes	Often	Always
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.					

Table A2

LEBA Long Form (23 Items):Latent Structure and Reliability

Factor names	Items	Reliability Coefficients
F1: Wearing blue light filters	01-03	0.96
F2: Spending time outdoors	4-9 (Item 4 is reversed)	0.83
F3: Using phone and smartwatch in bed	10-14	0.7
F4: Using light before bedtime	15-18	0.69
F5: Using light in the morning and during daytime	19-23	0.52
McDonald's Omega coefficient for the total scale		0.73

Table A3

LEBA Short Form (18 Items)

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 1 and 3 hours per day (in total) outside.					
06. I spend between 30 minutes and 1 hour per day (in total) outside.					
07.I spend more than 3 hours per day (in total) outside.					
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.					
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.					

Table A4

LEBA Short Form (18 Items): Latent Structure

Factor names	Items	Reliability Coefficients
F1: Wearing blue light filters	01-03	0.96
F2: Spending time outdoors	4-9 (Item 4 is reversed)	0.83
F3: Using phone and smartwatch in bed	10-14	0.7
F4: Using light before bedtime	15-18	0.69
F5: Using light in the morning and during daytime	19-23	0.52
McDonald's Omega coefficient for the total scale		0.73

Table A5

Minimum Average Partial (MAP) method of factor number determination. MAP Statistics is the lowest in the 5th row indicating five factors are required.

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

Table A6

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%)		

¹ Mean (SD); n (%)² False discovery rate correction for multiple testing³ Wilcoxon rank sum test⁴ Fisher's exact test⁵ Pearson's Chi-squared test

Table A7

Factor loadings and communality of the retained items (Minimum 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
item07		0.5				0.267	0.733
item08		-0.49				0.252	0.748
item09		0.32				0.113	0.887
item27			0.8			0.659	0.341
item03			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
item04					0.41	0.219	0.781
item01					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 A8

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
item08		-0.48					0.254	0.746
item07		0.47					0.257	0.743
item09		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
item03				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	
item04					0.48	0.332	0.668	
item25					0.4	0.238	0.762	
item01					0.35	0.134	0.866	
item26					0.35	0.161	0.839	
item37						-0.8	0.682	0.318
item38						0.39	0.245	0.755
% of Variance	0.11	0.1	0.09	0.09	0.06	0.05		

Table A8 continued

item	PA1	PA2	PA3	PA4	PA5	PA6	Communality	Uniqueness
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Note. Only loading higher than .30 is reported; Sixth factor has only two salient loadings

849

Table A9

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
item04	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
item08		-0.44				0.206	0.794
item07		0.42				0.226	0.774
item09		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
item03				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

Table A9 continued

item	PA1	PA2	PA5	PA3	PA4	Communality	Uniqueness
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

850

Table A10

*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
item04	0.49						0.296	0.704
item43	0.32					0.31	0.282	0.718
item10		0.81					0.67	0.33
item11		0.81					0.668	0.332
item12		0.58					0.408	0.592
item08		-0.45					0.218	0.782
item07		0.42					0.229	0.771
item09		0.33					0.115	0.885
item03			0.85				0.731	0.269
item27			0.77				0.606	0.394
item40			0.72				0.533	0.467

Table A10 continued

item	PA1	PA2	PA3	PA4	PA6	PA5	Communality	Uniqueness
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

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Table A11

Geographical Location of the participants (n =690)

Timezone	Number of Participants
Africa/Ceuta (UTC +01:00)	2.00
Africa/Douala (UTC +01:00)	1.00
Africa/Johannesburg (UTC +02:00)	5.00
Africa/Khartoum (UTC +02:00)	2.00
Africa/Lagos (UTC +01:00)	1.00
America/Adak (UTC -09:00)	2.00
America/Anchorage (UTC -08:00)	3.00

Table A11 continued

Timezone	Number of Participants
America/Araguaina (UTC -03:00)	2.00
America/Argentina/Buenos_Aires (UTC -03:00)	5.00
America/Argentina/Cordoba (UTC -03:00)	2.00
America/Argentina/Jujuy (UTC -03:00)	1.00
America/Bahia (UTC -03:00)	2.00
America/Blanc-Sablon (UTC -04:00)	1.00
America/Bogota (UTC -05:00)	2.00
America/Boise (UTC -06:00)	4.00
America/Cayman (UTC -05:00)	1.00
America/Chicago (UTC -05:00)	30.00
America/Costa_Rica (UTC -06:00)	2.00
America/Cuiaba (UTC -04:00)	1.00
America/Denver (UTC -06:00)	6.00
America/Detroit (UTC -04:00)	6.00
America/Edmonton (UTC -06:00)	14.00
America/Fortaleza (UTC -03:00)	1.00
America/Guatemala (UTC -06:00)	1.00
America/Guayaquil (UTC -05:00)	2.00
America/Halifax (UTC -03:00)	1.00
America/Indiana/Indianapolis (UTC -04:00)	3.00
America/Indiana/Tell_City (UTC -05:00)	1.00
America/Kentucky/Louisville (UTC -04:00)	3.00
America/Los_Angeles (UTC -07:00)	37.00
America/Martinique (UTC -04:00)	1.00
America/Mexico_City (UTC -06:00)	2.00
America/Moncton (UTC -03:00)	2.00
America/Monterrey (UTC -06:00)	1.00
America/New_York (UTC -04:00)	63.00
America/North_Dakota/Center (UTC -05:00)	1.00

Table A11 continued

Timezone	Number of Participants
America/North_Dakota/New_Salem (UTC -05:00)	1.00
America/Panama (UTC -05:00)	1.00
America/Phoenix (UTC -07:00)	7.00
America/Resolute (UTC -05:00)	1.00
America/Santiago (UTC -03:00)	8.00
America/Sao_Paulo (UTC -03:00)	19.00
America/Toronto (UTC -04:00)	16.00
America/Vancouver (UTC -07:00)	6.00
Antarctica/Macquarie (UTC +11:00)	1.00
Asia /Taipei City (UTC +08:00)	3.00
Asia/Amman (UTC +03:00)	2.00
Asia/Barnaul (UTC +07:00)	1.00
Asia/Dhaka (UTC +06:00)	1.00
Asia/Famagusta (UTC +02:00)	1.00
Asia/Ho_Chi_Minh (UTC +07:00),British - America/Tortola (UTC -04:00)	2.00
Asia/Hong_Kong (UTC +08:00)	2.00
Asia/Jakarta (UTC +07:00)	9.00
Asia/Jerusalem (UTC +02:00)	4.00
Asia/Karachi (UTC +05:00)	1.00
Asia/Kathmandu (UTC +05:45)	2.00
Asia/Kolkata (UTC +05:30)	38.00
Asia/Kuala_Lumpur (UTC +08:00)	7.00
Asia/Kuching (UTC +08:00)	2.00
Asia/Manila (UTC +08:00)	6.00
Asia/Novosibirsk (UTC +07:00)	1.00
Asia/Riyadh (UTC +03:00)	1.00
Asia/Seoul (UTC +09:00)	1.00
Asia/Shanghai (UTC +08:00)	7.00
Asia/Singapore (UTC +08:00)	1.00

Table A11 continued

Timezone	Number of Participants
Asia/Tokyo (UTC +09:00)	3.00
Asia/Tomsk (UTC +07:00)	1.00
Asia/Ulaanbaatar (UTC +08:00)	1.00
Asia/Vladivostok (UTC +10:00)	1.00
Asia/Yangon (UTC +06:30)	1.00
Asia/Yekaterinburg (UTC +05:00)	1.00
Atlantic/Canary (UTC)	1.00
Australia/Adelaide (UTC +10:30)	2.00
Australia/Brisbane (UTC +10:00)	4.00
Australia/Darwin (UTC +09:30)	1.00
Australia/Melbourne (UTC +11:00)	5.00
Australia/Perth (UTC +08:00)	2.00
Australia/Sydney (UTC +11:00)	2.00
East Africa/Dodoma (UTC +03:00)	1.00
Europe/Amsterdam (UTC +01:00)	19.00
Europe/Athens (UTC +02:00)	3.00
Europe/Belgrade (UTC +01:00)	3.00
Europe/Berlin (UTC +01:00)	53.00
Europe/Bratislava (UTC +01:00)	2.00
Europe/Brussels (UTC +01:00)	4.00
Europe/Bucharest (UTC +02:00)	3.00
Europe/Budapest (UTC +01:00)	2.00
Europe/Busingen (UTC +01:00)	3.00
Europe/Copenhagen (UTC +01:00)	3.00
Europe/Dublin (UTC)	5.00
Europe/Helsinki (UTC +02:00)	9.00
Europe/Istanbul (UTC +03:00)	6.00
Europe/Kiev (UTC +02:00)	1.00
Europe/Lisbon (UTC)	2.00

Table A11 continued

Timezone	Number of Participants
Europe/Ljubljana (UTC +01:00)	3.00
Europe/London (UTC)	57.00
Europe/Madrid (UTC +01:00)	7.00
Europe/Moscow (UTC +03:00)	8.00
Europe/Oslo (UTC +01:00)	3.00
Europe/Paris (UTC +01:00)	22.00
Europe/Prague (UTC +01:00)	3.00
Europe/Riga (UTC +02:00)	2.00
Europe/Rome (UTC +01:00)	9.00
Europe/Sofia (UTC +02:00)	1.00
Europe/Stockholm (UTC +01:00)	4.00
Europe/Tallinn (UTC +02:00)	2.00
Europe/Tirane (UTC +01:00)	1.00
Europe/Vienna (UTC +01:00)	1.00
Europe/Vilnius (UTC +02:00)	5.00
Europe/Warsaw (UTC +01:00)	15.00
Europe/Zagreb (UTC +01:00)	2.00
Europe/Zurich (UTC +01:00)	21.00
European /Skopje (UTC +01:00)	1.00
Iran /Tehran (UTC +0:30)	3.00
Pacific/Auckland (UTC +13:00)	6.00
Pacific/Chatham (UTC +13:45)	1.00
Pacific/Easter (UTC -05:00)	1.00
Pacific/Honolulu (UTC -10:00)	2.00

Table A12

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.
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.

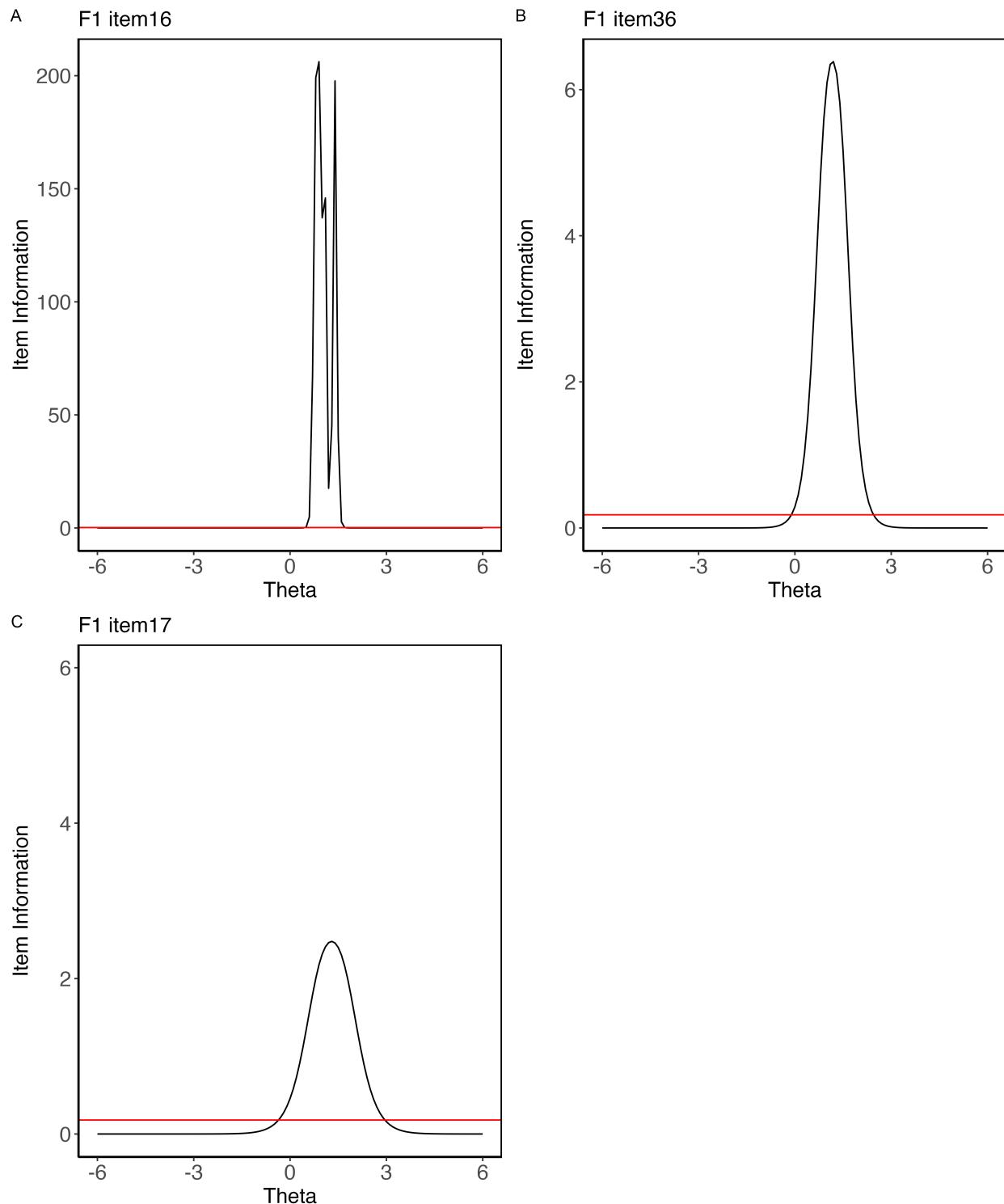


Figure A1. Item information curve of LEBA F1. Item 16 carried highest information among the three items

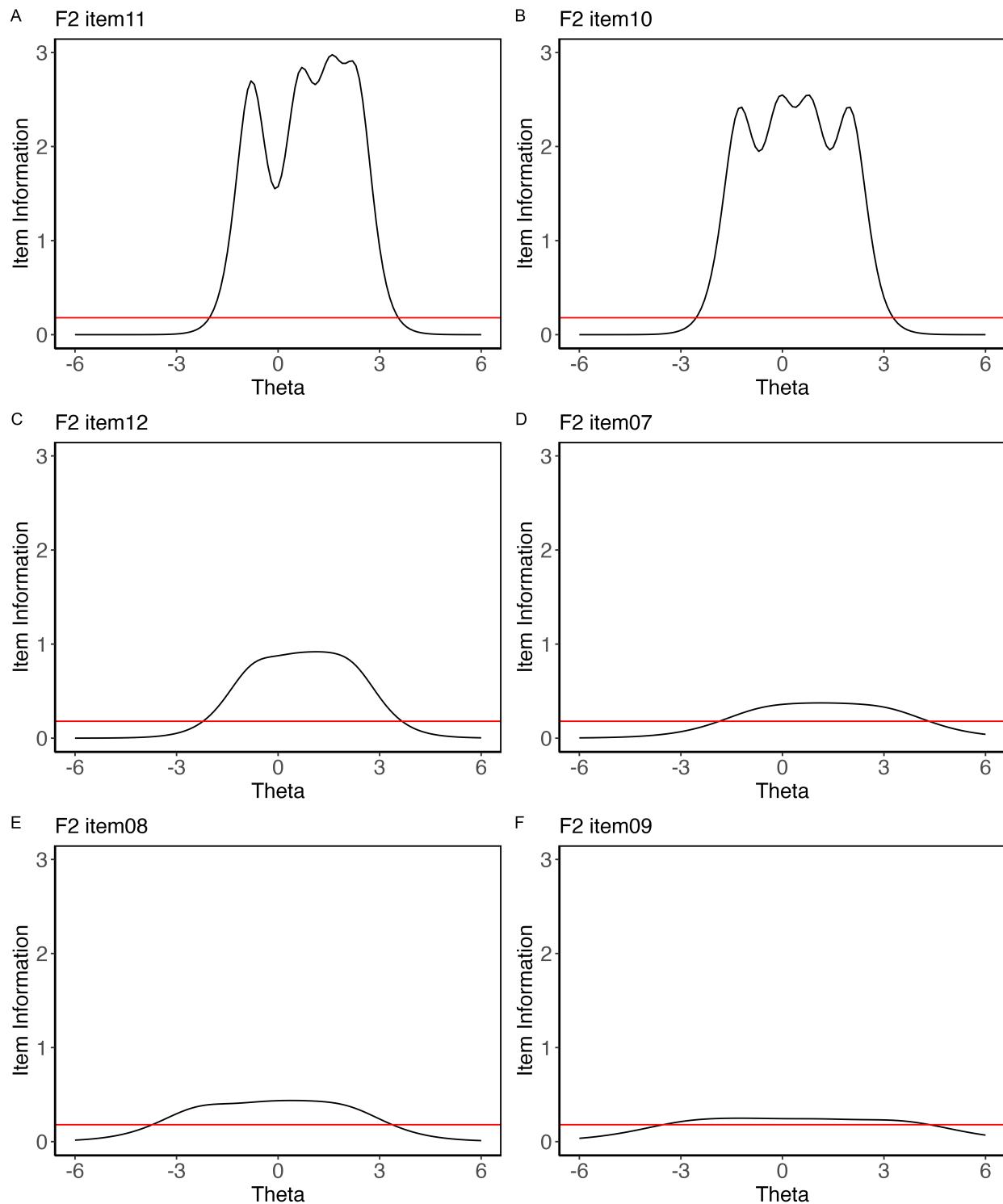


Figure A2. Item information curve of LEBA F2. All items carried information across the latent trait continuum. Item 09 carried the lowest information however its information peak was higher than .20.

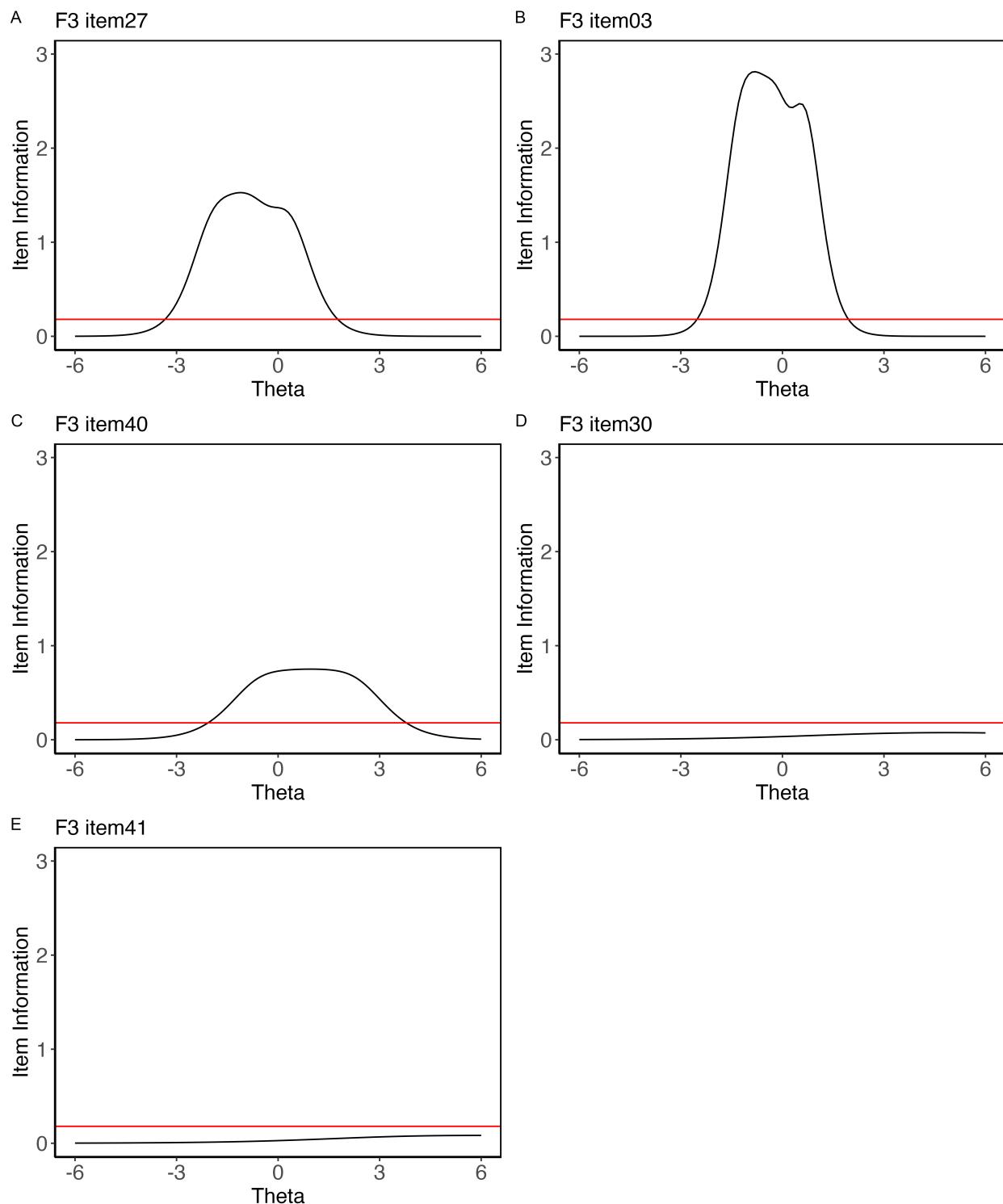


Figure A3. Item information curve of LEBA F3. Item 30 and 41 had relatively flat information curve with a peak lower than .20

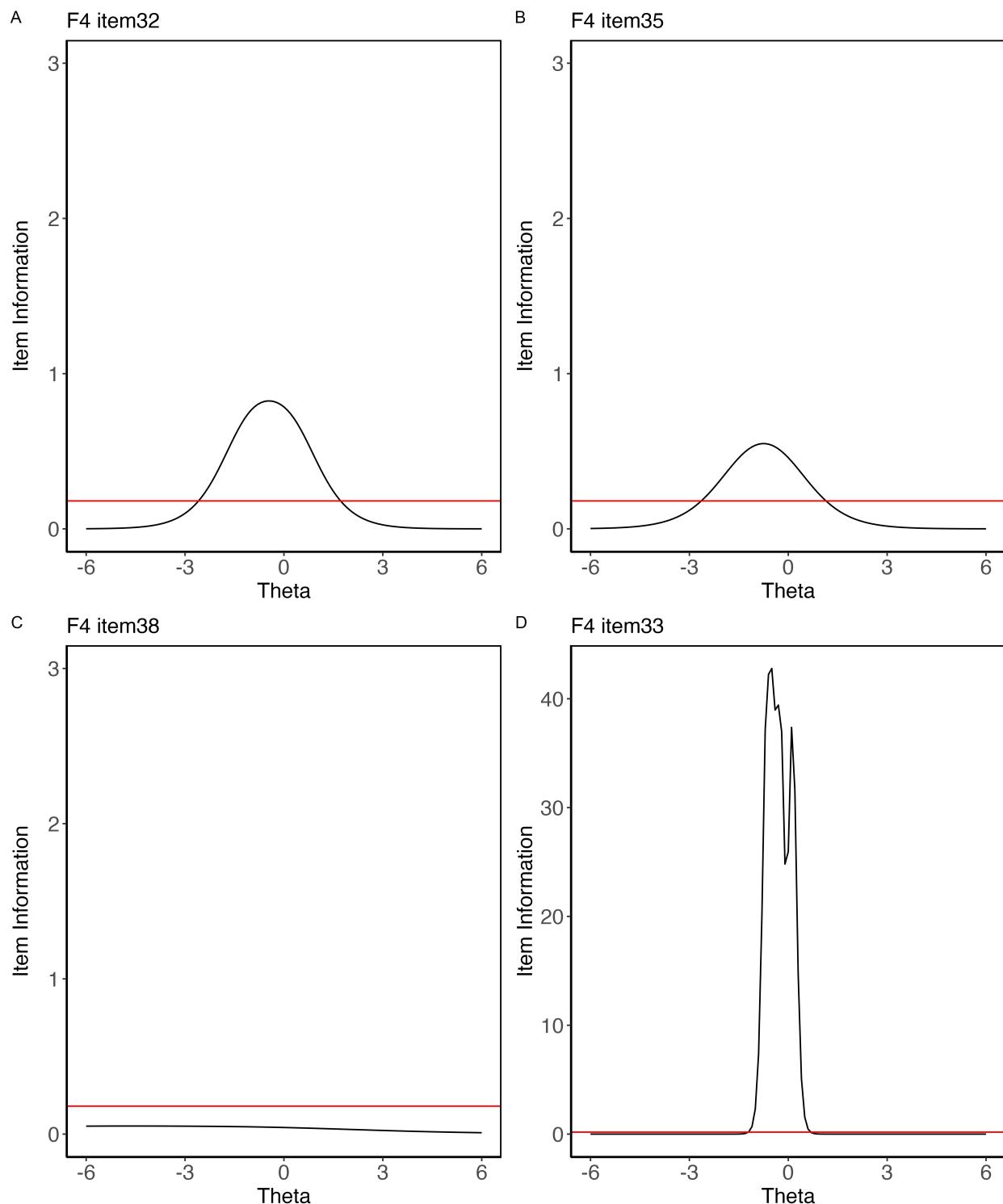


Figure A4. Item information curve of LEBA F4. Item 38 had relatively flat information curve with a peak lower than .20

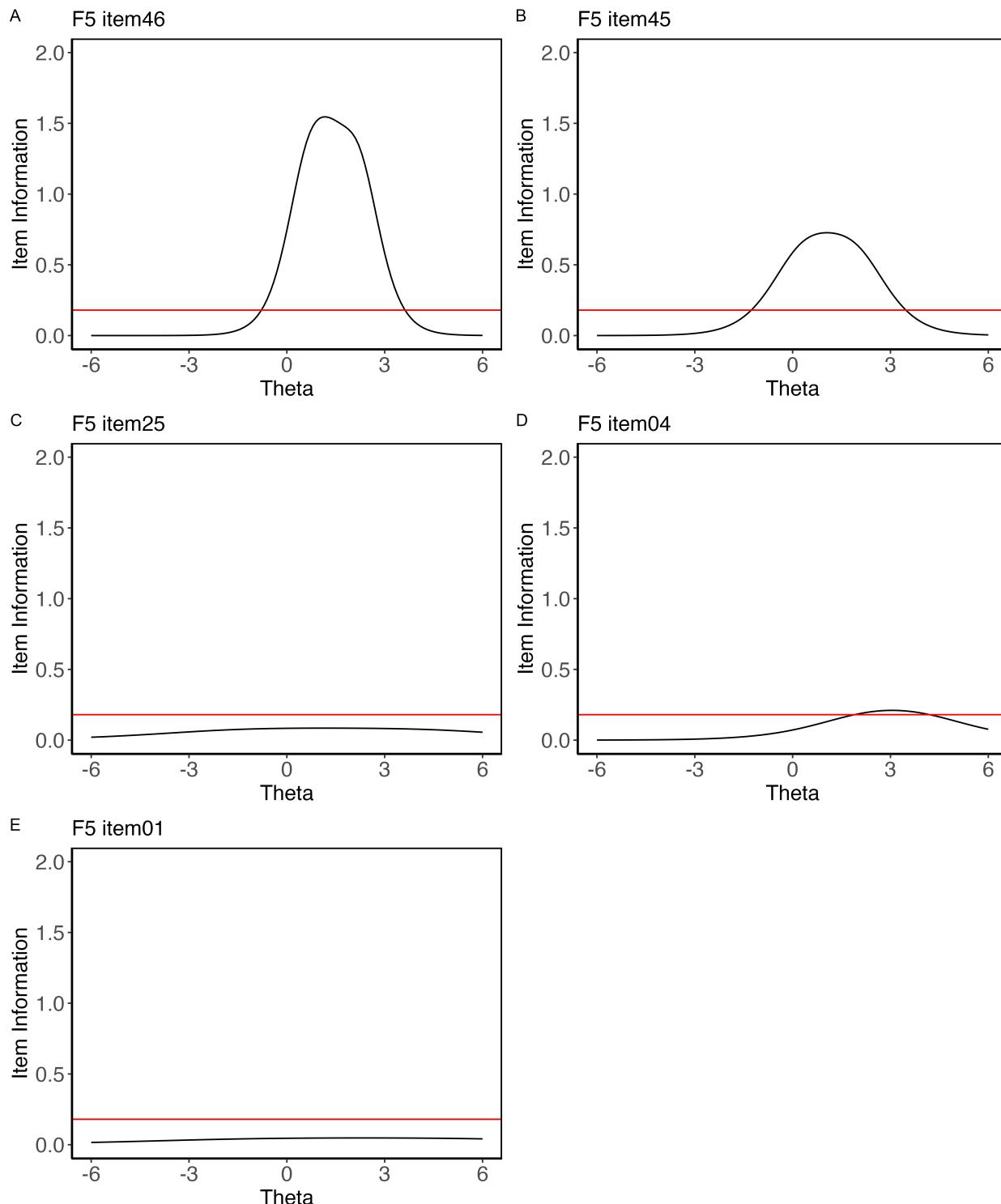


Figure A5. Item information curve of LEBA F5. Three items: 25, 04, 01 carried low information. However Item 04 had a bump at the right side of the center with a curve peak higher than .20. Item 01 and 25 had relatively flat information curve with a peak lower than .20

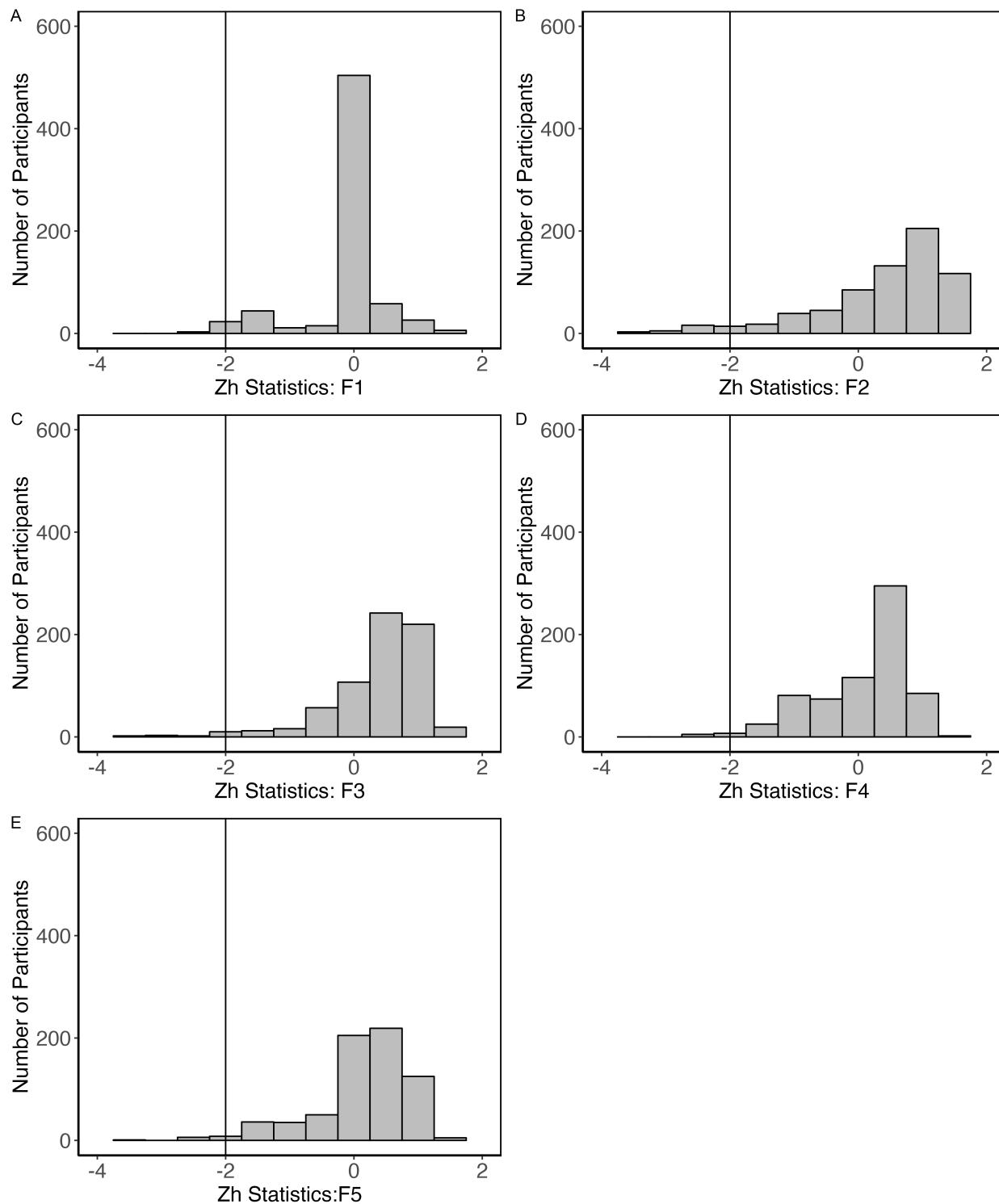


Figure A6. Person fit of the five fitted IRT models (a) Wearing blue light filters (b) Spending time outdoors (c) Using phone and smartwatchin bed (d) Using light before bedtime (e) Using light in the morning and during daytime