Running head: ONLINE 1

Light exposure related behavior and their influence on memory, concentration, chronotype, mood and sleep quality

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- ¹³ Conceptualization, Writing Original Draft Preparation, Writing Review & Editing;
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- ¹⁵ Conceptualization, Writing Review & Editing; Shamsul Haque: Conceptualization,
- 16 Writing Review & Editing.
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20 Abstract

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

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

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

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

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

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

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

35 Keywords: keywords

Word count: X

Light exposure related behavior and their influence on memory, concentration, chronotype, mood and sleep quality

##Objectives: 1. To explore the current light consumption pattern. 2. To investigate
whether light exposure related behaviours can predict Trouble in concentration, trouble
in memory Chronotype and sleep quality Mood.

42 Hypothesis

Light exposure related behaviour will successfully predict: Trouble in Concentration and memory Chronotype and Sleep quality Mood

45 Methods

A priori power analysis was done to determine adequate sample sizes with G*Power 3.0 (Faul, Erdfelder, Lang, & Buchner, 2007). To achieve a medium effect size (partial $\eta 2 = 0.14$)(Cohen, 1988) and 80% statistical power for liner multiple regression with seven predictors will need a total sample of 85 individuals. In our online survey we collected 124 participants data. 17 data were excluded due to incomplete data.

51 Participants

A online survey on 124 participants was conducted in September,2022 to understand the local people's light exposure related behavior. The survey was anonymous and a gift voucher of RM 10 was given to the fist 100 participants.

5 Material

Karolinska Sleepiness Scale. Subjective alertness wase evaluated with the Karolinska Sleepiness Scale (Åkerstedt & Gillberg, 1990). It is a self-reported subjective

assessment that uses a 10-point Likert-type scale, ranging from extremely sleepy (10 points) to extremely alert (1 point). This scale has been highly correlated with the actual workplace performance and the objective measures of fatigue.

PANAS. The positive and negative affect (PANAS) (Watson, Clark, & Tellegen, 1988) was used to measure positive and negative affects. PANAS is comprised of 20 items, 10 items measuring positive affects and 10 items measuring negative affects.

PANAS has a high internal consistency coefficient alpha (.87) and sufficient structural validity(Watson et al., 1988).

Office light Survey. We used office light survey (Eklund & Boyce, 1996) to 66 measure light satisfaction, trouble in concentration, trouble in memory and light's 67 influence on different physiological discomfort including fatigue, clumsiness, and 68 weakness. ### Harvard Light Exposure Assessment. To measure individual's average daily light exposure in lux from subjective reporting use was Harvard light exposure assessment (H-LEA) (Bajaj, Rosner, Lockley, & Schernhammer, 2011). H-LEA is a 71 reliable tool to assess daily light exposure as it showed significantly strong correlation (r 72 =.72, p<.001) with actual photopic light measures. However, one challenge we need to deal with was the absence of estimated corneal illuminance for LED luminaires. To address this issue, we take 10 random readings of illuminance at different position and averaged them.

Pittsburgh Sleep Quality Index. The Pittsburgh Sleep Quality Index (Buysse,
Reynolds C. F., Monk, Berman, & Kupfer, 1989) is a useful instrument used to measure
sleep quality and sleep patterns. It measures seven domains of sleep to differentiate
"poor" from "good" sleep. Scores range from zero to three on a Likert Scale, whereby 3
reflects the negative extreme on the Likert Scale. A sum of scores equal to or greater
than five indicates poor sleep quality. The PSQI has internal consistency and a reliability
coefficient (Cronbach's alpha) of 0.83 for its seven components. Scores also exhibit a
strong correlation with related sleep construct indicating high validity of the scale

Morningness-Eveningness Questionnaire. Morningness-Eveningness 85 questionnaire (Horne & Ostberg, 1976) is consist of 19 questions. The scores range 86 from 16 to 86. The higher scores indicate more morning propensity. Its internal 87 consistency in a New Zealand version was, Cronbach α coefficient=0.83. In the Original 88 study in student population (18-32 years) scores were validated with oral temperature 89 curves. ### Light exposure behavior assessment (LEBA). There was no suitable tool to 90 measure behaviors that may lead to different light exposure. As a result, we in 91 conjunction with a team based on University of Basel developed a tool to characterise different light exposure-related behaviours in people. A robust panel of expert created 48 93 pertaining items that captures different light exposure related behaviors. Our contribution 94 towards the development of this scale was to conduct formal psychometric analysis. Our exploratory factor analysis revealed five factor with 25 items. In our CFA analysis we obtain a best fit with a five factor model consisting 23 items (CFI = .97; TLI = .96; RMSEA = .05

$$.04 - .06, 90$$

, SRMR = .09) The internal consistency reliability coefficients were .93, .80, .61, .72, .45 respectively indicating satisfactory reliability for F1, F2 and F4. The manuscript of the development of LEBA is still a work in progress and will be submitted to a reputed Q1 journal.

Internal Consistency Indices for the scales

104 Name

103

105 Internal Consistency

Original Sample

107 Internal Consistency

Our Sample

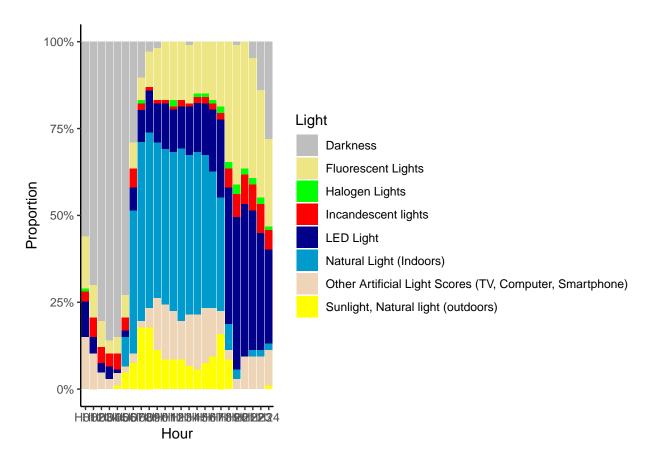
109	PANAS
110	.87
111	r apa(PANAS_IC.raw,2,T)
112	PSQI
113	.83
114	r apa(PSQI_IC.raw,2,T)
115	MEQ
116	.83
117	r apa(MEQ_IC.alpha.raw,2,T)
118	LEBA

: Internal Consistency Indices for the scales

120 Procedure

122

Data analysis

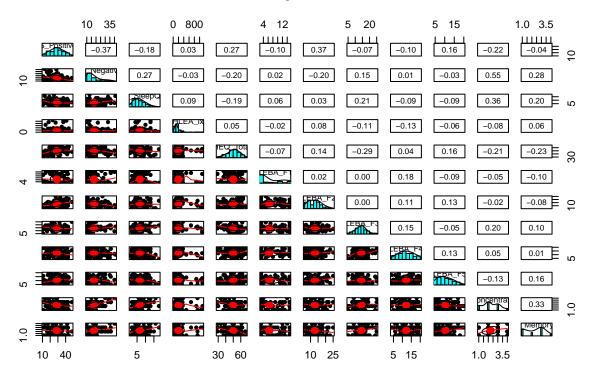


```
## corx::corx(data = LMR, caption = "Correlation Matrix")
123
   ##
124
   ## Correlation Matrix
125
126
   ##
                               PANAS PositiveAffect PANAS NegativeAffect
127
128
   ## PANAS_PositiveAffect
                                                                        -.37*
129
   ## PANAS_NegativeAffect
                                                -.37*
130
   ## PSQI_SleepQuality
                                                 -.18
                                                                         .27*
131
   ## HLEA_lx
                                                   .03
                                                                         -.03
132
```

133	##	MEQ_Total			27*		20*	
134	##	LEBA_F1		-	. 10		.02	
135	##	LEBA_F2		. 3	37*		20*	
136	##	LEBA_F3		-	. 07		.15	
137	##	LEBA_F4		-	. 10		.01	
138	##	LEBA_F5			. 16		03	
139	##	OLS_Concentration_rec		2	22*		.55*	
140	##	OLS_Memory_rec			.04		.28*	
141	##		PSQI_S1	eepQuality	HLEA_lx	$\texttt{MEQ_Total}$	LEBA_F1	LEBA_F2
142	##	PANAS_PositiveAffect		18	.03	.27*	10	.37*
143	##	PANAS_NegativeAffect		.27*	03	20*	.02	20*
144	##	PSQI_SleepQuality		-	.09	19	.06	.03
145	##	HLEA_1x		. 09	-	.05	02	.08
146	##	MEQ_Total		19	.05	-	07	.14
147	##	LEBA_F1		.06	02	07	-	.02
148	##	LEBA_F2		.03	.08	.14	.02	-
149	##	LEBA_F3		.21*	11	29*	.00	.00
150	##	LEBA_F4		09	13	.04	.18	.11
151	##	LEBA_F5		09	06	.16	09	.13
152	##	OLS_Concentration_rec		.36*	08	21*	05	02
153	##	OLS_Memory_rec		.20*	.06	23*	10	08
154	##		LEBA_F3	LEBA_F4 LE	EBA_F5 OI	LS_Concenti	ration_re	ec
155	##	PANAS_PositiveAffect	07	10	.16		22	2*
156	##	PANAS_NegativeAffect	. 15	.01	03		. 5	ō*
157	##	PSQI_SleepQuality	.21*	09	09		.36	ĵ*
158	##	HLEA_1x	11	13	06		(80
159	##	MEQ_Total	29*	.04	.16		2	1*

160	##	LEBA_F1	.00	.18	09	05	
161	##	LEBA_F2	.00	.11	.13	02	
162	##	LEBA_F3	_	.15	05	.20*	
163	##	LEBA_F4	.15	-	.13	.05	
164	##	LEBA_F5	05	.13	-	13	
165	##	OLS_Concentration_rec	.20*	.05	13	-	
166	##	OLS_Memory_rec	.10	.01	.16	.33*	
167	##		OLS_Memor	ry_rec			
168	##	PANAS_PositiveAffect		04			
169	##	PANAS_NegativeAffect		.28*			
170	##	PSQI_SleepQuality		.20*			
171	##	HLEA_1x		.06			
172	##	MEQ_Total		23*			
173	##	LEBA_F1		10			
174	##	LEBA_F2		08			
175	##	LEBA_F3		.10			
176	##	LEBA_F4		.01			
177	##	LEBA_F5		.16			
178	##	OLS_Concentration_rec		.33*			
179	##	OLS_Memory_rec		-			
180	##						
181	##	Note. * p < 0.05					

Scatterplot Matix



183

184

182

OLS_Concentration_rec

OLS_Concentration_rec

186 Coeffcient

187 Estimates

std. Beta

189 **p**

190 Estimates

std. Beta

192 **p**

(Intercept)

1.93(0.00)

```
0.00(-0.19 - 0.19)
```

- 196 <0.001
- 1.90(0.00)
- 198 0.00(-0.19 0.19)
- <0.001
- LEBA_F1
- -0.01(-0.06)
- -0.06(-0.26 0.13)
- 203 0.516
- LEBA_F2
- -0.00(-0.01)
- -0.01(-0.20 0.19)
- 207 0.954
- LEBA_F3
- 0.04(0.19)
- 0.19(-0.00 0.38)
- 211 0.055
- 212 0.04(0.20)
- 0.20(0.01 0.39)
- 214 0.041
- LEBA_F4
- 216 0.01(0.05)

```
0.05(-0.15 - 0.25)
```

218 0.647

219 LEBA_F5

-0.03(-0.13)

-0.13(-0.32 - 0.07)

222 0.196

-0.03(-0.12)

-0.12(-0.31 - 0.07)

225 0.220

Observations

227 107

228 107

R2 / R2 adjusted

0.060 / 0.014

0.055 / 0.037

232 AIC

287.253

281.824

log-Likelihood

-136.626

-136.912

```
OLS_Memory_rec
```

- OLS_Memory_rec
- Coeffcient
- Estimates
- std. Beta
- 244 **p**
- Estimates
- std. Beta
- 247 **p**
- (Intercept)
- 1.83(-0.00)
- -0.00(-0.19 0.19)
- 251 0.001
- 1.81(0.00)
- 253 0.00(-0.19 0.19)
- 254 <0.001
- LEBA_F1
- -0.02(-0.08)
- -0.08(-0.28 0.11)
- 258 0.395
- LEBA_F2
- -0.02(-0.10)

```
-0.10(-0.30 - 0.09)
```

262 0.290

LEBA_F3

0.03(0.11)

0.11(-0.09 - 0.30)

266 0.272

LEBA_F4

-0.00(-0.00)

-0.00(-0.20 - 0.20)

270 0.991

271 **LEBA_F5**

272 0.05(0.18)

0.18(-0.02 - 0.37)

274 0.079

275 0.04(0.16)

0.16(-0.03 - 0.35)

277 0.092

Observations

279 107

280 107

R2 / R2 adjusted

282 0.057 / 0.010

283	0.027 / 0.018
284	AIC
285	310.575
286	305.927
287	log-Likelihood
288	-148.287
289	-149.964
290	
291	PSQI_SleepQuality
292	PSQI_SleepQuality
293	Coeffcient
294	Estimates
295	std. Beta
296	р
297	Estimates
298	std. Beta
299	р
300	(Intercept)
301	4.85(0.00)
302	0.00(-0.19 – 0.19)
303	0.013

5.07(0.00)

```
0.00(-0.19 - 0.19)
```

- 306 0.001
- LEBA_F1
- 308 0.06(0.07)
- 309 0.07(-0.12 0.27)
- 310 0.449
- 311 LEBA_F2
- 312 0.04(0.05)
- 313 0.05(-0.14 0.24)
- 314 0.609
- 315 LEBA_F3
- 316 0.20(0.23)
- 317 0.23(0.04 0.42)
- 318 0.020
- 319 0.20(0.23)
- 0.23(0.04 0.42)
- 321 0.018
- LEBA_F4
- -0.11(-0.13)
- -0.13(-0.33 0.07)
- 325 0.185
- -0.10(-0.12)

```
-0.12(-0.31 - 0.07)
```

328 0.204

LEBA_F5

-0.05(-0.06)

-0.06(-0.25 – 0.14)

332 0.554

333 Observations

334 107

335 107

R2 / R2 adjusted

337 0.071 / 0.025

0.060 / 0.042

339 AIC

340 572.836

341 568.153

log-Likelihood

-279.418

-280.076

345

PANAS_PositiveAffect

PANAS_PositiveAffect

348 Coeffcient

```
Estimates
```

std. Beta

351 **p**

352 Estimates

std. Beta

354 **p**

(Intercept)

356 21.97(0.00)

357 0.00(-0.18 - 0.18)

<0.001

359 21.21(0.00)

360 0.00(-0.18 - 0.18)

361 <0.001

362 LEBA_F1

-0.13(-0.07)

-0.07(-0.26 - 0.11)

365 0.434

366 LEBA_F2

0.74(0.37)

0.37(0.19 - 0.55)

369 < 0.001

370 0.73(0.37)

```
0.37(0.19 - 0.55)
```

372 <0.001

373 LEBA_F3

-0.10(-0.05)

-0.05(-0.23 – 0.14)

376 0.621

-0.09(-0.04)

-0.04(-0.22 - 0.14)

379 0.642

380 LEBA_F4

-0.28(-0.14)

-0.14(-0.33 – 0.05)

383 0.144

-0.31(-0.15)

-0.15(-0.34 - 0.03)

386 0.100

387 LEBA_F5

388 0.27(0.12)

389 0.12(-0.06 - 0.30)

390 0.194

391 0.29(0.13)

0.13(-0.05-0.31)

393	0.162
394	Observations
395	107
396	107
397	R2 / R2 adjusted
398	0.182 / 0.142
399	0.177 / 0.145
400	AIC
401	752.005
402	750.656
403	log-Likelihood
404	-369.003
405	-369.328
406	
407	PANAS_NegativeAffect
408	PANAS_NegativeAffect
409	Coeffcient
410	Estimates
411	std. Beta
412	р
413	Estimates
414	std. Beta

```
415 p
```

(Intercept)

17.78(-0.00)

-0.00(-0.19 - 0.19)

<0.001

18.19(-0.00)

-0.00(-0.19 - 0.19)

<0.001

LEBA_F1

0.05(0.03)

0.03(-0.17 - 0.22)

426 0.789

LEBA_F2

-0.37(-0.20)

-0.20(-0.39 - -0.01)

430 0.043

-0.36(-0.20)

-0.20(-0.39 - -0.01)

433 0.039

LEBA_F3

0.30(0.16)

0.16(-0.04 - 0.35)

- 0.113
- 0.30(0.16)
- 0.16(-0.03 0.34)
- 440 0.105
- LEBA_F4
- -0.00(-0.00)
- -0.00(-0.20 0.20)
- 0.984
- LEBA_F5
- 0.02(0.01)
- 0.01(-0.19 0.20)
- 448 0.934
- Observations
- 450 107
- 451 107
- 452 **R2**
- 453 0.064
- 0.063 / 0.045
- 455 **AIC**
- 456 750.325
- 744.406
- log-Likelihood

```
-368.162
```

461

MEQ_Total

MEQ_Total

464 Coeffcient

465 Estimates

std. Beta

467 **p**

Estimates

std. Beta

470 **p**

(Intercept)

52.02(0.00)

0.00(-0.18 - 0.18)

<0.001

51.24(0.00)

0.00(-0.18 - 0.18)

<0.001

LEBA_F1

-0.14(-0.08)

-0.08(-0.26 - 0.11)

- 0.424
- LEBA_F2
- 483 0.23(0.12)
- 0.12(-0.07 0.30)
- 0.213
- 0.22(0.12)
- 0.12(-0.07 0.30)
- 488 0.218
- LEBA_F3
- -0.59(-0.29)
- -0.29(-0.48 -0.11)
- 492 0.002
- -0.59(-0.29)
- -0.29(-0.48 -0.10)
- 495 0.003
- LEBA_F4
- 0.14(0.07)
- 0.07(-0.12 0.27)
- 0.451
- 500 0.11(0.06)
- 0.06(-0.13 0.25)
- 502 0.541

```
LEBA_F5
```

- 0.26(0.12)
- o.12(-0.07 0.31)
- 506 0.212
- 0.28(0.13)
- 0.13(-0.06 0.32)
- 509 0.177
- 510 Observations
- 511 107
- 512 107
- R2 / R2 adjusted
- 0.128 / 0.085
- 0.123 / 0.088
- 516 AIC
- ⁵¹⁷ **752.026**
- 518 750.705
- log-Likelihood
- **-369.013**
- **-369.353**
- Durbin-Watson test is a test for a particular type of (lack of) independence; namely, 1st-order autocorrelation, which means that adjacent observations (specifically, their errors) are correlated (i.e., not independent) (Draper & Smith, 1998). D-W statistics less

than 1 or greater than 3 should definitely raise alarm bells. The closer to 2 that the value is, the better.

If the largest VIF is greater than 10 then there is cause for concern (Bowerman &O'Connell, 1990; Myers, 1990).

If the average VIF is substantially greater than 1 then the regression may be biased (Bowerman & O'Connell, 1990).

Tolerance below 0.1 indicates a serious problem. Tolerance below 0.2 indicates a potential problem (Menard, 1995).

```
lag Autocorrelation D-W Statistic p-value
533
                 0.05772839
                                   1.882874
   ##
                                               0.552
534
       Alternative hypothesis: rho != 0
   ##
535
       lag Autocorrelation D-W Statistic p-value
   ##
536
          1
                0.007636486
   ##
                                   1.982434
                                                0.93
537
   ##
       Alternative hypothesis: rho != 0
538
       lag Autocorrelation D-W Statistic p-value
   ##
539
          1
                 0.04662594
                                    1.89612
                                               0.528
   ##
540
       Alternative hypothesis: rho != 0
   ##
   ##
       lag Autocorrelation D-W Statistic p-value
               -0.005715252
   ##
                                   2.006302
                                               0.958
   ##
       Alternative hypothesis: rho != 0
       lag Autocorrelation D-W Statistic p-value
   ##
545
                                   2.026411
   ##
          1
                 -0.0192388
                                               0.904
546
```

Alternative hypothesis: rho != 0

```
^{548} ## lag Autocorrelation D-W Statistic p-value
```

549 **##** 1 0.05187488 1.858637 0.458

550 ## Alternative hypothesis: rho != 0

551 ## LEBA_F3 LEBA_F5

552 ## 1.002201 1.002201

553 ## [1] 1.002201

554 ## LEBA_F3 LEBA_F4

555 ## 1.023181 1.023181

556 ## [1] 1.023181

557 ## LEBA_F2 LEBA_F3 LEBA_F4 LEBA_F5

558 ## 1.026575 1.027925 1.052119 1.036737

559 ## [1] 1.035839

560 ## LEBA_F2 LEBA_F3 LEBA_F4 LEBA_F5

561 ## 1.026575 1.027925 1.052119 1.036737

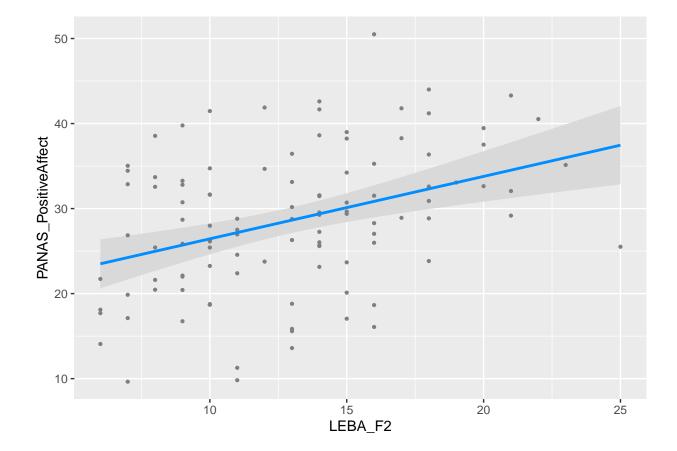
562 ## [1] 1.035839

563 ## LEBA_F2 LEBA_F3

564 **##** 1.000015 1.000015

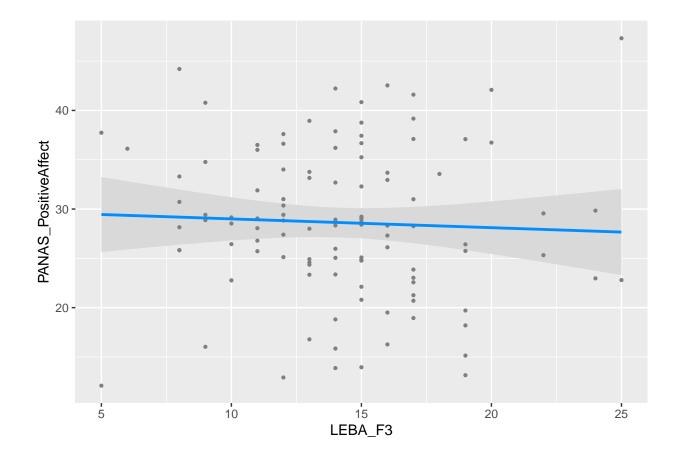
565 ## [1] 1.000015

566 ## [[1]]



567

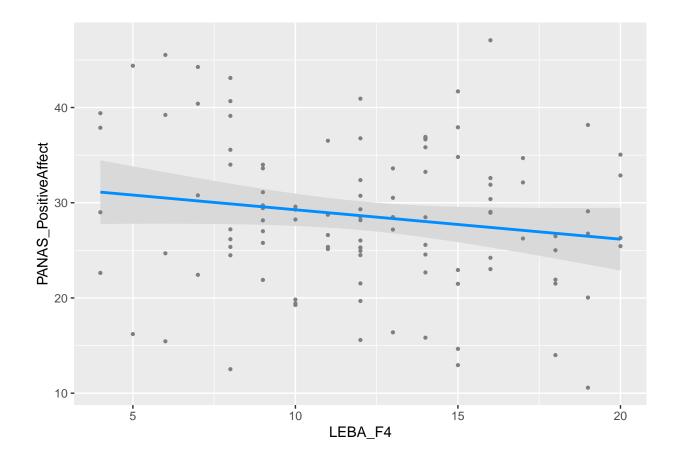
568 ## 569 ## [[2]]



570

571 ##

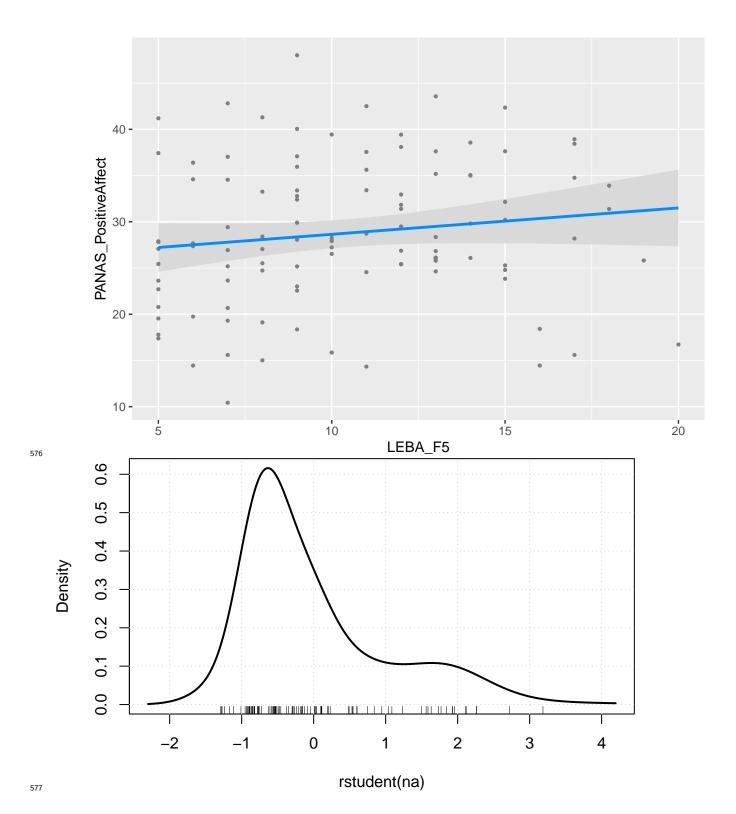
572 ## [[3]]

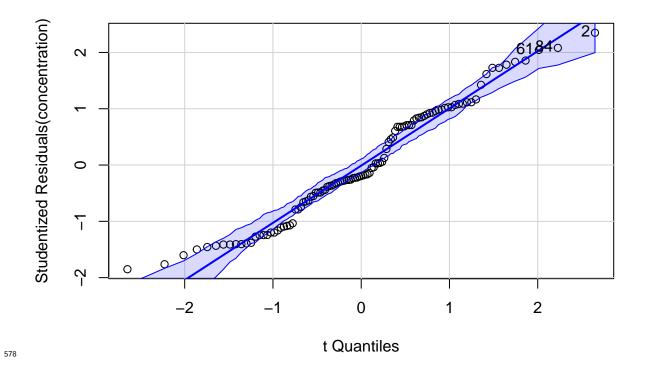


574 ##

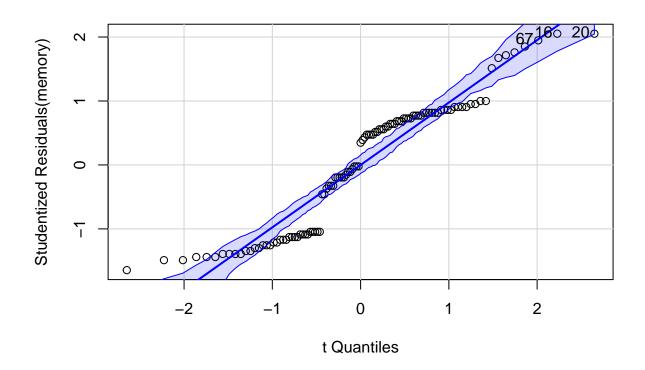
573

575 ## [[4]]

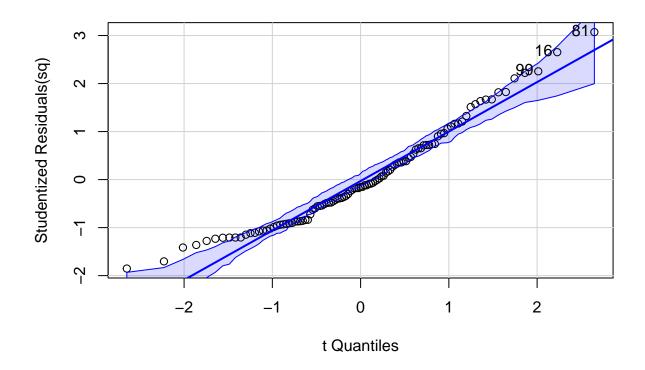




579 ## [1] 2 61 84

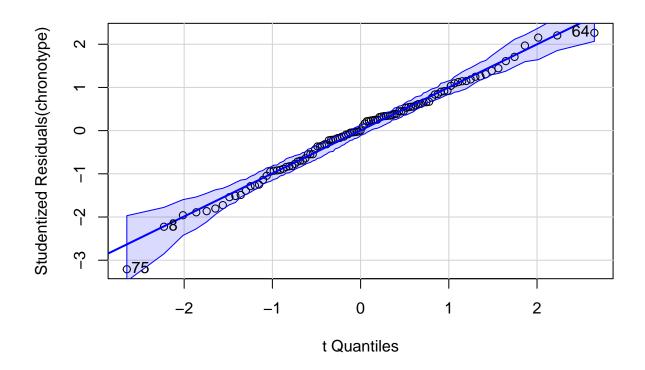


581 ## [1] 16 20 67

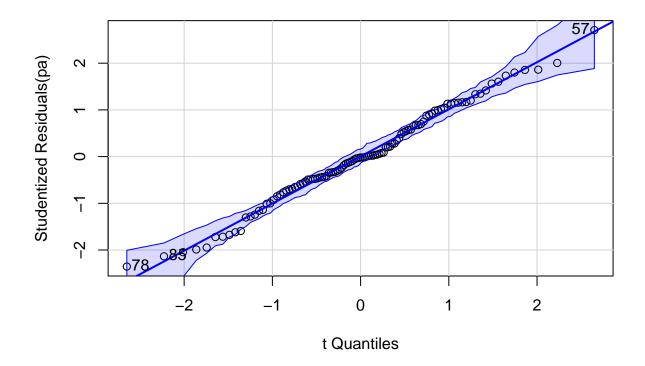


₅₈₃ ## [1] 16 81 99

582

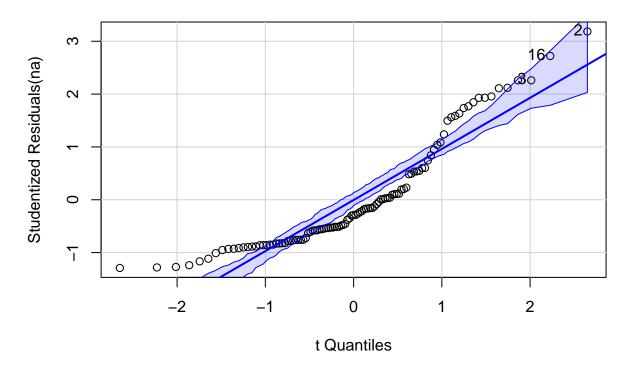


585 ## [1] 8 64 75



587 **##** [1] 57 78 85

586



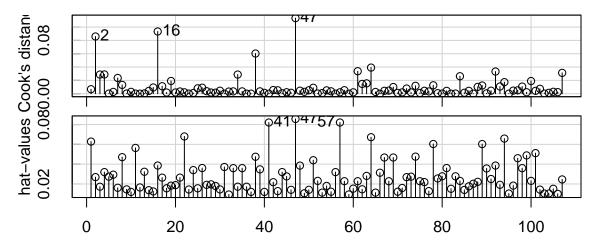
589 ## [1] 2 3 16

588

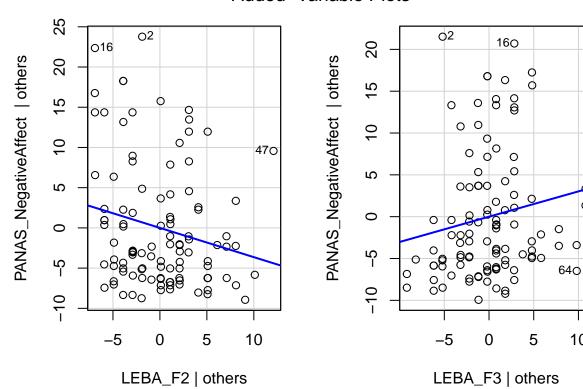
590 ## Non-constant Variance Score Test

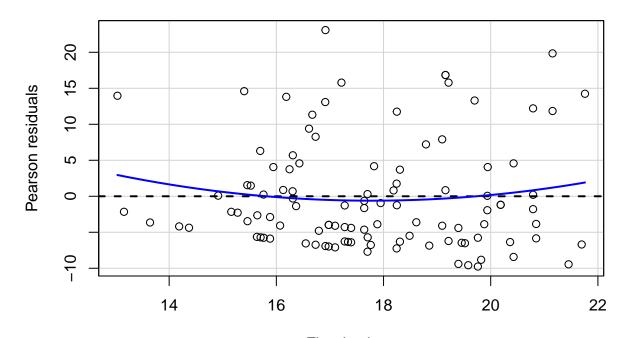
```
## Variance formula: ~ fitted.values
   ## Chisquare = 0.02201793, Df = 1, p = 0.88204
   ## Non-constant Variance Score Test
593
   ## Variance formula: ~ fitted.values
594
   ## Chisquare = 0.422668, Df = 1, p = 0.51561
595
   ## Non-constant Variance Score Test
   ## Variance formula: ~ fitted.values
   ## Chisquare = 1.729402, Df = 1, p = 0.18849
   ## Non-constant Variance Score Test
   ## Variance formula: ~ fitted.values
   ## Chisquare = 3.390299, Df = 1, p = 0.065581
601
   ## Non-constant Variance Score Test
   ## Variance formula: ~ fitted.values
   ## Chisquare = 0.001335325, Df = 1, p = 0.97085
   ## Non-constant Variance Score Test
   ## Variance formula: ~ fitted.values
   ## Chisquare = 3.098381, Df = 1, p = 0.07837
   ## No Studentized residuals with Bonferroni p < 0.05
   ## Largest |rstudent|:
         rstudent unadjusted p-value Bonferroni p
610
   ##
   ## 57 2.708363
                            0.0079428
                                           0.84988
```

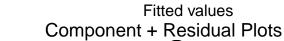
Diagnostic Plots

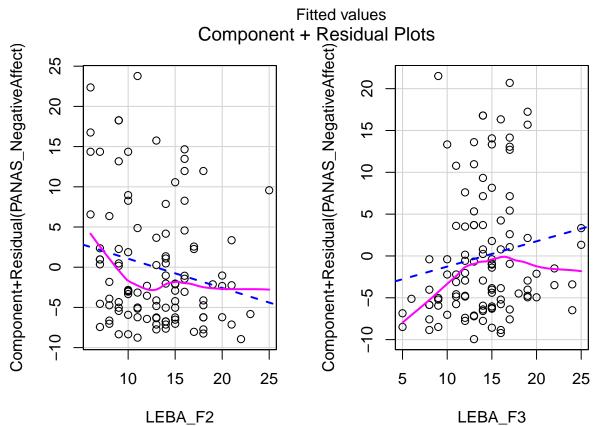


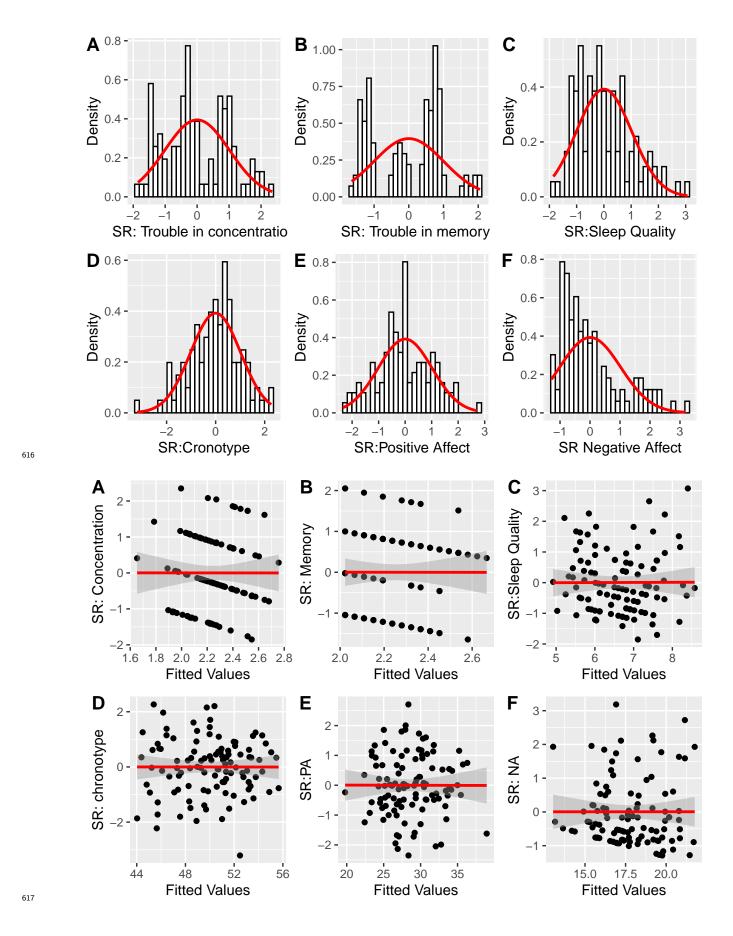
Index Added–Variable Plots











We used R [Version 4.1.2; R Core Team (2021)] and the R-packages }apaTables 618 [@}R-apaTables], boot [Version 1.3.28; Davison and Hinkley (1997)], car [Version 619 3.0.12; Fox and Weisberg (2019); Fox, Weisberg, and Price (2020)], carData [Version 620 3.0.4; Fox et al. (2020)], *dlookr* [Version 0.5.1; Ryu (2021)], *dplyr* [Version 1.0.7; 621 Wickham, François, Henry, and Müller (2021)], forcats [Version 0.5.1; Wickham (2021a)], 622 ggplot2 [Version 3.3.5; Wickham (2016)], gtExtras [Version 0.2.18; Mock (2021)], 623 atsummary [Version 1.4.2; Sjoberg et al. (2021)], kableExtra [Version 1.3.4; Zhu (2021)], 624 likert [Version 1.3.5; Bryer (2019)], MOTE [Version 1.0.2; Buchanan, Gillenwaters, 625 Scofield, and Valentine (2019)], packrat (Ushey, McPherson, Cheng, Atkins, & Allaire, 626 2021), papaja [Version 0.1.0.9997; Aust and Barth (2020)], plotly [Version 4.9.4.1; 627 Sievert (2020)], psych [Version 2.1.9; Revelle (2021)], purrr [Version 0.3.4; Henry and 628 Wickham (2020)], readr [Version 2.0.2; Wickham and Hester (2021)], scales [Version 1.1.1; Wickham and Seidel (2020)], shiny [Version 1.7.1; Chang et al. (2021)], sistats [Version 0.18.1; Lüdecke (2021)], stringr [Version 1.4.0; Wickham (2019)], tibble [Version 631 3.1.6; Müller and Wickham (2021)], tidyr [Version 1.1.4; Wickham (2021b)], tidyverse 632 [Version 1.3.1; Wickham et al. (2019)], visreg [Version 2.7.0; Breheny and Burchett 633 (2017)], and xtable [Version 1.8.4; Dahl, Scott, Roosen, Magnusson, and Swinton 634 (2019)] for all our analyses.

Results

637 Discussion

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Table 1

Dempgraphics

	Female, N =	Male, N =
Characteristic	71	36
Age	33 (9)	35 (14)
Religion		
Atheist	6 (8.5%)	3 (8.3%)
Buddhist	18 (25%)	13 (36%)
Christian	16 (23%)	6 (17%)
Hindu	13 (18%)	6 (17%)
Muslim	18 (25%)	8 (22%)
Ethnicity		
Malaysian Chinese	30 (42%)	16 (44%)
Malaysian Indian	11 (15%)	7 (19%)
Malaysian Malay	10 (14%)	1 (2.8%)
Others	20 (28%)	12 (33%)
Your marital status		
Divorced	1 (1.4%)	0 (0%)
Married	27 (38%)	18 (50%)
Single	43 (61%)	18 (50%)
Please state your current level of education - Selected		
Choice		
Bachelor's degree	20 (28%)	10 (28%)
Diploma	0 (0%)	2 (5.6%)
Doctor of Philosophy (PhD)	32 (45%)	8 (22%)
Master's degree	18 (25%)	15 (42%)
Pre-university	0 (0%)	1 (2.8%)

	Female, N =	Male, N =
Characteristic	71	36
Secondary School	1 (1.4%)	0 (0%)
Community_Stance	7.20 (1.92)	6.83 (1.99)
Time_of_ Day		
afternoon	46 (65%)	26 (72%)
evening	15 (21%)	6 (17%)
morning	5 (7.0%)	1 (2.8%)
night	5 (7.0%)	3 (8.3%)
Positive_Affect	29 (9)	28 (8)
Negative_Affect	18 (8)	18 (8)
PSQI	6.7 (3.4)	6.6 (3.6)
Sleep_Quality		
Good Sleep	23 (32%)	12 (33%)
Poor Sleep	48 (68%)	24 (67%)
Sleep_Environment	4.6 (4.4)	7.4 (5.5)
Avg_Corneal_Illuminance	206 (241)	185 (171)
MEQ	50 (8)	50 (8)
Chronotype		
Definite Evening	2 (2.8%)	1 (2.8%)
Intermediate	47 (66%)	26 (72%)
Moderate Evening	10 (14%)	4 (11%)
Moderate Morning	12 (17%)	5 (14%)
LEBA1	6.7 (4.5)	6.0 (4.6)
LEBA2	12.9 (4.4)	12.8 (4.0)
LEBA3	13.8 (4.1)	15.0 (3.7)
LEBA4	12.1 (4.2)	12.3 (4.3)

	Female, N =	Male, N =
Characteristic	71	36
LEBA5	10.5 (3.7)	10.1 (4.1)
Cups_of_Coffee(Weekday)	0.80 (1.06)	0.94 (1.39)
Cups_of Coffee(Weekend)	0.72 (1.01)	0.81 (1.33)
Your working /school/university shift		
Day Shift	52 (73%)	28 (78%)
Mixed shift work (Both Night and day in alternating way)	16 (23%)	3 (8.3%)
Night Shift	0 (0%)	1 (2.8%)
Off Work	3 (4.2%)	4 (11%)
Subjective_Alertness	3.49 (1.96)	4.28 (2.22)
Currently, How much are you bothered by: - 2. Trouble		
concentrating or thinking clearly		
Absent	17 (24%)	7 (19%)
Moderate	20 (28%)	12 (33%)
Severe	6 (8.5%)	3 (8.3%)
Slight	28 (39%)	14 (39%)
Currently, How much are you bothered by: - 5. Trouble with		
memory		
Absent	26 (37%)	9 (25%)
Moderate	12 (17%)	6 (17%)
Severe	5 (7.0%)	3 (8.3%)
Slight	28 (39%)	18 (50%)