

1 Psychometric Evaluation of the Bangla Communication Scale Among Adolescents: Latent
2 structure, reliability, validity, and gender invariance in a Bangladeshi adolescent sample

3 Nusrat Jahan¹, Mushfiqul Anwar Siraji², & Zinnatul Borak¹

4 ¹ Department of Educational and Counselling Psychology, University of Dhaka

5 ² Monash University, Department of Psychology, Jeffrey Cheah School of Medicine and
6 Health Sciences, Malaysia

8 Add complete departmental affiliations for each author here. Each new line herein
9 must be indented, like this line.

10 Enter author note here.

11 The authors made the following contributions. Nusrat Jahan: Conceptualization,
12 Project Management, Data Curation, Writing - Original Draft Preparation; Mushfiqul
13 Anwar Siraji: Conceptualization, Project Management, Formal Analysis & Data
14 Visualization, Writing - Original Draft Preparation; Review & Editing; Zinnatul Borak:
15 Conceptualization, Writing - Review & Editing.

16 Correspondence concerning this article should be addressed to Zinnatul Borak,
17 Department of Educational and Counselling Psychology, University of Dhaka, Dhaka
18 1000. E-mail: institutional email of Bonhee mam

Abstract

19

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

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

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

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

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

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

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

34 *Keywords:* keywords

35 Word count: X

Psychometric Evaluation of the Bangla Communication Scale Among Adolescents: Latent structure, reliability, validity, and gender invariance in a Bangladeshi adolescent sample

Communication is a complex behaviour of exchanging information among individuals (Tanner, 2006). Communication plays a central role among adolescents in developing self-identity, social relationships and creates the foundation of collective social activity (Conti-Ramsden & Botting, 2008; Haslett & Bowen, 1989; Spencer, Clegg, & Stackhouse, 2013). Inadequate communication skill may cause poor peer relationship resulting long-term socio-emotional difficulties including social anxiety, stress, low self-esteem and poor academic performance (Brinton & Fujiki, 2004; Reed & Trumbo, 2020).

Often adults picture adolescents having inadequate and inept communication skills (Stern, 2005; Thurlow, 2003). Media representation of adolescents often includes “storm-and stress,” self absorbed and disengaged type behaviours (Porteous & Colston, 1980; Stern, 2005). As such adolescents are often labelled as “lazy” and “disrespectful” by the adults (Agenda & America, 1999). On the contrary adolescents are highly engaged in work, community services and extracurricular activities and also more aspiring to earn an college degrees (DeBard, 2004; Schneider & Stevenson, 1999). Such a discrepancy between the reality of adolescent’s image and adult’s perception of the adolescents might be attributed to the mismatch of communication skills. The communication pattern of adolescents might not necessarily same as the adults. In addition to face to face communication, adolescents are vastly exposed to different virtual communication platforms. This may cause them to face more complex social challenges than the adults (Thurlow, 2003). “Communication capital” expresses the potential of civic-engagement that incorporates developing social relationships and influences collective social activity. The more communication capital an individual has the easier the instances of civic engagement become.

Understanding the adolescents' communication skill vital as it is considered as the "key skill" in the education (Thurlow, 2001) and employment market (Olszewski, Panorska, & Gillam, 2017). In the western society, adolescents are now facing high unemployment (Lindsay et al., 2014). Lack of adequate communication skill is one of the root causes of this high unemployment (Lindsay et al., 2014). Similarly, lack of proper communication skill often promotes the propensity of anti-social behaviours and risk of exclusion from schools (Clegg, Stackhouse, Finch, Murphy, & Nicholls, 2009; Conti-Ramsden & Botting, 2004).

To promote better understanding of subject contents assessing the communication skill among adolescents is highly required. For this purpose "Communication Skill" sub-skill set (Barkman & Machtmes, 2002) was developed in 2002 as a part of The National On-line Youth Life Skills Evaluation System (Mincemoyer, Perkins, & Munyua, 2005) and since then it has been extensively used (Fitzpatrick, Gagne, Jones, Loble, & Phelps, 2005).

Study-1: Translations and Exploratory Factor Analysis

Study-1 had two objectives. First we translated the 23 items of the Communication Scale (Barkman & Machtmes, 2002) from English to Bangla. Second, we conducted an Exploratory Factor Analysis (EFA) to identify the latent structure of the scale.

Methods

Participants. A cross-sectional survey was used to collect data from a large sample of students of grade 8-12 ($n = 300$) from 8 schools following convenience sampling method. One participant was excluded due to incomplete data. Participants were recruited following convenience sampling technique. For exploring the initial factor structure the recommended sample size is 250-300 (Comrey & Lee, 1992; Schönbrodt &

Perugini, 2013). Among 300 participants, 218 were female aged between 12 to 21 years (15.89±1.46). 82 were male with an age range between 13 participants 23 years (16.62±1.38). 282 (94%) participants belonged to middle socio-economic status. 13 (4.3%) and 5 (1.7%) participants belonged to lower and upper socio-economic status-respectively.

Materials.

Communication Scale.

Communication Scale (Barkman & Machtmes, 2002) is a sub-skill set of The National On-line Youth Life Skills Evaluation System (Mincemoyer et al., 2005). It has 23 items with a 5 Point Likert Type response scale (0=Never, 1=Rarely, 2=Sometimes, 3=Often, 4=Always). The total score range is 0-92 where a higher score would indicate higher communication skills among adolescents (age range 12-18). The internal consistency of the total scale, Cronbach α was .79

Bangla Communication Scale.

We followed the International Test Commission guidelines (Bartram et al., 2018) while translating the Communication Scale in Bangla. At first two bilingual researchers did the forward translation. These two forward translations were synthesized by the authors. Another four bilingual researchers did the backward translation of the synthesized Bangla Communication Scale. The authors again synthesized the back-translations, compared it with original scale and made necessary amendments.

Data collection. The project received institutional ethics clearance from Department of Educational and Counselling Psychology (Project ID: —). Prior to data collection necessary authorization from school's authority and assent from the participating were obtained. Necessary explanations were given in oral and written forms. It was also mentioned in the explanatory statement that their participation was voluntary and that they could withdraw from participation any time without being

penalized. Data collection was commenced between November 2021 to January 2022. The data collection took place in the classroom where students were at first briefed about 'communication skill.' Next, they filled up their socio-demographics information and responded to our Bangla Communication Scale. All personal information (name, school, class) was codified and encrypted, producing a anonymous database.

Analytic strategies. We used R (version 4.1.0) (R Core Team, 2021), including several R packages (Chalmers, 2012; Revelle, 2021; Rosseel, 2012; Siraji, 2021), for our analyses. Prior to the data analysis normality assumptions were checked. Our data violeted both univariate and multivariate normality. As such we used polychoric correlation matrix and 'principal axis" factor extraction method to conduct the exploratory factor analysis (Watkins, 2020). An oblique rotation method was employed. A posterior sampling adequacy was estimated using KMO statistics (Kaiser, 1974). TO identify the optimum number of factors required to explain the latent structure of our scale we used Scree plot (Cattell, 1966), Horn's parallel analysis (Horn, 1965), Hull method (Lorenzo-Seva, Timmerman, & Kiers, 2011) and Minimum average partials method (MAP) (Velicer, 1976). Lastly, to identify the simple structure, we followed the following guidelines (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 (Child, 2006; Mulaik, 2009; Watkins, 2020)

Results & Discussion.

Sampling adequacy.

Kaiser-Meyer-Olkin (KMO) (Kaiser, 1974) statistics was used to check the sampling adequacy. The overall KMO value for 23 items was 0.74, which was above the cut-off value of .50, indicating an adequate sample.

Descriptive statistics and item analysis.

Table1 reports univariate descriptive statistics for the 23 items. All items were

skewed and violated univariate normality assessed by The Shapiro-Wilk test indicated all the items violated normality assumptions (Shapiro & Wilk, 1965). Mardia's Test of multivariate normality (Mardia, 1970) yielded Multivariate skew = 4030.49 ($p < 0.001$) and multivariate kurtosis = 15.1 ($p < 0.001$) indicating the violation of multivariate normality as well. As such we used polychoric correlation matrix which is more robust towards these violations. Our initial item analysis yielded internal consistency coefficient ordinal alpha = 0.79. Alpha drop statistics (Table1) showed no substantial increase of ordinal alpha if any item is deleted and ranged between 0.8-0.8. As such we subjected all items to EFA.

Supplementary Table 1 and Figure 1 depict the inter-item correlation coefficients of BCS. Bartlett's test of sphericity (Bartlett, 1954), $\chi^2(253) = \text{rbartlet\$schisq}$, $p < .001$ indicated the inter-item correlations are significantly different than zero. However, only 9.88% of the inter-item correlation coefficients were greater than absolute value of .30 in the obtained matrix. The corrected item-total correlations range was 0.05- 0.56 (Table1).

#####Exploratory factor analysis

To identify optimum number of factor required to express the latent structure adequately we at first used Horn's parallel analysis (Horn, 1965). Horn's parallel analysis offered two factor solution to the latent structure (Figure 2). However, The Scree-plot (Cattell, 1966) suggested one factor solution. One factor solution was also supported by MAP method (Velicer, 1976) and Hull method (Lorenzo-Seva et al., 2011). Minimum average partial (MAP) method expects the average squared off-diagonal values of the calculated partial correlation matrix to be minimum when the correct number of factors are extracted (Velicer, 1976). In our data set this value reached the minimum after extracting the first factor (Supplementary Table 2). Hull method tried to find an optimal number of factors to balance model fit and the number of parameters and offered one factor solution in our data set (Figure 2). As a result, we tested both one-factor and two-factor solutions.

At first we fitted one factor solution in our dataset with all 23 items. The initial solution had six items with poor factor loadings (Item 1, 2, 5, 9, 16, 23). We discarded these items from the model and run another EFA. This iteration of EFA yielded a one factor simple structure with 17 items. The one factor solution explained 20.87% of the total variance. Cronbach's alpha coefficient has a tendency to deflate the estimates for Likert type data as the calculation is based on pearson-correlation matrix which requires data with continuous measurement level (Gadermann, Guhn, & Zumbo, 2012; Zumbo, Gadermann, & Zeisser, 2007). For better estimates of reliability we reported ordinal alpha using polychoric-correlation (Zumbo et al., 2007). The internal consistency reliability coefficient for this model was, ordinal $\alpha = .81$.

The initial two-factor solution with all 23 items had several items (item 01, 03, 05, 09, 16) with poor factor loading ($<.3$) . We conducted another two round of EFA and identified and discarded problematic items (with poor factor loadings and/or cross loadings). Finally, a two-factor EFA solution with 15 items was accepted with RMSR = 0.06, no loading smaller than .30 and no cross-loading greater than .30. The first factor retained 8 items, and the second factor retained 6 items. The first factor explained 15.42% of the total variance and the second factor explained only 10.75. Ordinal alpha coefficient value ranges from 0 to 1 and higher value represents better reliability. The internal consistency reliability coefficient for the two factors were, ordinal $\alpha = .75$ and .66 respectively. Items clustered under factor 1 were stemmed from several common theme including efficiency of verbal communication, individual's efficiency in verbal communication. All items except item12 (I try to see the other person's point of view) investigated individual's efficiency in non-verbal communication. The average variance extracted (AVE) is the amount of joint variance captured by the components and not by measurement error. Fornell and Larcker (1981) recommended .50 as a benchmark value of AVE to establish the converging validity and purity of particular factor. AVE for these two factors were .25 and .23 indicating high measurement error. Due to such high

measurement error of the factor we rejected the two factor model and accepted the one factor model which had satisfactory reliability

Measurement Invariance

To gather more information on our retained one-factor solution, we sought Item Response Theory (IRT). IRT complements the conventional classical test theory-based analysis by gathering information on item discrimination and item difficulty. IRT judges an item's quality by providing item information in the light of participants' trait level (θ). We gathered evidence on item quality as well as item fit, person fit and model by fitting a graded response model in RStudio with the "mirt" package (Chalmers, 2012) (Chalmers, 2012). We did a Monte Carlo simulation using "SimDesign" package (Chalmers & Adkins, 2020) with sample sizes varying from 50-350 and calculated average root mean squared error(RMSE) to estimate the optimal sample size for the graded response model with 23 items. The RMSE became stable for $n = 200$ to 300 (RMSE ranging between .25-.35). Our sample size within the estimated sample size for stability.

Marginal reliability is based on the true score model (Lord & Novick, 1968) and is an estimate of the overall reliability of a test based on the average conditional standard errors. Often it is close in value to coefficient alpha (and sometimes it may even be identical). Alpha provides a lower estimate of marginal reliability.

Results

Discussion

Ethical Consideration

All procedures performed in studies involving human participants were in accordance with the 1964 Helsinki declaration and its later amendments or comparable

214 ethical standards. This article does not contain any studies with animals performed by
215 any of the authors.

216 **Data and code availability**

217 All code and data underlying this article is available on a public GitHub repository
218 (<https://github.com/masiraji/Communication>).

References

- Agenda, P., & America, U. S. of. (1999). Kids these days' 99: What americans really think about the next generation.
- Aust, F., & Barth, M. (2020). *papaja: Create APA manuscripts with R Markdown*. Retrieved from <https://github.com/crsh/papaja>
- Barkman, S., & Machtmes, K. (2002). Four-fold: A research model for designing and evaluating the impact of youth development programs. *News and Views*, 4(4), 4–6.
- Barnier, J., Briatte, F., & Larmarange, J. (2021). *Questionr: Functions to make surveys processing easier*. Retrieved from <https://CRAN.R-project.org/package=questionr>
- Bartlett, M. (1954). A note on the multiplying factors for Various#«2 approximations. *Journal of the Royal Statistical Society Series b-Methodological*, 16, 296–298.
- Bartram, D., Berberoglu, G., Grégoire, J., Hambleton, R., Muniz, J., & Vijver, F. van de. (2018). ITC Guidelines for Translating and Adapting Tests (Second Edition). *International Journal of Testing*, 18(2), 101–134. <https://doi.org/10.1080/15305058.2017.1398166>
- Brinton, B., & Fujiki, M. (2004). Social and affective factors in children with language impairment: Implications for literacy learning. *Handbook of Language and Literacy*, 130–153.
- Bryer, J., & Speerschneider, K. (2016). *Likert: Analysis and visualization likert items*. Retrieved from <https://CRAN.R-project.org/package=likert>
- Buchanan, E. M., Gillenwaters, A., Scofield, J. E., & Valentine, K. D. (2019). *MOTE: Measure of the Effect: Package to assist in effect size calculations and their confidence intervals*. Retrieved from <http://github.com/doomlab/MOTE>
- Butts, C. T. (2008). Network: A package for managing relational data in r. *Journal*

246 *of Statistical Software*, 24(2). Retrieved from

247 <https://www.jstatsoft.org/v24/i02/paper>

248 Butts, C. T. (2020). *Sna: Tools for social network analysis*. Retrieved from

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

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

252 https://doi.org/10.1207/s15327906mbr0102_10

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

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

256 Chalmers, R. P., & Adkins, M. C. (2020). Writing effective and reliable Monte
257 Carlo simulations with the SimDesign package. *The Quantitative Methods for*
258 *Psychology*, 16(4), 248–280. <https://doi.org/10.20982/tqmp.16.4.p248>

259 Chang, W., Cheng, J., Allaire, J., Sievert, C., Schloerke, B., Xie, Y., ... Borges, B.
260 (2021). *Shiny: Web application framework for r*. Retrieved from

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

262 Child, D. (2006). *Essentials of factor analysis* (3rd ed.). New York: Continuum.

263 Clegg, J., Stackhouse, J., Finch, K., Murphy, C., & Nicholls, S. (2009). Language
264 abilities of secondary age pupils at risk of school exclusion: A preliminary
265 report. *Child Language Teaching and Therapy*, 25(1), 123–139.

266 Comrey, A. L., & Lee, H. B. (1992). *A first course in factor analysis*, 2nd ed.
267 Hillsdale, NJ, US: Lawrence Erlbaum Associates, Inc.

268 Conti-Ramsden, G., & Botting, N. (2004). Social difficulties and victimization in
269 children with SLI at 11 years of age.

270 Conti-Ramsden, G., & Botting, N. (2008). Emotional health in adolescents with
271 and without a history of specific language impairment (SLI). *Journal of Child*
272 *Psychology and Psychiatry*, 49(5), 516–525.

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

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

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

276 DeBard, R. (2004). Millennials coming to college. *New Directions for Student*

277 *Services*, 2004(106), 33–45.

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

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

280 Epskamp, S. (2019). *semPlot: Path diagrams and visual analysis of various SEM*

281 *packages' output*. Retrieved from

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

283 Epskamp, S., Cramer, A. O. J., Waldorp, L. J., Schmittmann, V. D., & Borsboom,

284 D. (2012). qgraph: Network visualizations of relationships in psychometric

285 data. *Journal of Statistical Software*, 48(4), 1–18.

286 Fitzpatrick, C., Gagne, K. H., Jones, R., Loble, J., & Phelps, L. (2005). Life skills

287 development in youth: Impact research in action. *Learning*, 72, 77.

288 Fornell, C., & Larcker, D. F. (1981). Evaluating structural equation models with

289 unobservable variables and measurement error. *Journal of Marketing*

290 *Research*, 18(1), 39–50.

291 Fox, J., & Weisberg, S. (2019). *An R companion to applied regression* (Third).

292 Thousand Oaks CA: Sage. Retrieved from

293 <https://socialsciences.mcmaster.ca/jfox/Books/Companion/>

294 Fox, J., Weisberg, S., & Price, B. (2020). *carData: Companion to applied*

295 *regression data sets*. Retrieved from

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

297 Gadermann, A. M., Guhn, M., & Zumbo, B. D. (2012). Estimating ordinal reliability

298 for likert-type and ordinal item response data: A conceptual, empirical, and

299 practical guide. *Practical Assessment, Research, and Evaluation*, 17(1), 3.

Golino, H., & Christensen, A. P. (2021). *EGAnet: Exploratory graph analysis – a framework for estimating the number of dimensions in multivariate data using network psychometrics*.

Haslett, B., & Bowen, S. P. (1989). Children's strategies in initiating interaction with peers.

Henry, L., & Wickham, H. (2020). *Purrr: Functional programming tools*. Retrieved from <https://CRAN.R-project.org/package=purrr>

Horn, J. L. (1965). A rationale and test for the number of factors in factor analysis. *Psychometrika*, 30(2), 179–185. <https://doi.org/10.1007/BF02289447>

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

Iannone, R. (2020). *DiagrammeR: Graph/network visualization*. Retrieved from <https://CRAN.R-project.org/package=DiagrammeR>

Iannone, R., Cheng, J., & Schloerke, B. (2021). *Gt: Easily create presentation-ready display tables*. Retrieved from <https://CRAN.R-project.org/package=gt>

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

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

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

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

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

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

Kowarik, A., & Templ, M. (2016). Imputation with the R package VIM. *Journal of Statistical Software*, 74(7), 1–16. <https://doi.org/10.18637/jss.v074.i07>

Krivitsky, P. N. (2021). *Statnet.common: Common r scripts and utilities used by the statnet project software*. The Statnet Project (<https://statnet.org>).

Retrieved from <https://CRAN.R-project.org/package=statnet.common>

Lindsay, S., Adams, T., Sanford, R., McDougall, C., Kingsnorth, S., & Menna-Dack, D. (2014). Employers' and employment counselors' perceptions of desirable skills for entry-level positions for adolescents: How does it differ for youth with disabilities? *Disability & Society*, 29(6), 953–967.

Lorenzo-Seva, U., Timmerman, M., & Kiers, H. (2011). The Hull Method for Selecting the Number of Common Factors. *Multivariate Behavioral Research*, 46, 340–364. <https://doi.org/10.1080/00273171.2011.564527>

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

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

Michalke, M. (2020a). *koRpus.lang.en: Language support for 'koRpus' package: english*. Retrieved from <https://reaktanz.de/?c=hacking&s=koRpus>

Michalke, M. (2020b). *Sylly: Hyphenation and syllable counting for text analysis*. Retrieved from <https://reaktanz.de/?c=hacking&s=sylly>

Michalke, M. (2021). *koRpus: Text analysis with emphasis on POS tagging, readability, and lexical diversity*. Retrieved from <https://reaktanz.de/?c=hacking&s=koRpus>

Mincemoyer, C. C., Perkins, D. F., & Munyua, C. (2005). Measuring the impact of youth development programs: A national on-line youth life skills evaluation system. In *The forum for family and consumer issues* (Vol. 10, pp. 1–9).

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

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

Mulaik, S. A. (2009). *Foundations of Factor Analysis* (Vol. 7). London: London:

Chapman; Hall/CRC. <https://doi.org/10.1201/b15851>

Müller, K., & Wickham, H. (2021). *Tibble: Simple data frames*. Retrieved from

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

Navarro-Gonzalez, D., & Lorenzo-Seva, U. (2021). *EFA.MRFA: Dimensionality*

assessment using minimum rank factor analysis. Retrieved from

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

Neuwirth, E. (2014). *RColorBrewer: ColorBrewer palettes*. Retrieved from

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

O'Hara-Wild, M. (2022). *Icons: SVG icons for r documents and apps*.

Olszewski, A., Panorska, A., & Gillam, S. L. (2017). Training verbal and nonverbal

communication interview skills to adolescents. *Communication Disorders*

Quarterly, 38(4), 206–218.

Ooms, J. (2021a). *Magick: Advanced graphics and image-processing in r*.

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

Ooms, J. (2021b). *Rsvg: Render SVG images into PDF, PNG, PostScript, or*

bitmap arrays. Retrieved from <https://CRAN.R-project.org/package=rsvg>

Pornprasertmanit, S., Miller, P., Schoemann, A., & Jorgensen, T. D. (2021).

Simsem: SIMulated structural equation modeling. Retrieved from

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

Porteous, M., & Colston, N. (1980). How adolescents are reported in the british

press. *Journal of Adolescence*, 3(3), 197.

R Core Team. (2020). *Foreign: Read data stored by 'minitab', 's', 'SAS', 'SPSS',*

'stata', 'sysstat', 'weka', 'dBase', ... Retrieved from

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

- 381 R Core Team. (2021). *R: A language and environment for statistical computing*.
382 Vienna, Austria: R Foundation for Statistical Computing. Retrieved from
383 <https://www.R-project.org/>
- 384 Reed, V. A., & Trumbo, S. (2020). The relative importance of selected
385 communication skills for positive peer relations: American adolescents'
386 opinions. *Communication Disorders Quarterly*, 41(3), 135–150.
- 387 Revelle, W. (2021). *Psych: Procedures for psychological, psychometric, and*
388 *personality research*. Evanston, Illinois: Northwestern University. Retrieved
389 from <https://CRAN.R-project.org/package=psych>
- 390 Rosseel, Y. (2012). lavaan: An R package for structural equation modeling.
391 *Journal of Statistical Software*, 48(2), 1–36. Retrieved from
392 <https://www.jstatsoft.org/v48/i02/>
- 393 Ryu, C. (2021). *Dlookr: Tools for data diagnosis, exploration, transformation*.
394 Retrieved from <https://CRAN.R-project.org/package=dlookr>
- 395 Sarkar, D. (2008). *Lattice: Multivariate data visualization with r*. New York:
396 Springer. Retrieved from <http://lmdvr.r-forge.r-project.org>
- 397 Schneider, B., & Stevenson, D. (1999). *The ambitious generation*. New Haven,
398 CT: Yale University Press.
- 399 Schönbrodt, F. D., & Perugini, M. (2013). At what sample size do correlations
400 stabilize? *Journal of Research in Personality*, 47(5), 609–612.
401 <https://doi.org/10.1016/j.jrp.2013.05.009>
- 402 Shapiro, S. S., & Wilk, M. B. (1965). An analysis of variance test for normality
403 (complete samples). *Biometrika*, 52(3/4), 591–611.
- 404 Siraji, M. A. (2021). *Tabledown: A companion pack for the book "basic &*
405 *advanced psychometrics in r"*. Retrieved from
406 <https://github.com/masiraji/tabledown>
- 407 Sjöberg, D. D., Whiting, K., Curry, M., Lavery, J. A., & Larmarange, J. (2021).

Reproducible summary tables with the gtsummary package. *The R Journal*,
13, 570–580. <https://doi.org/10.32614/RJ-2021-053>

Spencer, S., Clegg, J., & Stackhouse, J. (2013). Language, social class and
education: Listening to adolescents' perceptions. *Language and Education*,
27(2), 129–143.

Stauffer, R., Mayr, G. J., Dabernig, M., & Zeileis, A. (2009). Somewhere over the
rainbow: How to make effective use of colors in meteorological visualizations.
Bulletin of the American Meteorological Society, 96(2), 203–216.
<https://doi.org/10.1175/BAMS-D-13-00155.1>

Stern, S. R. (2005). Self-absorbed, dangerous, and disengaged: What popular
films tell us about teenagers. *Mass Communication & Society*, 8(1), 23–38.

Tanner, D. C. (2006). *An advanced course in communication sciences and
disorders*. Plural Pub Incorporated.

Thurlow, C. (2001). Talkin"bout my communication: Communication awareness in
mid-adolescence. *Language Awareness*, 10(2-3), 213–231.

Thurlow, C. (2003). Teenagers in communication, teenagers on communication.
Journal of Language and Social Psychology, 22(1), 50–57.

Velicer, W. (1976). Determining the Number of Components from the Matrix of
Partial Correlations. *Psychometrika*, 41, 321–327.
<https://doi.org/10.1007/BF02293557>

Venables, W. N., & Ripley, B. D. (2002). *Modern applied statistics with s* (Fourth).
New York: Springer. Retrieved from <https://www.stats.ox.ac.uk/pub/MASS4/>

Watkins, M. (2020). *A Step-by-Step Guide to Exploratory Factor Analysis with R
and RStudio*. <https://doi.org/10.4324/9781003120001>

Wickham, H. (2007). Reshaping data with the reshape package. *Journal of
Statistical Software*, 21(12). Retrieved from
<http://www.jstatsoft.org/v21/i12/paper>

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

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

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

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

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

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

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

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

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

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

457 Xiao, N. (2018). *Ggsci: Scientific journal and sci-fi themed color palettes for*
458 *'ggplot2'*. Retrieved from <https://CRAN.R-project.org/package=ggsci>

459 Xie, Y., Cheng, J., & Tan, X. (2021). *DT: A wrapper of the JavaScript library*
460 *'DataTables'*. Retrieved from <https://CRAN.R-project.org/package=DT>

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

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

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

465 Zeileis, A., Hornik, K., & Murrell, P. (2009). Escaping RGBland: Selecting colors
466 for statistical graphics. *Computational Statistics & Data Analysis*, 53(9),
467 3259–3270. <https://doi.org/10.1016/j.csda.2008.11.033>

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

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

470 Zumbo, B. D., Gadermann, A. M., & Zeisser, C. (2007). Ordinal versions of
471 coefficients alpha and theta for likert rating scales. *Journal of Modern Applied*
472 *Statistical Methods*, 6(1), 4.

Table 1

Items	Mean	SD	Skew	Kurtosis	Shapiro-Wilk Statistics	Item-Total Correlation	Alpha Dro
CS01	2.97	1.02	-0.65	-0.43	0.84*	0.20	0.79
RCS02	2.51	1.18	-0.26	-0.96	0.89*	0.26	0.79
CS03	2.94	1.09	-0.72	-0.44	0.84*	0.35	0.78
CS04	2.47	1.23	-0.42	-0.79	0.89*	0.33	0.79
RCS05	2.50	1.31	-0.48	-0.87	0.87*	0.08	0.80
CS06	2.82	0.97	-0.62	-0.01	0.87*	0.40	0.78
CS07	2.70	1.16	-0.63	-0.59	0.86*	0.41	0.78
CS08	2.92	1.07	-0.87	0.00	0.84*	0.36	0.78
CS09	3.60	0.76	-2.24	4.73	0.56*	0.18	0.79
CS10	3.08	0.97	-0.81	-0.18	0.82*	0.34	0.78
CS11	3.17	1.00	-1.07	0.40	0.78*	0.52	0.77
CS12	2.58	1.19	-0.40	-0.83	0.89*	0.43	0.78
CS13	3.22	1.12	-1.35	0.81	0.72*	0.29	0.79
CS14	2.82	1.10	-0.79	-0.06	0.86*	0.43	0.78
CS15	3.34	0.82	-1.19	1.19	0.76*	0.51	0.77
CS16	2.63	1.26	-0.45	-1.00	0.87*	0.28	0.79
CS17	2.20	1.33	-0.04	-1.22	0.89*	0.36	0.78
CS18	2.86	1.07	-0.76	-0.11	0.85*	0.38	0.78
CS19	2.03	1.23	-0.08	-0.93	0.91*	0.44	0.78
CS20	2.73	1.00	-0.44	-0.47	0.88*	0.52	0.78
CS21	2.79	1.08	-0.66	-0.25	0.87*	0.56	0.77
CS22	3.00	1.06	-0.94	0.26	0.82*	0.44	0.78
CS23	2.31	1.22	-0.12	-0.91	0.90*	0.05	0.80

Table 2

Factor loadings of two factor model

item	PA1	PA2	Communality	Uniqueness
CS15	0.71		0.494	0.506
CS22	0.65		0.319	0.681
CS20	0.53		0.333	0.667
CS11	0.5		0.368	0.632
CS18	0.48		0.195	0.805
CS14	0.45		0.248	0.752
CS13	0.42		0.163	0.837
CS06	0.34		0.166	0.834
CS10	0.32		0.115	0.885
CS19		0.64	0.372	0.628
CS04		0.61	0.302	0.698
CS17		0.51	0.231	0.769
CS12		0.47	0.25	0.75
CS08		0.39	0.171	0.829
CS07		0.35	0.198	0.802
% of Variance	0.15	0.11		

Table 3

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*	NA
Configural	245.13	238.00	0.99	0.99	0.01	0.00	0.04	0.08	-	-
Metric	280.35	254.00	0.98	0.97	0.03	0.00	0.04	0.08	13.481a	0.637
Scalar	290.78	270.00	0.98	0.98	0.02	0.00	0.04	0.08	13.002b	0.673
Residual	303.44	287.00	0.98	0.98	0.02	0.00	0.04	0.09	14.008c	0.667

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

Table 4

Items discrimination and response category difficulty thresholds of 17 items in Bangla Communication

Items	a	b1	b2	b3	b4	Standardized Outfit	Standardized Infit	S-Chi-square
CS03	0.67	-5.51	-3.25	-1.18	0.69	-0.73	-0.59	80.74
CS04	0.92	-2.92	-1.46	-0.15	1.61	-1.34	-1.09	86.36
CS06	0.72	-5.36	-3.25	-1.09	1.50	-0.50	-0.61	101.14
CS07	0.90	-3.83	-2.04	-0.76	1.16	-1.18	-0.95	77.50
CS08	0.88	-4.44	-2.63	-1.19	0.75	-1.10	-0.82	71.65
CS10	0.64	-6.98	-3.98	-1.57	0.84	-0.62	-0.50	89.19
CS11	1.22	-4.06	-2.83	-1.40	-0.10	-1.18	-0.87	60.34
CS12	0.82	-3.82	-1.97	-0.34	1.23	-0.80	-0.91	76.70
CS13	0.75	-4.81	-3.17	-1.89	-0.29	-0.75	-0.53	76.21
CS14	0.75	-4.88	-2.93	-0.99	1.34	-0.96	-0.80	64.03
CS15	1.16	-4.00	-3.30	-1.70	-0.09	-1.10	-0.77	64.76
CS17	0.89	-2.74	-0.94	0.43	1.78	-1.14	-1.26	75.92
CS18	0.71	-5.54	-2.98	-1.37	1.02	-0.61	-0.61	92.33
CS19	1.01	-2.11	-0.92	0.60	2.31	-0.87	-1.13	96.62
CS20	1.15	-4.16	-2.10	-0.60	1.22	-1.56	-1.49	69.01
CS21	1.20	-3.46	-2.08	-0.73	0.91	-1.67	-1.39	56.50
CS22	0.66	-5.71	-3.67	-1.66	0.68	-0.79	-0.53	70.85

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

Table 5

Correlation matrix of the main variables

	1	2	3	4	5
1 Communication					
2 F1	.83**				
3 F2	.83**	.45**			
4 Hopelessness	-.16*	-.11	-.15+		
5 Life Satisfaction	-.02	-.02	.00	.02	
6 SE	.04	.02	.01	-.58**	-.07

Note. **p < .001

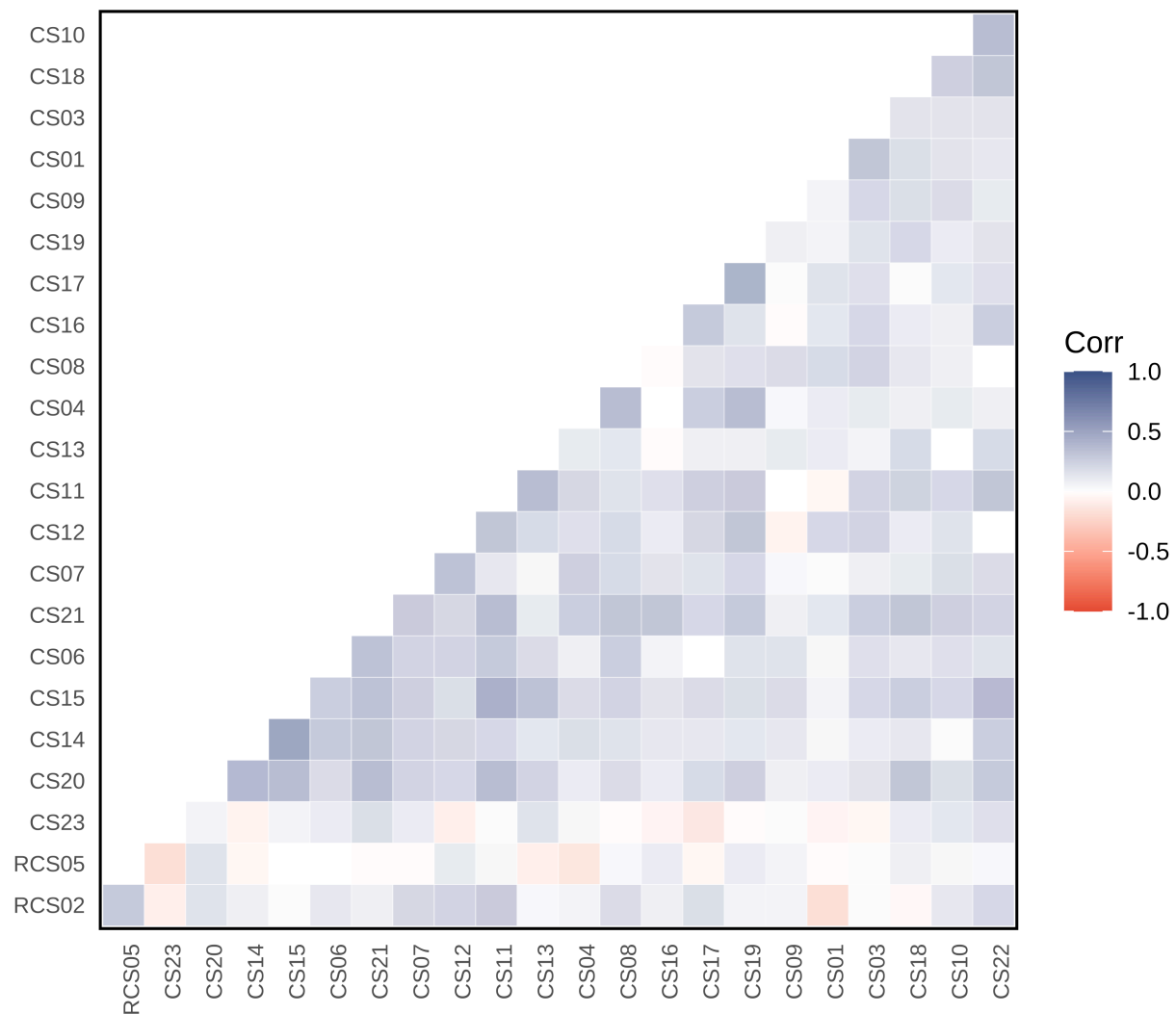


Figure 1. Inter item polychoric correlation coefficients for the 23 items. 9.8 % inter-item correlation coefficients were higher than $|\text{.30}|$

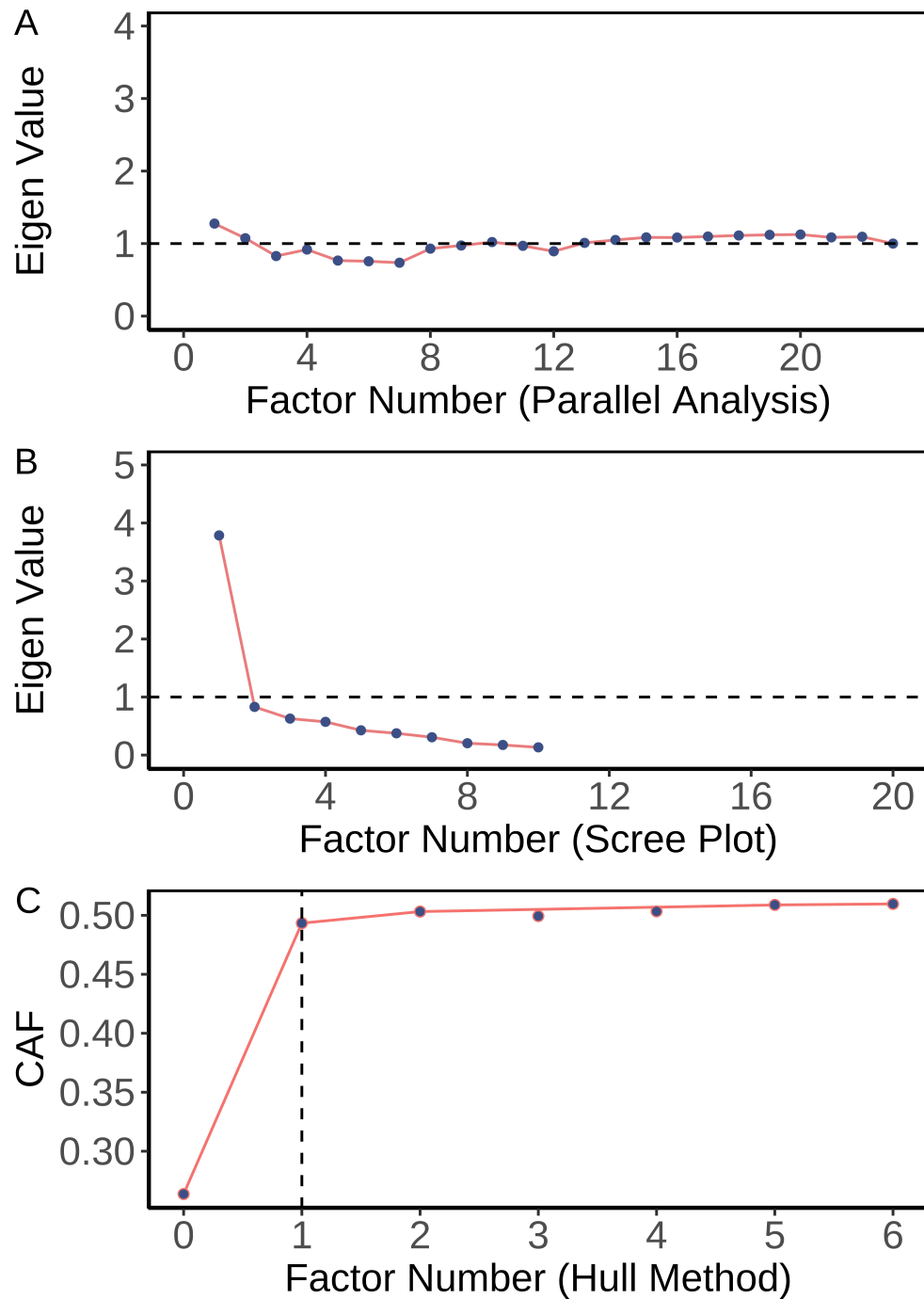


Figure 2. Factor Identification Methods (A) Parallel analysis indicated the optimal number of factors were two. (B) Scree plot suggested one factor. (C) Hull method indicated one factor were required to balance the model fit and number of parameters.

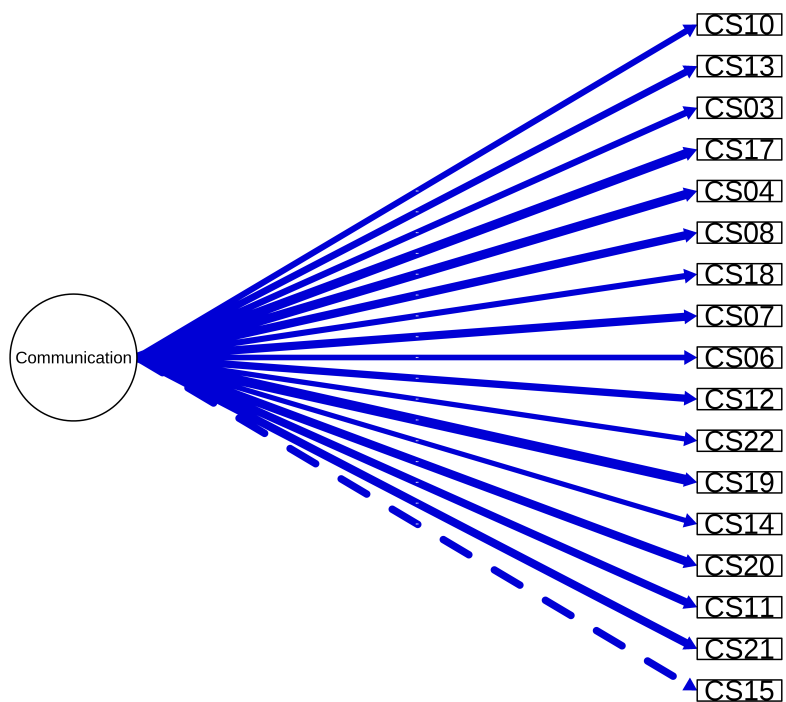


Figure 3. CFA Plot.

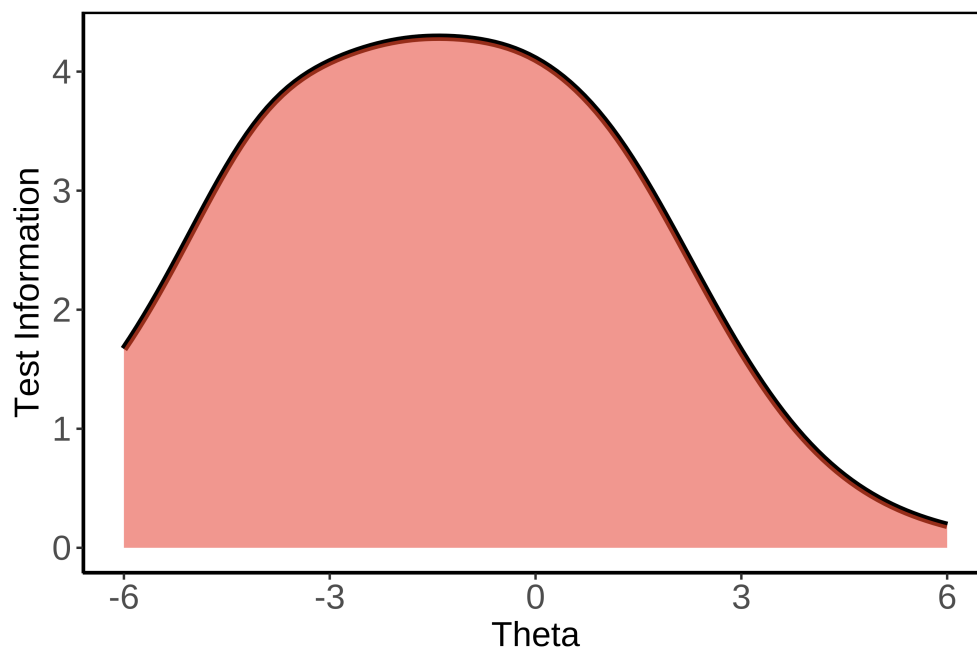


Figure 4. TIC.

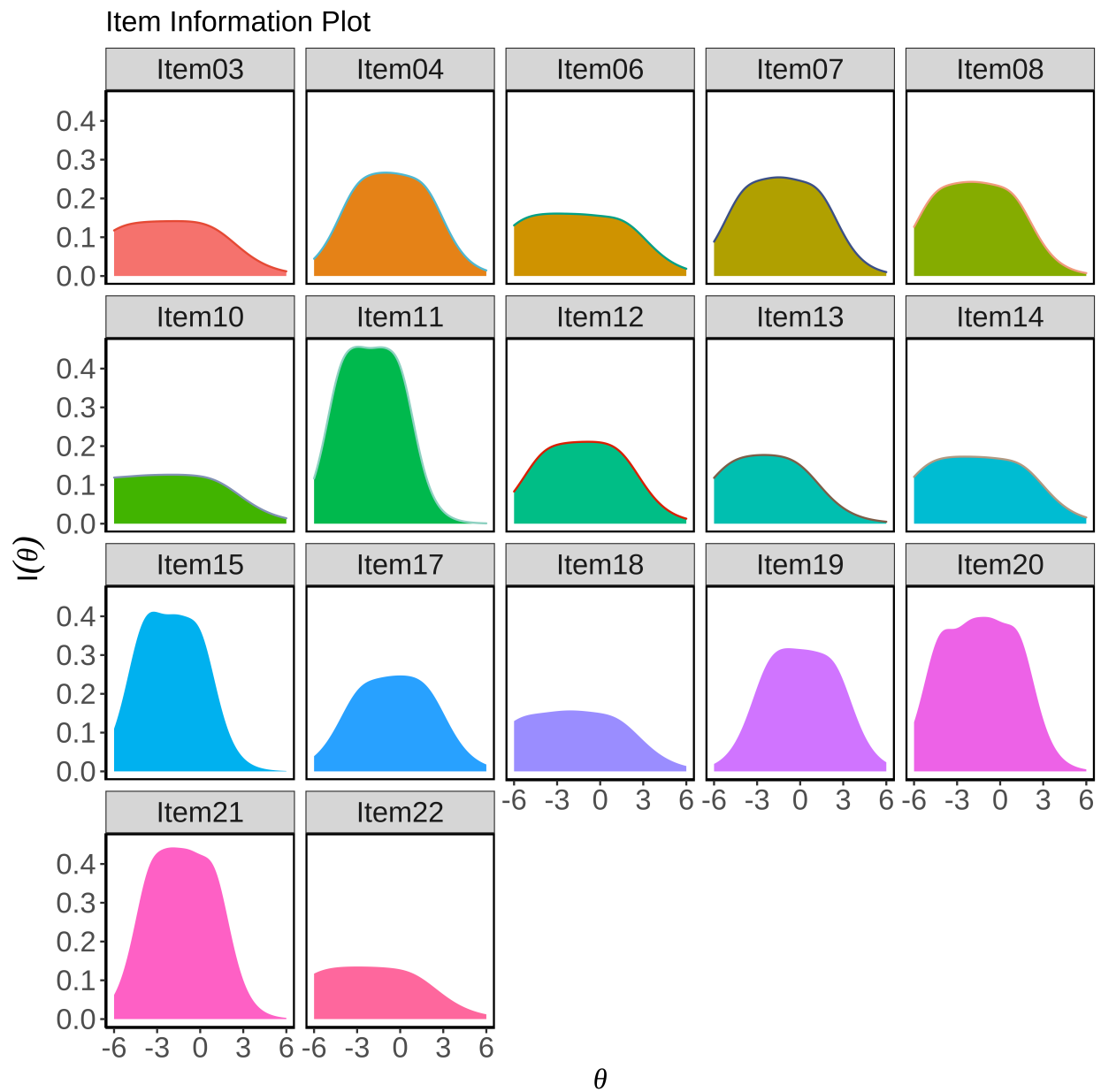


Figure 5. Item Information Curve.

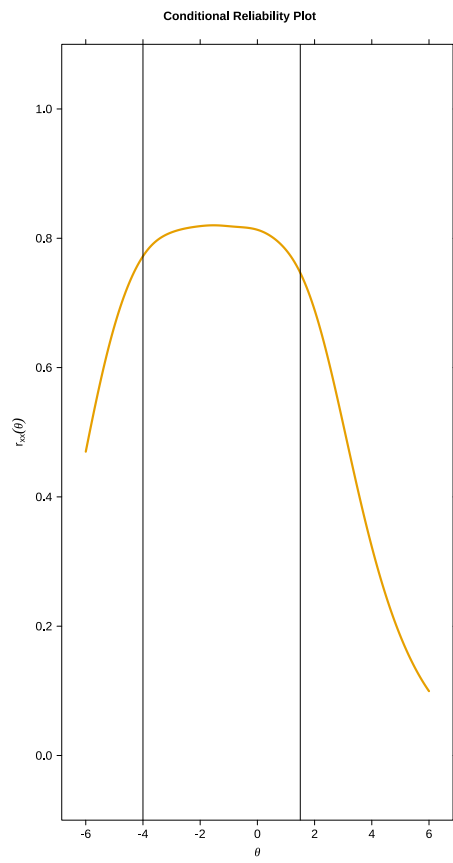


Figure 6. Conditional Reliability.