

Supplementary materials 2 - Appendix

In this appendix, we describe procedures to establish measurement invariance and measurement model fit, as well as report item parameter estimates and plots showing the reliability and the fit of the measurement model.

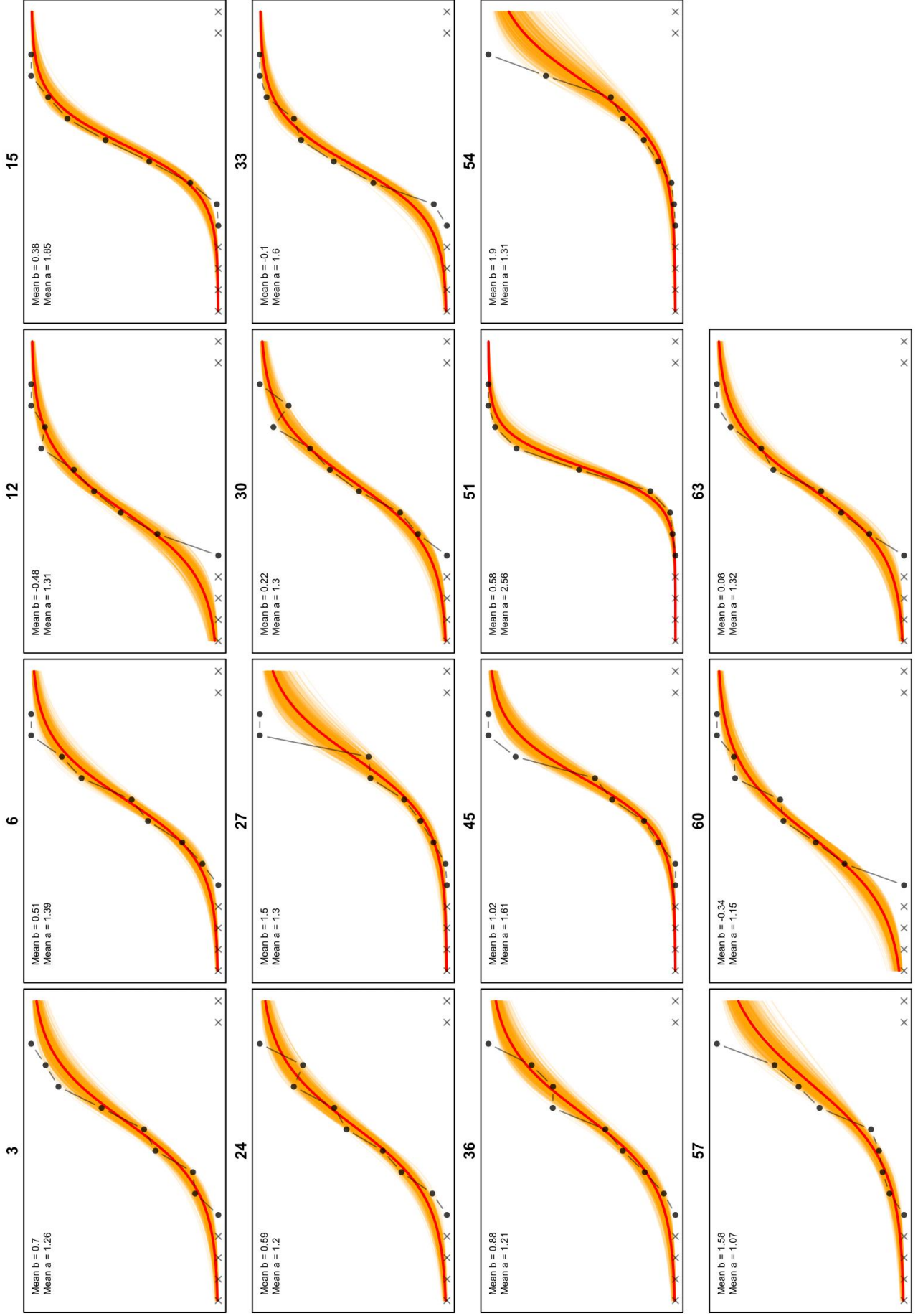
Establishing measurement invariance using leave-one-out crossvalidation

Latent difference score modelling requires measurement invariance across informants (de Haan et al., 2018). Following the procedure outlined by Verhagen and Fox (2013), we fitted IRT models to the CBQ data with item parameters allowed to differ between parent and adolescent informants and models with item parameters constrained to be equal. We then compared the fit of these models for each item using approximate leave-one-out cross-validation (Vehtari et al., 2019). Seven items (CBQ item numbers 9, 18, 21, 39, 42, 48, and 66) were found to have substantially better fit with item parameters allowed to differ between parents and adolescents, and these items were removed in the analysis.

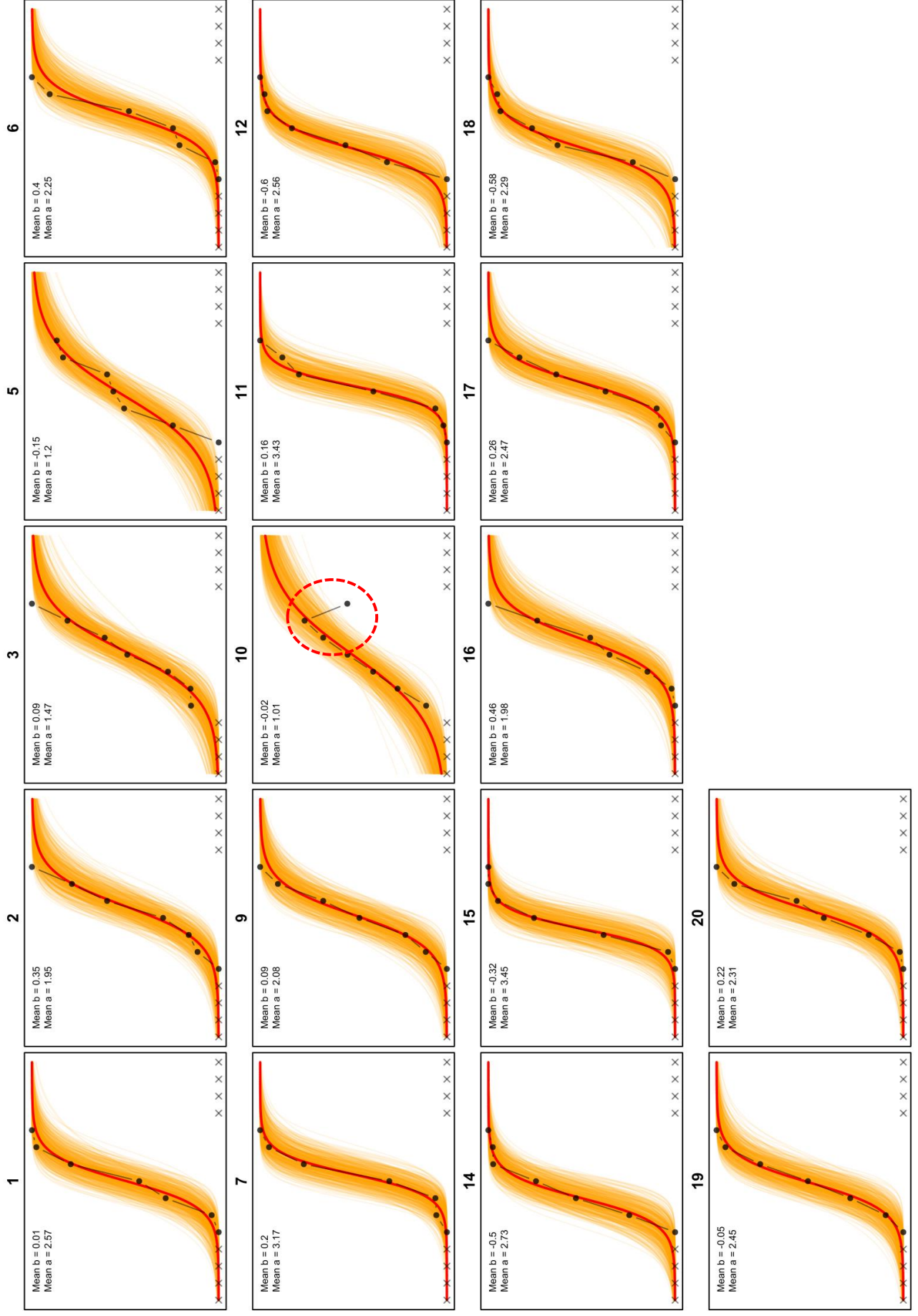
Evaluating IRT model fit by plotting

To evaluate the fit of the IRT model we plotted the observed probabilities of endorsing an item at different levels of the latent conflict variable together with item characteristic curves from the fitted model. We used the posterior means of the latent conflict variable and divided those into groups by 0.5 increments of the latent variable. For each item we then computed the proportion of endorsements in each group, giving the observed probability of endorsement at that approximate level of conflict. We plotted these probabilities along with 500 item characteristic curves sampled from the posterior distributions of the item parameters. This allowed us to assess the fit of the IRT model to the data, by comparing the observed probabilities of endorsement at different levels of latent trait with the posterior distribution of endorsement probabilities predicted by the model (Orlando & Thissen, 2000). These plots (see the following three pages) indicated acceptable or good fit of the IRT models for most items, except item 10 of the BHS. This item was also excluded from the analysis, bringing the final number of included items to 15 (CBQ) and 16 (BHS).

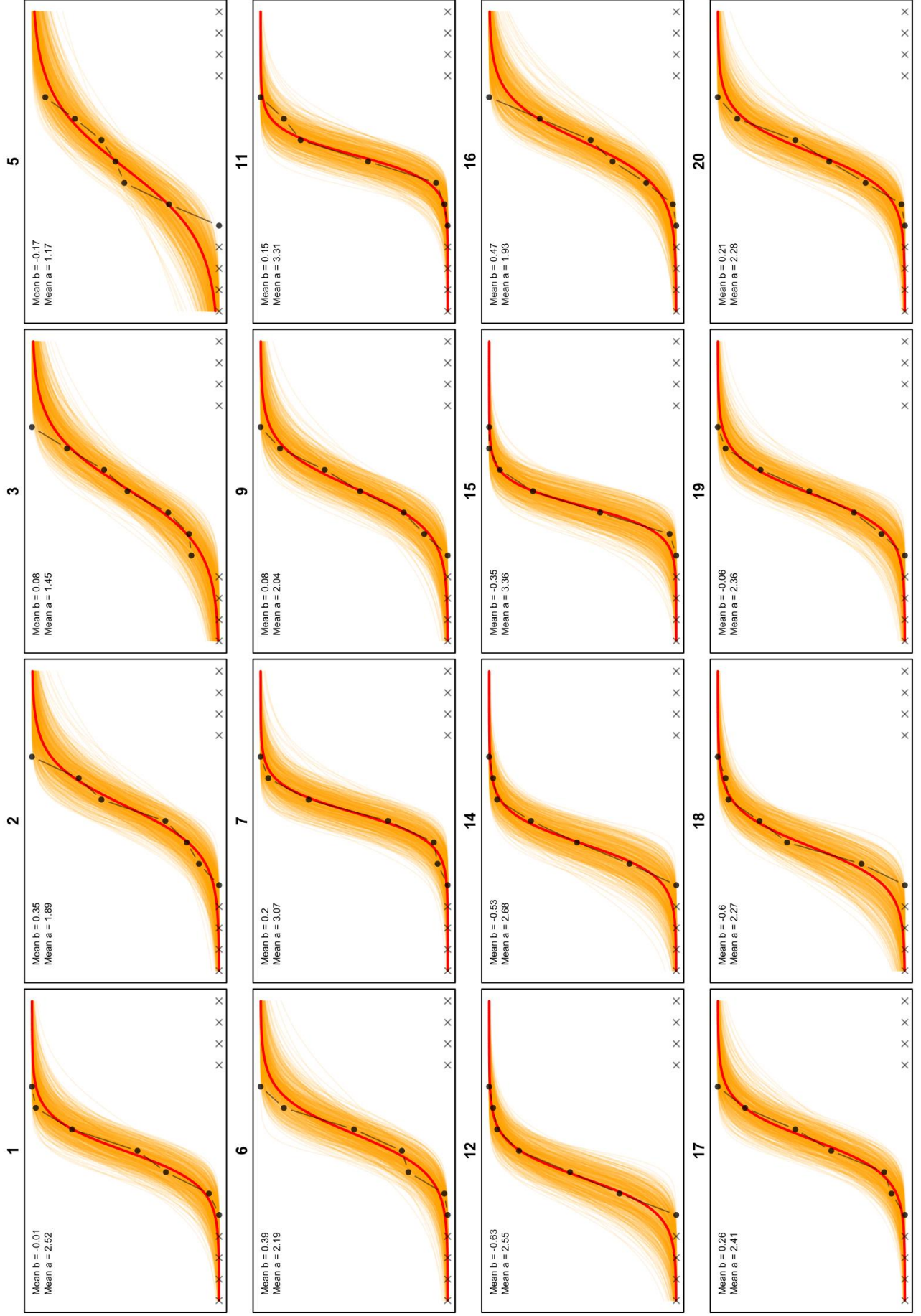
Items from the CBQ Perception of the Dyad Subscale (15)



Items from the Beck Hopelessness Inventory (17)



Items from the Beck Hopelessness Inventory (16)



The posterior estimates for the IRT model item parameters are summarised in the following tables. The estimates are from the final measurement models, without the excluded items:

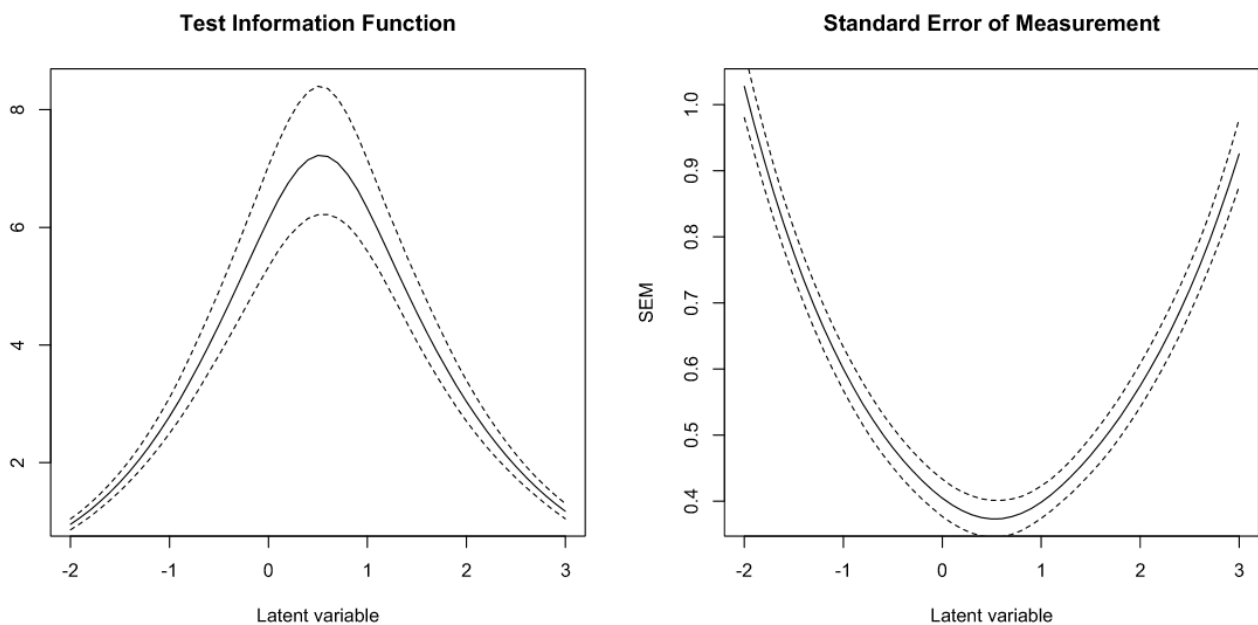
Conflict Behavior Questionnaire

Parameter	Mean	SD	66% CI	90% CI	ESS Bulk	ESS Tail
Item discrimination parameters						
<i>Item 3</i>	1.28	0.13	1.15 ; 1.40	1.07 ; 1.51	3541	3101
<i>Item 6</i>	1.40	0.15	1.26 ; 1.54	1.17 ; 1.64	2859	2831
<i>Item 12</i>	1.38	0.15	1.24 ; 1.52	1.15 ; 1.63	3083	2775
<i>Item 15</i>	1.93	0.19	1.75 ; 2.11	1.63 ; 2.25	2867	2987
<i>Item 24</i>	1.25	0.13	1.13 ; 1.37	1.04 ; 1.47	3640	2973
<i>Item 27</i>	1.46	0.16	1.30 ; 1.61	1.20 ; 1.74	3169	2912
<i>Item 30</i>	1.36	0.14	1.22 ; 1.49	1.14 ; 1.59	3061	3215
<i>Item 33</i>	1.56	0.16	1.41 ; 1.71	1.31 ; 1.84	3113	3147
<i>Item 36</i>	1.29	0.14	1.16 ; 1.42	1.07 ; 1.53	3661	2716
<i>Item 45</i>	1.63	0.17	1.47 ; 1.78	1.36 ; 1.91	3300	3693
<i>Item 51</i>	2.77	0.29	2.50 ; 3.04	2.33 ; 3.28	2256	2973
<i>Item 54</i>	1.54	0.18	1.36 ; 1.71	1.24 ; 1.84	3801	2680
<i>Item 57</i>	1.19	0.14	1.06 ; 1.33	0.97 ; 1.43	3675	2982
<i>Item 60</i>	1.25	0.13	1.12 ; 1.37	1.04 ; 1.47	3353	3298
<i>Item 63</i>	1.32	0.14	1.19 ; 1.45	1.10 ; 1.55	3477	3177
Item threshold parameters						
<i>Item 3</i>	0.62	0.09	0.53 ; 0.71	0.46 ; 0.78	2540	2532
<i>Item 6</i>	0.45	0.09	0.37 ; 0.53	0.31 ; 0.59	2904	3086
<i>Item 12</i>	-0.50	0.09	-0.58 ; -0.41	-0.64 ; -0.36	2214	2995
<i>Item 15</i>	0.29	0.07	0.23 ; 0.36	0.18 ; 0.41	2638	3101
<i>Item 24</i>	0.51	0.09	0.43 ; 0.60	0.37 ; 0.67	2593	3170
<i>Item 27</i>	1.30	0.13	1.18 ; 1.42	1.10 ; 1.52	2666	3266
<i>Item 30</i>	0.15	0.08	0.08 ; 0.23	0.02 ; 0.29	2556	3252
<i>Item 33</i>	-0.18	0.08	-0.25 ; -0.10	-0.31 ; -0.05	2638	2883
<i>Item 36</i>	0.77	0.10	0.68 ; 0.87	0.61 ; 0.95	2844	3342
<i>Item 45</i>	0.95	0.10	0.85 ; 1.04	0.79 ; 1.12	2427	3318
<i>Item 51</i>	0.47	0.07	0.41 ; 0.54	0.36 ; 0.59	2324	3163
<i>Item 54</i>	1.60	0.15	1.46 ; 1.74	1.36 ; 1.86	2711	2771
<i>Item 57</i>	1.37	0.15	1.23 ; 1.51	1.14 ; 1.64	3057	3339
<i>Item 60</i>	-0.38	0.09	-0.47 ; -0.30	-0.53 ; -0.24	2482	3123
<i>Item 63</i>	0.04	0.08	-0.03 ; 0.12	-0.09 ; 0.17	2795	2737

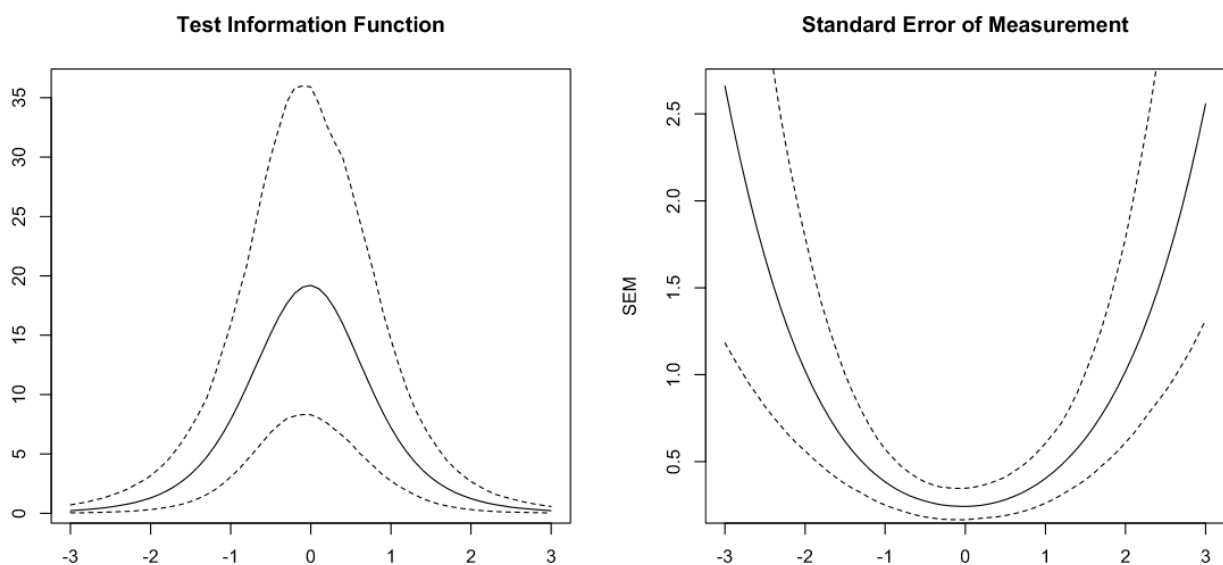
Beck Hopelessness Inventory

Parameter	Mean	SD	66% CI	90% CI	ESS Bulk	ESS Tail
Item discrimination parameters						
<i>Item 1</i>	2.47	0.57	1.92 ; 3.00	1.63 ; 3.48	669	1192
<i>Item 2</i>	1.89	0.44	1.46 ; 2.30	1.23 ; 2.67	714	1363
<i>Item 3</i>	1.41	0.34	1.09 ; 1.73	0.89 ; 2.02	689	1414
<i>Item 5</i>	1.15	0.29	0.88 ; 1.42	0.72 ; 1.66	805	1496
<i>Item 6</i>	2.17	0.51	1.69 ; 2.64	1.43 ; 3.09	634	1174
<i>Item 7</i>	3.10	0.71	2.43 ; 3.77	2.04 ; 4.34	591	1519
<i>Item 9</i>	2.00	0.47	1.56 ; 2.43	1.30 ; 2.83	671	1511
<i>Item 11</i>	3.30	0.77	2.58 ; 4.01	2.16 ; 4.67	644	1437
<i>Item 12</i>	2.46	0.57	1.92 ; 3.00	1.63 ; 3.50	694	1342
<i>Item 14</i>	2.54	0.60	1.97 ; 3.11	1.68 ; 3.62	666	1354
<i>Item 15</i>	3.38	0.79	2.62 ; 4.12	2.21 ; 4.79	637	1285
<i>Item 16</i>	1.89	0.45	1.47 ; 2.31	1.25 ; 2.67	659	1358
<i>Item 17</i>	2.47	0.58	1.91 ; 3.01	1.62 ; 3.49	679	1361
<i>Item 18</i>	2.27	0.54	1.76 ; 2.76	1.50 ; 3.21	727	1289
<i>Item 19</i>	2.33	0.54	1.82 ; 2.84	1.54 ; 3.30	700	1568
<i>Item 20</i>	2.32	0.54	1.80 ; 2.82	1.52 ; 3.27	685	1528
Item threshold parameters						
<i>Item 1</i>	0.03	0.24	-0.20 ; 0.26	-0.36 ; 0.43	789	1580
<i>Item 2</i>	0.39	0.26	0.14 ; 0.64	-0.03 ; 0.82	670	1642
<i>Item 3</i>	0.13	0.25	-0.10 ; 0.38	-0.27 ; 0.56	802	1627
<i>Item 5</i>	-0.15	0.25	-0.39 ; 0.10	-0.56 ; 0.28	874	1671
<i>Item 6</i>	0.44	0.26	0.18 ; 0.70	0.03 ; 0.87	665	1578
<i>Item 7</i>	0.23	0.24	0.00 ; 0.47	-0.16 ; 0.63	721	1556
<i>Item 9</i>	0.13	0.24	-0.10 ; 0.36	-0.26 ; 0.53	763	1595
<i>Item 11</i>	0.20	0.24	-0.04 ; 0.43	-0.20 ; 0.59	732	1613
<i>Item 12</i>	-0.62	0.27	-0.87 ; -0.36	-1.09 ; -0.18	801	1371
<i>Item 14</i>	-0.51	0.26	-0.76 ; -0.26	-0.95 ; -0.09	796	1246
<i>Item 15</i>	-0.32	0.25	-0.55 ; -0.08	-0.72 ; 0.10	803	1358
<i>Item 16</i>	0.51	0.27	0.24 ; 0.76	0.08 ; 0.96	657	1697
<i>Item 17</i>	0.28	0.25	0.05 ; 0.53	-0.12 ; 0.70	690	1558
<i>Item 18</i>	-0.57	0.27	-0.82 ; -0.31	-1.02 ; -0.13	791	1314
<i>Item 19</i>	-0.02	0.24	-0.25 ; 0.21	-0.41 ; 0.38	770	1476
<i>Item 20</i>	0.23	0.24	0.00 ; 0.47	-0.17 ; 0.65	754	1465

As we use latent variables from IRT models rather than sum-scores in the analysis, conventional estimates of reliability (such as coefficient alpha) are not directly applicable. According to item response theory the reliability of trait estimates is not uniform across the range of the latent variable, but can be described by the test information function as long as the item response model fits the data (de Ayala, 2009). We computed and plotted the test information function for the final item sets of the CBQ and the BHS, which indicated acceptable reliability across the observed range of the latent variables.



Test information curve and plot of the standard error of measurement across the latent variable for the 15 items from the CBQ Perception of the Dyad subscale



Test information curve and plot of the standard error of measurement across the latent variable for the 16 items from the BHS

References

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