

irtgraph icc — Item characteristic curve plot

[Description](#)
[Options](#)
[Also see](#)

[Quick start](#)
[Remarks and examples](#)

[Menu](#)
[Stored results](#)

[Syntax](#)
[Reference](#)

Description

`irtgraph icc` plots item characteristic curves (ICCs) for binary items and category characteristic curves (CCCs) for categorical items for the currently fitted IRT model.

Quick start

IPL model for binary items `b1` to `b10`

```
irt 1pl b1-b10
```

Plot ICCs for all items

```
irtgraph icc
```

Plot ICCs and item difficulties for items `b1`, `b5`, and `b9`

```
irtgraph icc b1 b5 b9, blocation
```

GRM for ordinal items `o1` to `o5`, items coded 1, 2, 3

```
irt grm o1-o5
```

Plot CCCs for selected item categories

```
irtgraph icc 1.o1 3.o5 2.o1
```

Plot CCCs for the first category of all items

```
irtgraph icc 1.o*
```

Menu

Statistics > IRT (item response theory)

Syntax

Basic syntax

```
irtgraph icc [varlist] [, options]
```

Full syntax

```
irtgraph icc (varlist [, plot_options]) (varlist [, plot_options]) [...] [, options]
```

varlist is a list of items from the currently fitted IRT model.

<i>options</i>	Description
Plots	
<code>blocation</code> [(<i>line_options</i>)]	add vertical lines for estimated item difficulties
<code>plocation</code> [(<i>line_options</i>)]	add horizontal lines for midpoint probabilities
<code>bcc</code>	plot boundary characteristic curves for categorical items
<code>ccc</code>	plot category characteristic curves
<code>range</code> (# #)	plot over $\theta = \#$ to $\#$
Line	
<i>line_options</i>	affect rendition of the plotted curves
Add plots	
<code>addplot</code> (<i>plot</i>)	add other plots to the ICC plot
Y axis, X axis, Titles, Legend, Overall	
<i>twoway_options</i>	any options other than <code>by()</code> documented in [G-3] <i>twoway_options</i>
Data	
<code>n</code> (#)	evaluate curves at # points; default is <code>n(300)</code>
<code>data</code> (<i>filename</i> [, <code>replace</code>])	save plot data to a file

<i>plot_options</i>	Description
<code>blocation</code> [(<i>line_options</i>)]	add vertical lines for estimated item difficulties
<code>plocation</code> [(<i>line_options</i>)]	add horizontal lines for midpoint probabilities
<code>bcc</code>	plot boundary characteristic curves for categorical items
<code>ccc</code>	plot category characteristic curves
<i>line_options</i>	affect rendition of the plotted curves

varlist may use factor-variable notation; see [U] 11.4.3 **Factor variables**.

line_options in *plot_options* override the same options specified in *options*.

Options

Plots

`blocation`[(*line_options*)] specifies that for each ICC, a vertical line be drawn from the estimated difficulty parameter on the x axis to the curve. The optional *line_options* specify how the vertical lines are rendered; see [G-3] *line_options*. This option implies option `bcc`.

`plocation`[(*line_options*)] specifies that for each ICC, a horizontal line be drawn from the midpoint probability on the y axis to the curve. The optional *line_options* specify how the horizontal lines are rendered; see [G-3] *line_options*. This option implies option `bcc`.

`bcc` specifies that boundary characteristic curves (BCCs) be plotted for categorical items. The ICCs for the individual item categories are plotted by default. This option has no effect on binary items.

`ccc` specifies that category characteristic curves (CCCs) be plotted for all items. This is the default behavior for categorical items. For binary items, this option will plot ICCs for both outcomes.

`range`(# #) specifies the range of values for θ . This option requires a pair of numbers identifying the minimum and maximum. The default is `range(-4 4)` unless the estimated difficulty parameters exceed these values, in which case the range is extended.

Line

line_options affect the rendition of the plotted ICCs; see [G-3] *line_options*.

Add plots

`addplot`(*plot*) allows adding more `graph twoway` plots to the graph; see [G-3] *addplot_option*.

Y axis, X axis, Titles, Legend, Overall

twoway_options are any of the options documented in [G-3] *twoway_options*, excluding `by()`. These include options for titling the graph (see [G-3] *title_options*) and for saving the graph to disk (see [G-3] *saving_option*).

Data

`n`(#) specifies the number of points at which the ICCs, CCCs, and BCCs are to be evaluated. The default is `n(300)`.

`data`(*filename* [, `replace`]) saves the plot data to a Stata data file.

Remarks and examples

[stata.com](http://www.stata.com)

`irtgraph icc` plots ICCs for binary items and CCCs for categorical items after estimating the parameters of an IRT model using `irt`.

ICCs are also known as item response functions and item response curves.

CCCs are also known as category response functions, option response functions, operating characteristic curves, and category response curves.

For categorical items, `irtgraph icc` also plots BCCs, which are probability curves for crossing a boundary. BCCs are also known as “category boundary curves”.

`irtgraph icc` is very flexible, and the best way to learn its capabilities is through examples.

▷ Example 1: ICCs for binary outcomes

We continue with the model from [example 1](#) of [\[IRT\] irt 1pl](#). Recall that we fit a 1PL model to the nine binary items. Here we use `estat` report to rearrange the estimated IRT parameters sorted by item difficulty.

```
. use http://www.stata-press.com/data/r15/masc1
(Data from De Boeck & Wilson (2004))
```

```
. irt 1pl q1-q9
```

```
Fitting fixed-effects model:
```

```
Iteration 0: log likelihood = -4275.6606
```

```
Iteration 1: log likelihood = -4269.7861
```

```
Iteration 2: log likelihood = -4269.7825
```

```
Iteration 3: log likelihood = -4269.7825
```

```
Fitting full model:
```

```
Iteration 0: log likelihood = -4153.3609
```

```
Iteration 1: log likelihood = -4142.374
```

```
Iteration 2: log likelihood = -4142.3516
```

```
Iteration 3: log likelihood = -4142.3516
```

```
One-parameter logistic model
```

```
Number of obs = 800
```

```
Log likelihood = -4142.3516
```

		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
	Discrim	.852123	.0458445	18.59	0.000	.7622695	.9419765
q1	Diff	-.7071339	.1034574	-6.84	0.000	-.9099066	-.5043612
q2	Diff	-.1222008	.0963349	-1.27	0.205	-.3110138	.0666122
q3	Diff	-1.817693	.1399523	-12.99	0.000	-2.091994	-1.543391
q4	Diff	.3209596	.0976599	3.29	0.001	.1295498	.5123695
q5	Diff	1.652719	.1329494	12.43	0.000	1.392144	1.913295
q6	Diff	.6930617	.1031842	6.72	0.000	.4908243	.8952991
q7	Diff	1.325001	.1205805	10.99	0.000	1.088668	1.561335
q8	Diff	-2.413443	.1691832	-14.27	0.000	-2.745036	-2.08185
q9	Diff	-1.193206	.1162054	-10.27	0.000	-1.420965	-.965448

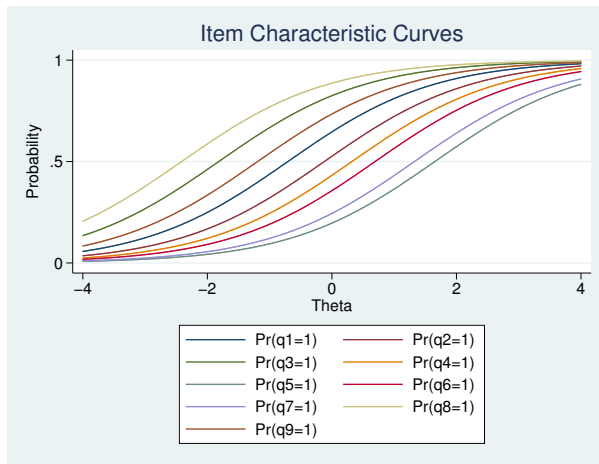
```
. estat report, sort(b) byparm
```

```
One-parameter logistic model          Number of obs   =          800
Log likelihood = -4142.3516
```

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Discrim	.852123	.0458445	18.59	0.000	.7622695	.9419765
Diff						
q8	-2.413443	.1691832	-14.27	0.000	-2.745036	-2.08185
q3	-1.817693	.1399523	-12.99	0.000	-2.091994	-1.543391
q9	-1.193206	.1162054	-10.27	0.000	-1.420965	-.965448
q1	-.7071339	.1034574	-6.84	0.000	-.9099066	-.5043612
q2	-.1222008	.0963349	-1.27	0.205	-.3110138	.0666122
q4	.3209596	.0976599	3.29	0.001	.1295498	.5123695
q6	.6930617	.1031842	6.72	0.000	.4908243	.8952991
q7	1.325001	.1205805	10.99	0.000	1.088668	1.561335
q5	1.652719	.1329494	12.43	0.000	1.392144	1.913295

To plot ICCs for all items in the model, we simply type

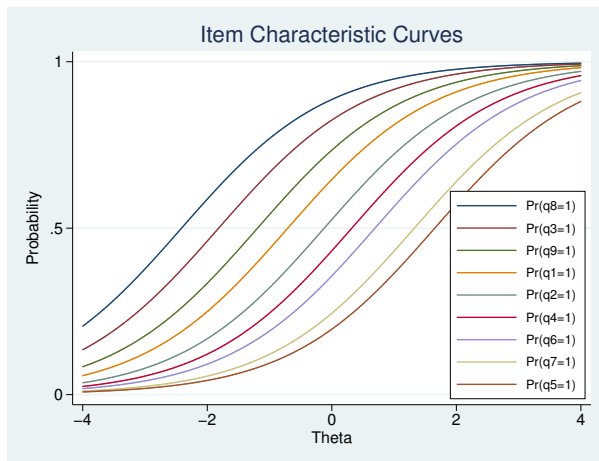
```
. irtgraph icc
```



With many items, the default graph is not pretty. The legend takes up too much space. One way to make the graph more readable is to shrink the legend and move it inside the plot region. We also specify a list of items explicitly so that the legend lists the ICCs in the order they appear in the graph. At the end of our interactive session, we came up with the following.

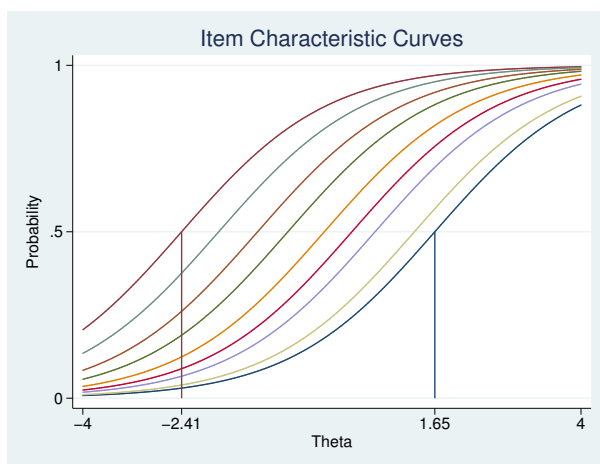
6 irtgraph icc — Item characteristic curve plot

```
. irtgraph icc q8 q3 q9 q1 q2 q4 q6 q7 q5,  
> legend(pos(4) col(1) ring(0) size(small))
```



Another way to present the ICCs is to turn off the legend and highlight the items with the lowest and highest estimated difficulty parameter. From the output of `estat report`, we see that item q8 has the lowest estimated difficulty parameter and item q5 has the highest estimated difficulty parameter. Here we put those items in a separate plotting group, request their estimated difficulty locations be plotted, and put the remaining items in another plotting group.

```
. irtgraph icc (q5 q8, blocation) (q1-q4 q6 q7 q9), legend(off)
```

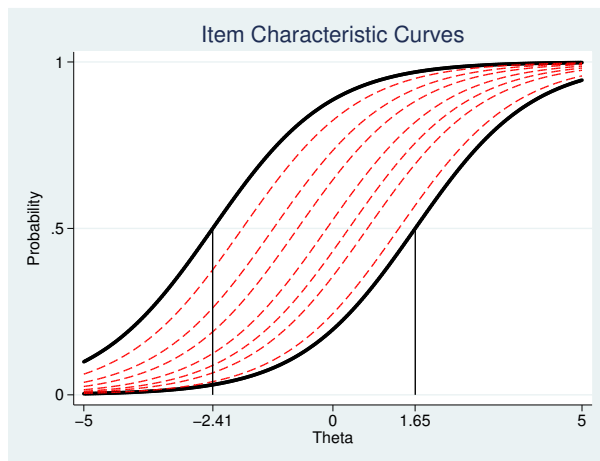


This plot shows the information we wanted, but we can tinker further to make items q5 and q8 stand out and make the whole plot more dramatic. Here is the final result.

```

. irtgraph icc
> (q8 q5, lcolor(black) lwidth(thick) bloc(lcolor(black)))
> (q1-q4 q6 q7 q9, lpattern(dash)),
> range(-5 5) xlabel(-5 -2.41 0 1.65 5) legend(off) lcolor(red)

```



We admit the above works nicely for a 1PL model because the ICCs do not cross; 2PL and 3PL models may require a different approach, but the general idea remains the same—we rarely obtain the desired ICC plot on the first try and need to work incrementally to arrive at the graph that best suits the estimated model parameters.

◀

□ Technical note

For a binary item, it is standard practice to plot only the ICC for the probability of the positive outcome. Thus the following commands are equivalent.

```

. irtgraph icc q1
. irtgraph icc 1.q1

```

However, there are in fact two ICCs we could plot: one for the probability of the positive outcome and one for the probability of the negative outcome. To plot both ICCs, we can use any of the following:

```

. irtgraph icc 0.q1 1.q1
. irtgraph icc i.q1
. irtgraph icc q1, ccc

```

Because the two probabilities sum to 1, the ICC for the negative outcome is a mirror image of the ICC for the positive outcome reflected about the y axis at 0.5.

□

▷ Example 2: CCCs for categorical outcomes

We continue with the model introduced in [example 1](#) of [\[IRT\] irt grm](#). To easily present some graphical features, we collapse the last two categories into one for all items and refit the GRM.

8 *irtgraph icc* — Item characteristic curve plot

```
. use http://www.stata-press.com/data/r15/charity
(Data from Zheng & Rabe-Hesketh (2007))
. recode ta1-ta5 (3=2)
(ta1: 58 changes made)
(ta2: 102 changes made)
(ta3: 55 changes made)
(ta4: 36 changes made)
(ta5: 86 changes made)
. irt grm ta1-ta5
```

Fitting fixed-effects model:

```
Iteration 0: log likelihood = -4887.6873
Iteration 1: log likelihood = -4736.7618
Iteration 2: log likelihood = -4713.4582
Iteration 3: log likelihood = -4713.2515
Iteration 4: log likelihood = -4713.2513
```

Fitting full model:

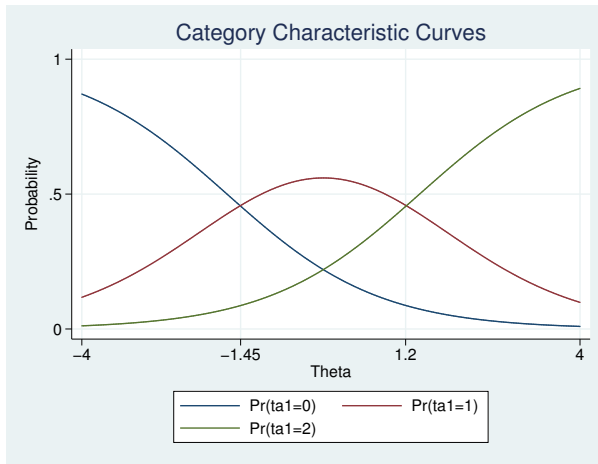
```
Iteration 0: log likelihood = -4519.7175
Iteration 1: log likelihood = -4434.2749
Iteration 2: log likelihood = -4430.3549
Iteration 3: log likelihood = -4430.343
Iteration 4: log likelihood = -4430.343
```

```
Graded response model                               Number of obs   =           945
Log likelihood = -4430.343
```

		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
ta1	Discrim	.8184045	.0951662	8.60	0.000	.6318823	1.004927
	Diff						
	>=1	-1.668228	.1931402	-8.64	0.000	-2.046775	-1.28968
	=2	1.423835	.1696051	8.39	0.000	1.091415	1.756255
ta2	Discrim	1.031415	.1119233	9.22	0.000	.8120498	1.250781
	Diff						
	>=1	-1.563834	.1578408	-9.91	0.000	-1.873196	-1.254472
	=2	-.007408	.077598	-0.10	0.924	-.1594973	.1446814
ta3	Discrim	1.595058	.1519538	10.50	0.000	1.297234	1.892882
	Diff						
	>=1	-1.128531	.09129	-12.36	0.000	-1.307457	-.9496062
	=2	1.082083	.0889538	12.16	0.000	.9077368	1.256429
ta4	Discrim	1.814543	.1837944	9.87	0.000	1.454313	2.174773
	Diff						
	>=1	-.3493225	.0601007	-5.81	0.000	-.4671178	-.2315273
	=2	1.527393	.1080719	14.13	0.000	1.315576	1.73921
ta5	Discrim	1.528601	.1497992	10.20	0.000	1.235	1.822202
	Diff						
	>=1	-.825658	.0811847	-10.17	0.000	-.9847772	-.6665389
	=2	.6543883	.0739536	8.85	0.000	.509442	.7993347

For a model with many categorical items, we do not recommend using *irtgraph icc* without *varlist*, because the resulting graph will contain far too many plotted curves. With 5 items, each with 3 categories, the total number of CCCs in the default plot is 15. Here we focus on item *ta1*.

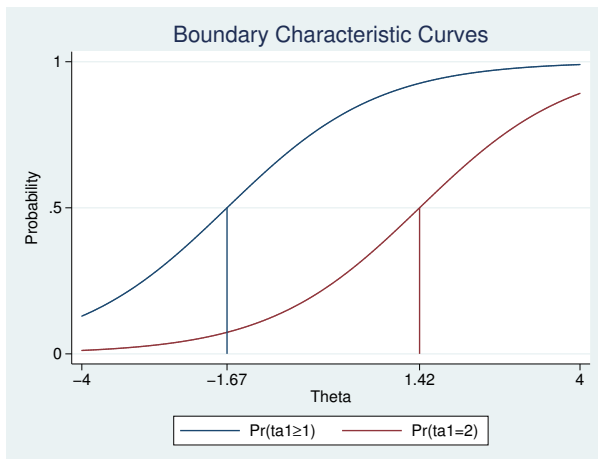

```
. irtgraph icc ta1, xlabel(-4 -1.45 1.20 4, grid)
```



In a GRM, the adjacent probability curves do not cross at the estimated difficulty parameters. Each crossing point represents the level of the latent trait at which an examinee “transitions” from responding in one category versus the next. Thus, in the graph above, respondents whose trait level is below approximately -1.45 are most likely to answer 0 (strongly agree); respondents whose trait level is between approximately -1.45 and 1.20 are most likely to answer 1 (somewhat agree); and respondents whose trait level is above approximately 1.20 are most likely to answer 2 (somewhat or strongly disagree).

Because the GRM is defined in terms of cumulative probabilities, the estimated difficulties represent a point at which a person with $\theta = b_{ik}$ has a 50% chance of responding in category k or higher. We can use `irtgraph` to plot these probabilities with the corresponding estimated category difficulties. These probability curves are known as BCCs. We specify option `blocation`, which plots the category difficulties and also implies option `bcc`.

```
. irtgraph icc ta1, blocation
```



□ Technical note

In the example above, we typed

```
. irtgraph icc ta1
```

to plot the CCCs for item `ta1`. Because item `ta1` is coded 0, 1, or 2, we could have typed

```
. irtgraph icc 0.ta1 1.ta1 2.ta1
```

or

```
. irtgraph icc i.ta1
```

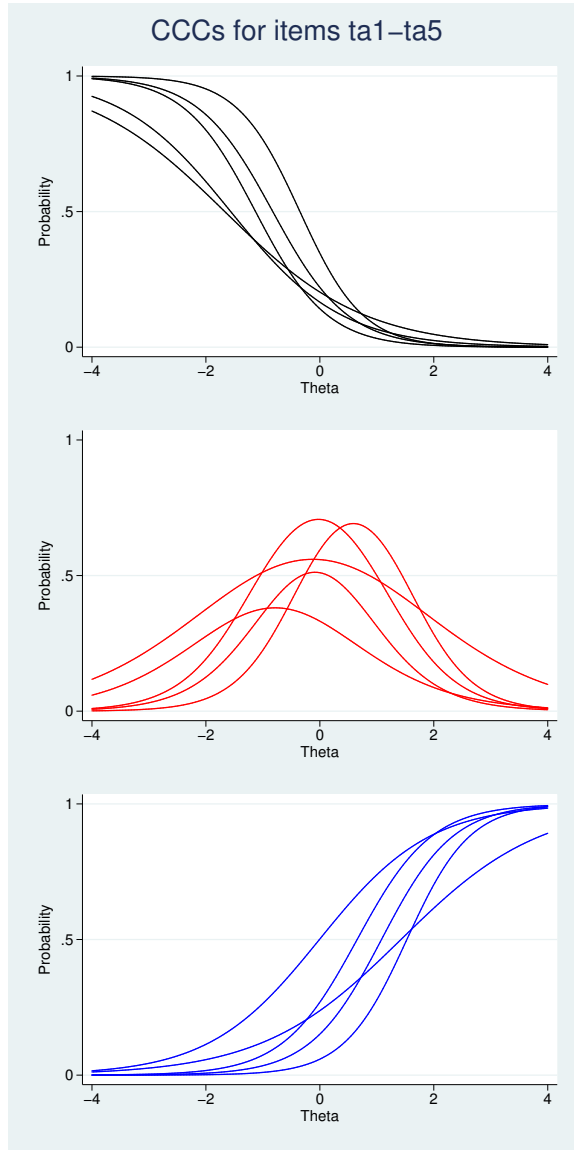
However, the first notation is most convenient to type. The factor notation comes in handy when we want to plot a particular category or change its appearance in the graph.

□

▷ Example 3: Combining graphs

Sometimes, it is useful to focus on one category and plot its CCCs for all items. Below we show one way of presenting this information. We do not label the curves, because we want to see only the overall shape and location of the CCCs. We could always play with the legend to identify the individual curves, as we did above in [example 1](#).

```
. irtgraph icc 0.ta*, legend(off) title("") lcolor(black) nodraw
> name(out0,replace)
. irtgraph icc 1.ta*, legend(off) title("") lcolor(red) nodraw
> name(out1,replace)
. irtgraph icc 2.ta*, legend(off) title("") lcolor(blue) nodraw
> name(out2,replace)
. graph combine out0 out1 out2, col(1) xsize(3) ysize(6)
> title("CCCs for items ta1-ta5")
```



4

Stored results

`irtgraph icc` stores the following in `r()`:

Macros

`r(xvals)`

values used to label the x axis

`r(yvals)`

values used to label the y axis

Reference

Raciborski, R. 2015. Spotlight on irt. *The Stata Blog: Not Elsewhere Classified*.
<http://blog.stata.com/2015/07/31/spotlight-on-irt/>.

Also see

- [IRT] **irt** — Introduction to IRT models
- [IRT] **irt 1pl** — One-parameter logistic model
- [IRT] **irt 2pl** — Two-parameter logistic model
- [IRT] **irt 3pl** — Three-parameter logistic model
- [IRT] **irt grm** — Graded response model
- [IRT] **irt hybrid** — Hybrid IRT models
- [IRT] **irt nrm** — Nominal response model
- [IRT] **irt pcm** — Partial credit model
- [IRT] **irt rsm** — Rating scale model
- [IRT] **irtgraph tcc** — Test characteristic curve plot