



Agenda

- ◊ A few more IRT concepts:
 - ◊ Local independence
 - ◊ Item parameter invariance
 - ◊ Information
 - ◊ Linking IRT and CTT
 - ◊ Polytomous models

Local Independence

- o (most) IRT models require that the test be unidimensional – or **sufficiently unidimensional**.
 - o Judgment call – a multidimensional test with a large general factor may be “sufficiently unidimensional,” or you can divide a multidimensional test into unidimensional subtests.
- o If our test is really unidimensional, the items are **locally independent**.
 - o The latent trait accounts for **all** of the observed relationship between any two item responses.
 - o At a given level of θ , the items are **independent**.
 - o I.e., uncorrelated!

Local Independence

- o How does that work?
 - o Key words: **at a given level of θ** .
 - o If we don't know θ , we expect the items to be correlated!
 - o Items are only independent *locally* – that is, at a particular value of θ .
- o This is analogous (conceptually) to the idea of item uniquenesses being uncorrelated with factor scores and with each other.
 - o All of the common variance is accounted for by the common factor (or latent trait); any other variance in item responses is unique.
- o Need to test local independence – if it does not hold, there may be another latent trait affecting item responses.

Item Parameter Invariance

- In CTT, all of our parameter estimates are linked to our specific sample of test-takers and items.
 - E.g., item-total correlations or factor loadings depend on the other items in the test.
 - Item difficulty depends on the group ability level.
- In IRT, that's not the case.
 - We estimate item parameters for each item individually based on θ estimates.
 - So item parameters should be **invariant** across different combinations of items and populations of test-takers.
 - We do get the θ estimates from the items (it's an iterative process), but they also aren't linked to a specific set of items.
 - We can *equate θ metrics* across groups of examinees.

Estimating θ

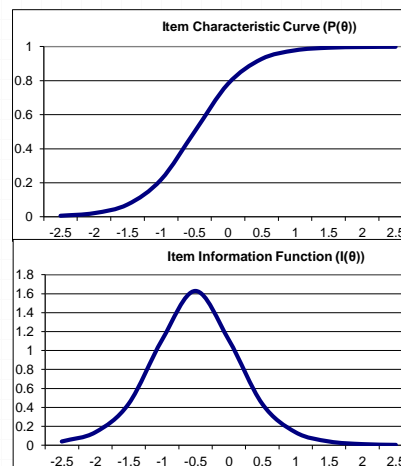
- We can estimate item parameters easily if we know participants' θ values...
 - And we can estimate θ if we know the item parameters!
- So how do we start?
 - Choose arbitrary start values for ability parameters.
 - Use these values to estimate the item parameters; then use these item parameters to estimate the ability parameters.
 - Keep going back and forth until these estimates converge.
- If we use **maximum likelihood estimation**, lots of items, and lots of participants, we can usually get good estimates this way.
 - One reason why IRT is so data-intensive!

Information

- Obviously, we want items that give us a lot of information.
 - Conceptually, information is the opposite of uncertainty.
- In IRT, information has a specific technical meaning:
 - The inverse (reciprocal) of the squared standard error of θ .
- But in IRT, this value is **not the same** at all values of θ !
 - It's an **information function** or curve.
 - Calculus-based, not a straightforward hand calculation.
 - $I(\theta)$ is highest at b – remember that we measure best around the item difficulty parameter.

Interpreting Information Curves

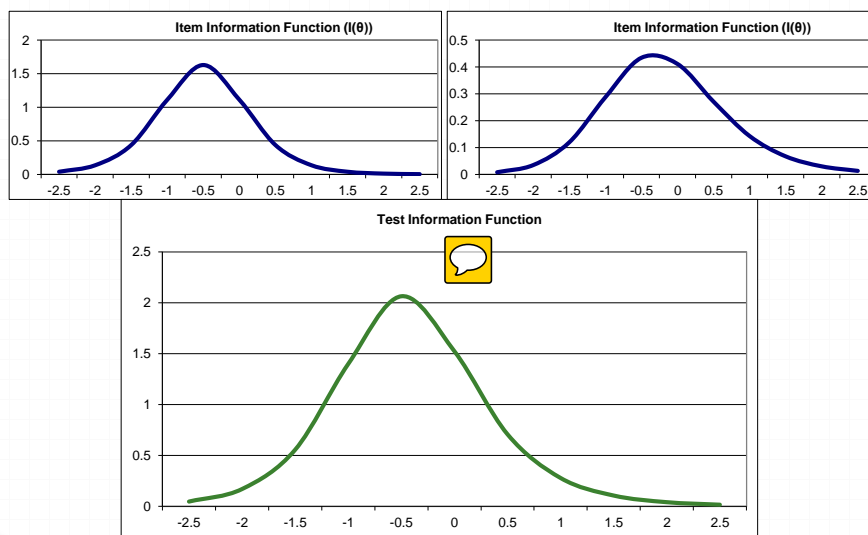
- IRT software will estimate these.
 - $1/SE(\theta)^2$ for lots of different values of θ .
- Shows you not only *how* precise your θ estimates are, but *where* they are precise.



Test Information Curves

- We can add the item information curves together to get a **test information function**.
- Information for one item is **independent** of information for the others!
- Why is this useful?
 - We can identify the contribution of each item to the total information in the test.
 - We can choose items to cover the range of θ we are concerned about. No single item measures across the construct; but we can choose the combination we want.

Test Information Curves



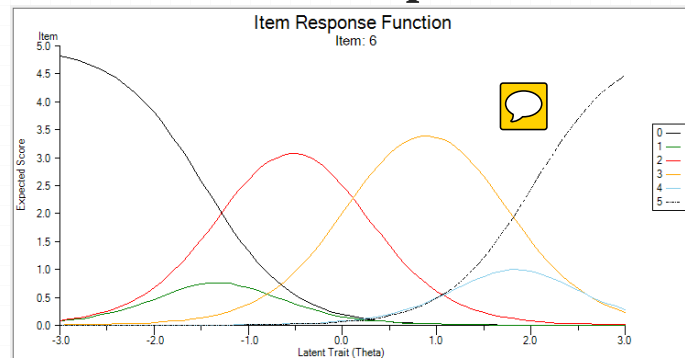
Linking CFA and IRT

- We've talked about IRT as an alternative to CFA...
- ... but CFA models can be **reparameterized** to be equivalent to IRT models!
 - Both are fundamentally concerned with estimating the underlying latent construct.
 - See McDonald (1999) for details.
- This work is the basis for some modern IRT approaches.
 - E.g., multidimensional IRT models.

Polytomous Models

- What to do when you have more than 2 response options?
 - **Polytomous IRT models**
- Again, lots of choices:
 - Graded response model
 - Generalized partial credit model
 - Partial credit model
 - Rating scale model
 - (ordered from most flexible to most restrictive).

Graded Response Model Example



- The item has one overall discrimination parameter **and** a difficulty parameter for every “threshold” (# of response options – 1).

What Can You Do With a Polytomous Model?

- Use more of the information in rating-scale type items!
 - Not dichotomizing – but note the response is still seen as a categorical variable.
- Get some insight into the response scale.
 - Are the responses really ordered as we think they should be?
- Estimate **ideal point** models.
 - When you can have too much of a good thing – e.g., conscientiousness.

Questions?

Please do course evals! (online)

For next time: Final exam

Read: Look over the final exam, posted later today!

Lab Friday: Intro to IRT in Mplus