

An IRT analysis of a verbal aggression data set

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While common, verbal aggression can be harmful

Others see verbal aggression as a negative character trait

Verbal aggression can cause mutual aggressive reactions and further escalation

People differ in the degree to which they are able to inhibit their verbally aggressive impulses

Verbal aggression data set

Collected at the University of Leuven, Belgium, by Kristof Vansteelandt

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Responses of 316 subjects (243 women and 73 men) to a questionnaire of 24 items, about verbal aggression

Items describe a frustrating situation together with a possible verbal aggression response

Verbal aggression data set

The subset we will consider has 8 items. Each item starts with one of the situations below:

- *A bus fails to stop for me*
- *I miss a train because a clerk gave me faulty information*
- *The grocery store closes just as I am about to enter*
- *The operator disconnects me when I had used up my last dime for a call*

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The corresponding item response is 1 if the respondent agrees with that sentence, and 0 if not

Formal description of the data subset

Item-side variables

- Scenario $\in \{1, 2, 3, 4\}$
- Behavior mode $\in \{\text{Do}, \text{Want}\}$

Person-side variable

- Gender: $\in \{M, F\}$

Dependent variable

- Response $\in \{0, 1\}$

Research questions

- **Item effect:** Do we see a difference between *wanting* and *doing*?
- **Person effect:** Do we see a gender difference in saying they would exhibit a verbally aggressive response?
- **Person-by-item interaction effect:** Do we see a gender difference in the difference between wanting and doing?

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A modern extension of the Rasch model considers more sophisticated views of the person and item parameters (De Boeck & Wilson, 2004)

Explanatory item response models

For example, the item parameter β_i can be re-expressed as $\beta_i = \gamma \times W_i$:

$$W_i = \begin{cases} 1/2 & \text{if } i \text{ is want-item} \\ -1/2 & \text{if } i \text{ is do-item} \end{cases}$$

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person part	item part	model type
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$\alpha G_p + \epsilon_p$	$-\gamma W_i$	doubly explanatory
$\alpha G_p + \epsilon_p - \gamma W_i - \delta W_i G_p$		interaction explanatory

Research questions

Do we see a difference between *wanting* and *doing*?

→ Need an item explanatory model

Do we see a gender difference in saying they would exhibit a verbally aggressive response?

→ Need a person explanatory model

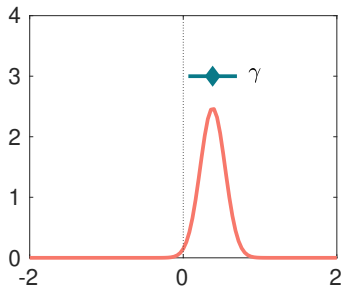
Do we see a gender difference in the difference between wanting and doing?

→ Need an interaction explanatory model

Item-side question (wanting, $W = \frac{1}{2}$, vs. doing, $W = -\frac{1}{2}$)

$$P(X_{ip}) = \text{ilogit}(\theta_p - \gamma W_i)$$

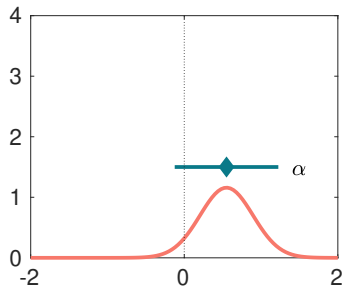
Parameter	Posterior			95% Cred. Int.	
	Mean	Median	SD	Lower	Upper
gamma	0.385	0.383	0.163	0.064	0.706



Person-side question (female, $G = -\frac{1}{2}$, vs. male, $G = \frac{1}{2}$)

$$P(X_{ip}) = \text{ilogit}(\alpha G_p + \epsilon_p - \beta_i)$$

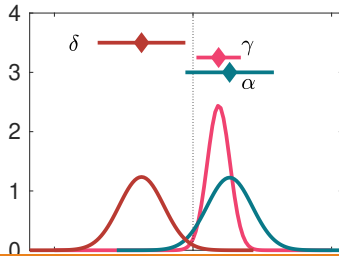
Parameter	Posterior			95% Cred. Int.	
	Mean	Median	SD	Lower	Upper
alpha	0.542	0.536	0.342	-0.116	1.228



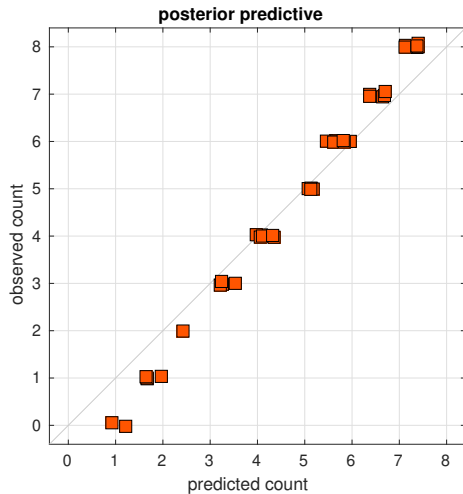
Interaction question (gender \times wanting vs. doing)

$$P(X_{ip}) = \text{ilogit}(\alpha G_p + \epsilon_p - \gamma W_i - \delta G_p W_i)$$

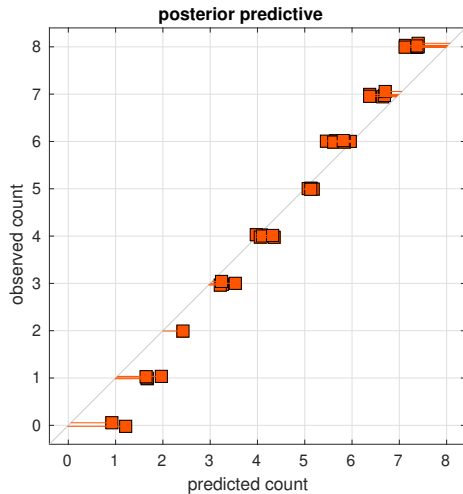
Parameter	Posterior			95% Cred. Int.	
	Mean	Median	SD	Lower	Upper
alpha	0.522	0.520	0.330	-0.133	1.172
delta	-0.736	-0.735	0.315	-1.365	-0.117
gamma	0.365	0.365	0.166	0.035	0.692



Posterior predictive



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Do we see a gender difference in saying they would exhibit a verbally aggressive response?

Do we see a gender difference in the difference between wanting and doing?

Do we see a difference between *wanting* and *doing*?

→ Yes, wanting is more commonly selected than doing

Do we see a gender difference in saying they would exhibit a verbally aggressive response?

Do we see a gender difference in the difference between wanting and doing?

Conclusions

Do we see a difference between *wanting* and *doing*?

→ Yes, wanting is more commonly selected than doing

Do we see a gender difference in saying they would exhibit a verbally aggressive response?

→ We don't see a main effect

Do we see a gender difference in the difference between wanting and doing?

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→ Yes, wanting is more commonly selected than doing

Do we see a gender difference in saying they would exhibit a verbally aggressive response?

→ We don't see a main effect

Do we see a gender difference in the difference between wanting and doing?

→ Yes, the difference is smaller among male respondents

References

De Boeck, P., & Wilson, M. (2004). *Explanatory item response models*. Wiley Online Library.

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Verbal aggression complete data set (part 1)

Item	Situations	Mode		Blame		Behavior		
		Want	Do	Other	Self	Curse	Scold	Shout
1	S1	1	0	1	0	1	0	0
2	S1	1	0	1	0	0	1	0
3	S1	1	0	1	0	0	0	1
4	S2	1	0	1	0	1	0	0
5	S2	1	0	1	0	0	1	0
6	S2	1	0	1	0	0	0	1
7	S3	1	0	0	1	1	0	0
8	S3	1	0	0	1	0	1	0
9	S3	1	0	0	1	0	0	1
10	S4	1	0	0	1	1	0	0
11	S4	1	0	0	1	0	1	0
12	S4	1	0	0	1	0	0	1

Verbal aggression complete data set (part 2)

Item	Situations	Mode		Blame		Behavior		
		Want	Do	Other	Self	Curse	Scold	Shout
13	S1	0	1	1	0	1	0	0
14	S1	0	1	1	0	0	1	0
15	S1	0	1	1	0	0	0	1
16	S2	0	1	1	0	1	0	0
17	S2	0	1	1	0	0	1	0
18	S2	0	1	1	0	0	0	1
19	S3	0	1	0	1	1	0	0
20	S3	0	1	0	1	0	1	0
21	S3	0	1	0	1	0	0	1
22	S4	0	1	0	1	1	0	0
23	S4	0	1	0	1	0	1	0
24	S4	0	1	0	1	0	0	1