

# Ordinal Models

PSY 504

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Princeton University

# What is an ordinal variable?

- A variable that consists of ordered, but discrete categories; there is a natural ordering to the categories
  - Examples:
    - Likert-scales
      - Strongly disagree, disagree, neutral, agree, strongly agree
      - Never, sometimes, often, always
    - Education level
      - High school degree, bachelor's degree, master's degree, etc.
  - By contrast, metric variables are continuous and inherently quantitative (e.g., score on a math test).

# Problems with treating ordinal data as metric

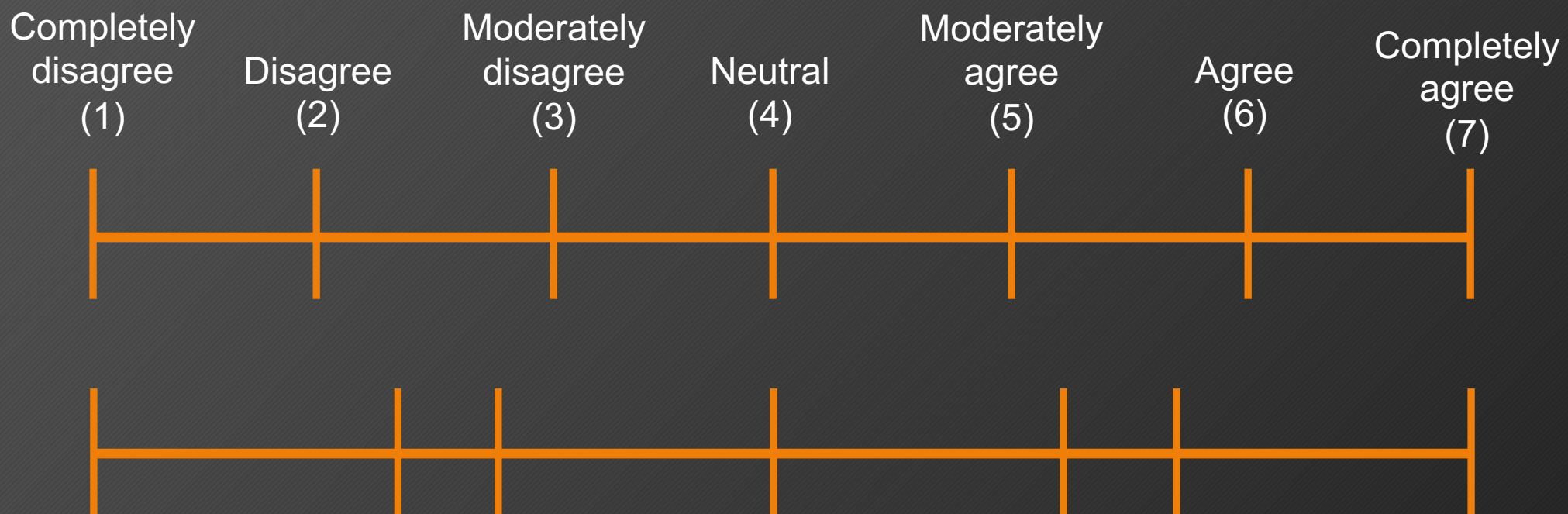
- Using models that assume data is metric (t-tests, ANOVA) for ordinal data, runs the risk of:
  - Low rates of correct detection
  - Distorted effect size estimates
  - Increased Type I error
  - Inversions of differences between groups



# Why these problems?



- Assumption of equidistance

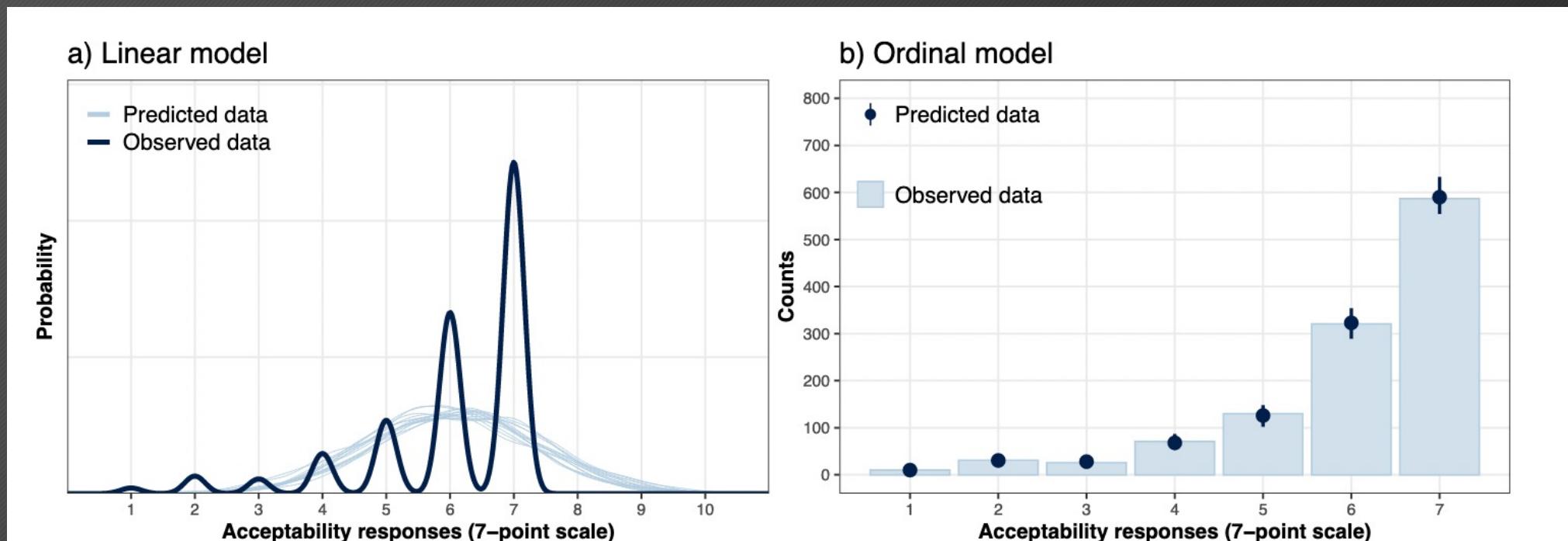


# Why these problems?



- Assumption of equidistance
- Distribution of ordinal variables may be non-normal
- Unequal variances of unobserved variables

# Treating ordinal data as continuous vs. ordinal



# LAW & ORDINAL MODELS



# What do ordinal models do?

- Predict discrete response categories
- Can handle non-normality
- Allow for unequal distances between responses

# Three classes of ordinal models

Cumulative models

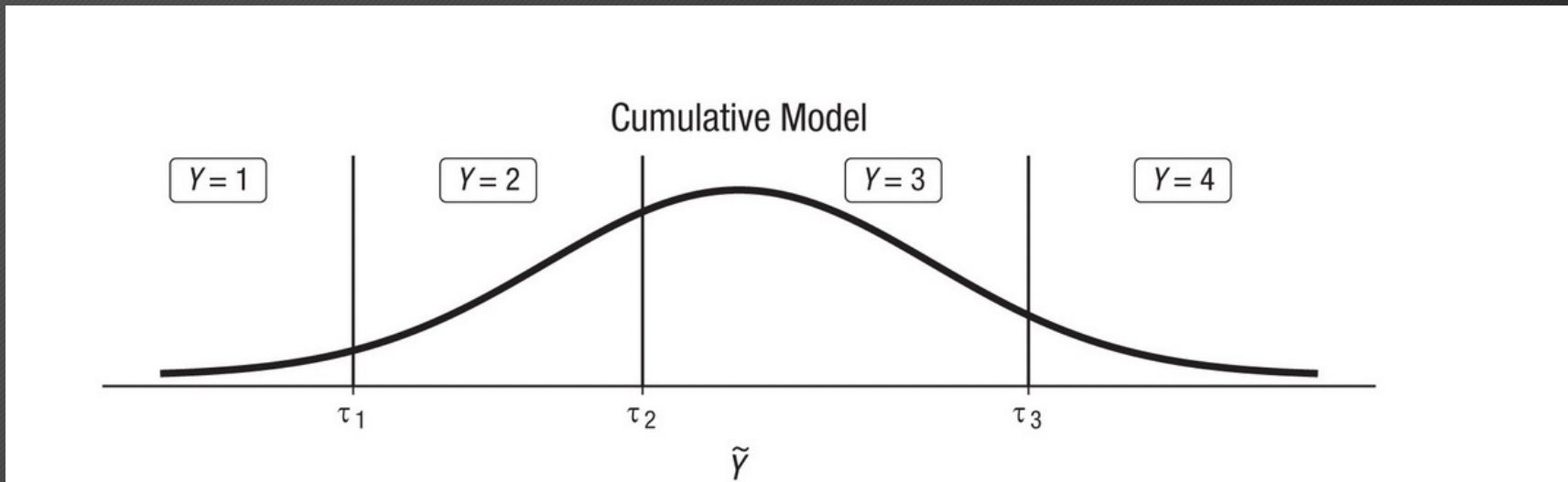
Sequential models

Adjacent-category models

# Cumulative model

- Good for Likert DVs, rating scales, etc.
- Assumes that observed ordinal variable  $Y$  originates from categorization of a latent (not observable) continuous variable  $\tilde{Y}$
- Assumes  $\tilde{Y}$  (the latent continuous variable) is normally distributed, with mean = 0 and SD = 1
- Assumes  $K$  thresholds ( $\tau_k$ ), which partition  $\tilde{Y}$  into  $K + 1$  observable and ordered categories of  $Y$

# Cumulative model

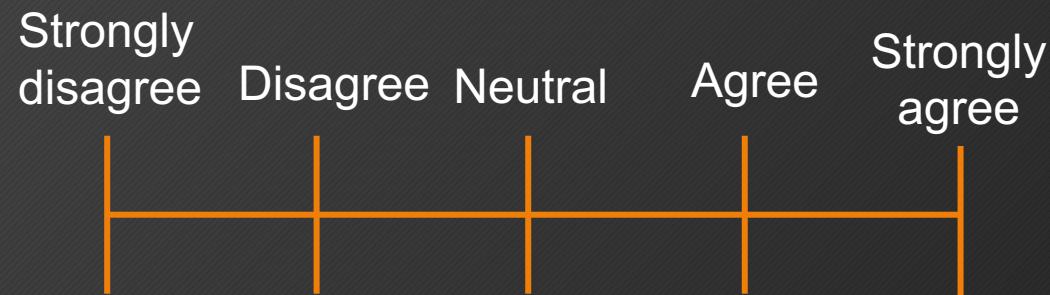


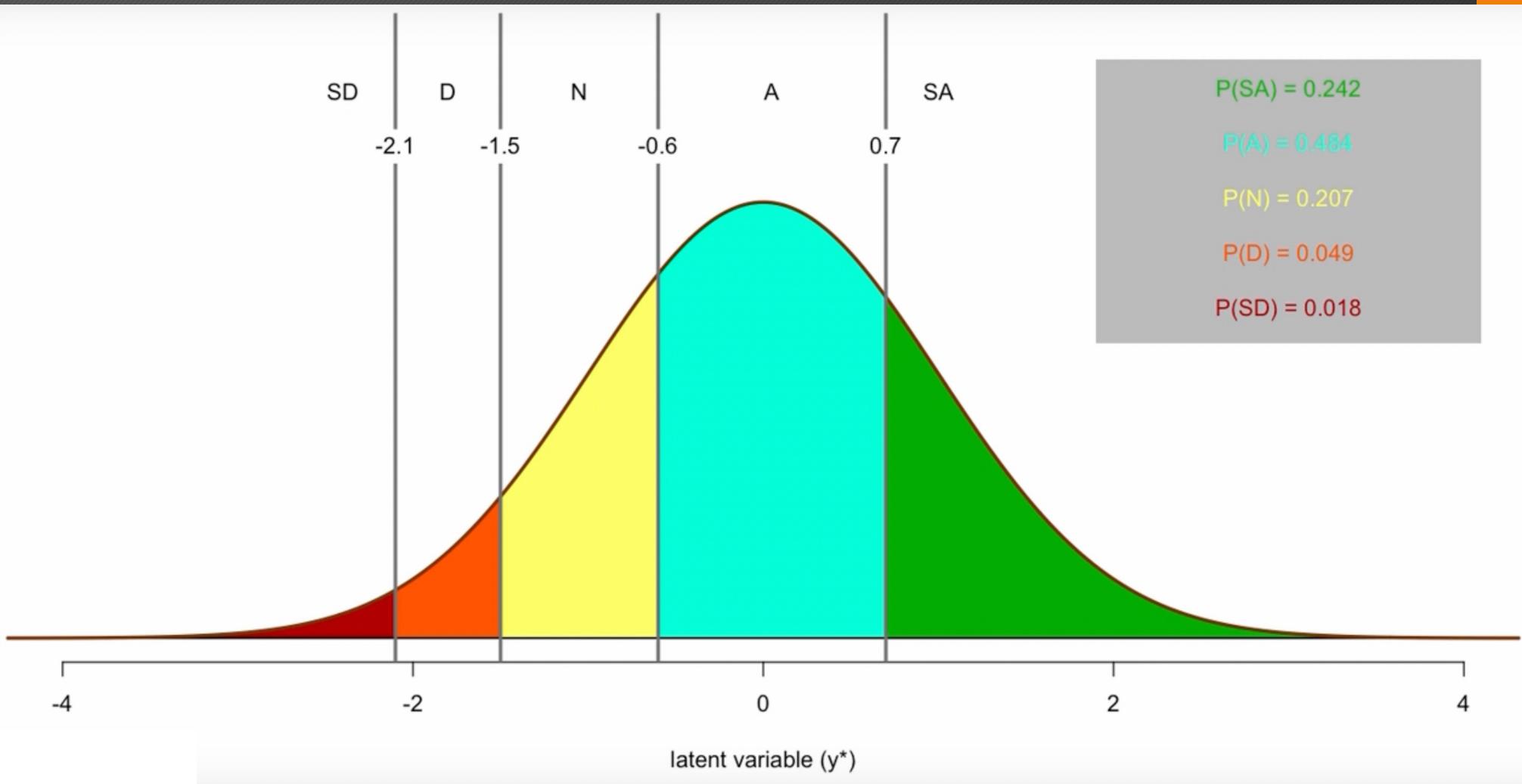
- Area under the curve in each bin represents the probability of the observed ordinal response  $Y$

# Example

**Please rate your agreement with the statement below.**

*Graphic design is clearly the passion of the person who created this image:*





# Sequential model

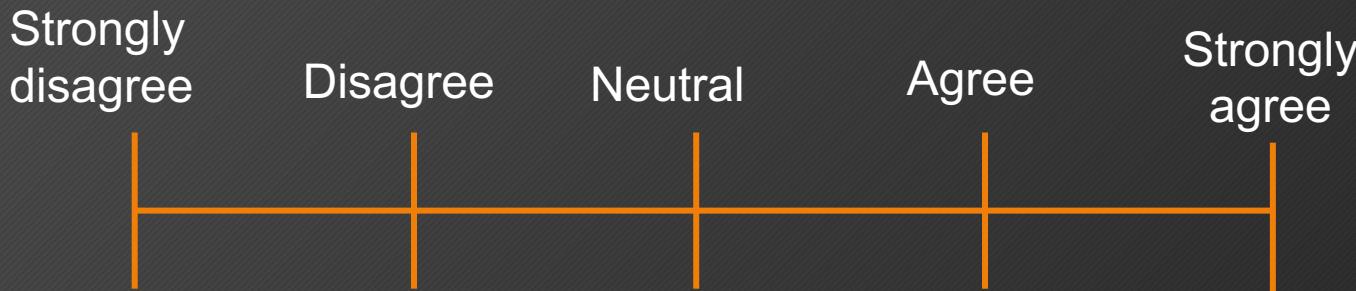
- Good for DVs resulting from a sequential process
- How do I know if it's a sequential process?
  - “When a higher response category is possible only after all lower categories are achieved” (Burkner & Vuorre, 2019)
  - E.g., years of marriage before divorce

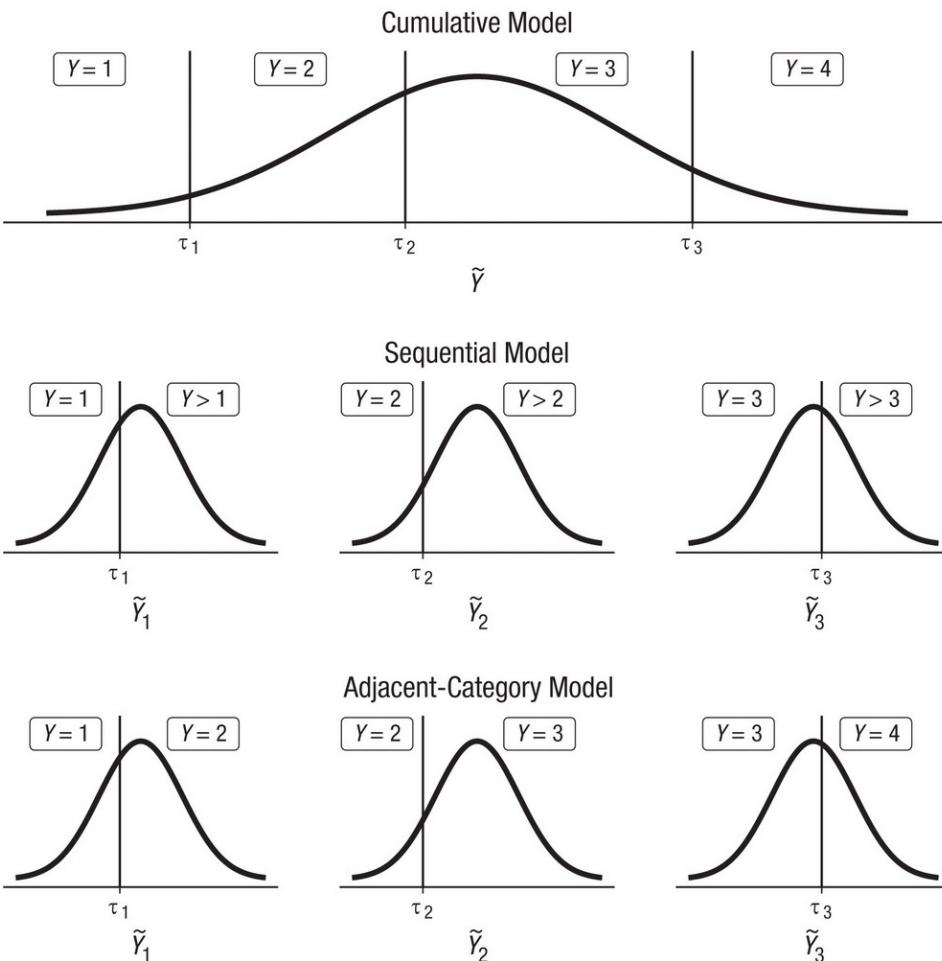
# Sequential model

- Assumes that for every category  $k$ , there is a latent continuous variable  $\tilde{Y}_k$  (standard normally distributed) that determines the transition between the  $k$ th and  $k + 1$ th category
- Thresholds ( $\tau_k$ ) separate categories.
- If  $\tilde{Y}_k > \tau_k$ , the sequential process continues.
- The probability that  $Y$  falls in category  $k$  equals the probability that it did not fall in one of the former categories 1 to  $k-1$ , multiplied by the probability the sequential process stopped at  $k$

# Adjacent-category model

- Models the decision between two adjacent categories of  $\tilde{Y}$
- Models the decision between  $Y = k$  and  $Y = k + 1$ 
  - Contrast this to sequential models, which model decision between  $Y = k$  and  $Y > k$
- Example: modeling the probability of choosing “agree” over “strongly agree”





# Generalizing across the three classes

- Category-specific effects
  - A predictor could have different effects on the different response categories of  $Y$ 
    - E.g., sympathy may affect people's ratings of "agree" vs. "neutral" in the graphic design example from Slide 12, but not people's ratings of "agree" vs. "strongly agree."
    - Sequential and adjacent-category models work best here.
- Unequal variances
  - When variance differs between groups, conditions, time, etc.
  - All three classes of models are good for this.

# Lingering questions

- Are there any cases in which differentiating between cumulative and sequential models becomes especially difficult?
- How do you know if some of your effects might be category-specific?

# References

- Bürkner, P. C., & Vuorre, M. (2019). Ordinal regression models in psychology: A tutorial. *Advances in Methods and Practices in Psychological Science*, 2(1), 77-101.
- Veríssimo, J. (2021). Analysis of rating scales: A pervasive problem in bilingualism research and a solution with Bayesian ordinal models. *Bilingualism: Language and Cognition*, 24(5), 842-848.
- <https://www.youtube.com/watch?v=jWIJ7P1G9P4>