

# Some Examples of Using Bayesian Statistics in Modeling Human Cognition

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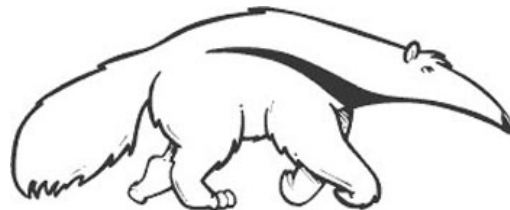
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<https://osf.io/fn6c2/>



# David Scott



# (Don't) Tell Me Another One



- Holds the record of **40 hours and 8 minutes** for the longest stand-up comedy show by an individual, at the Diamond Jo Casino in Dubuque, Iowa, in late April 2013
- He was allowed only a 5 minute break each hour, was required not to repeat any joke within a 4 hour period, and needed to have at least 10 people in the audience at all times.

# (Don't) Tell Me Another One

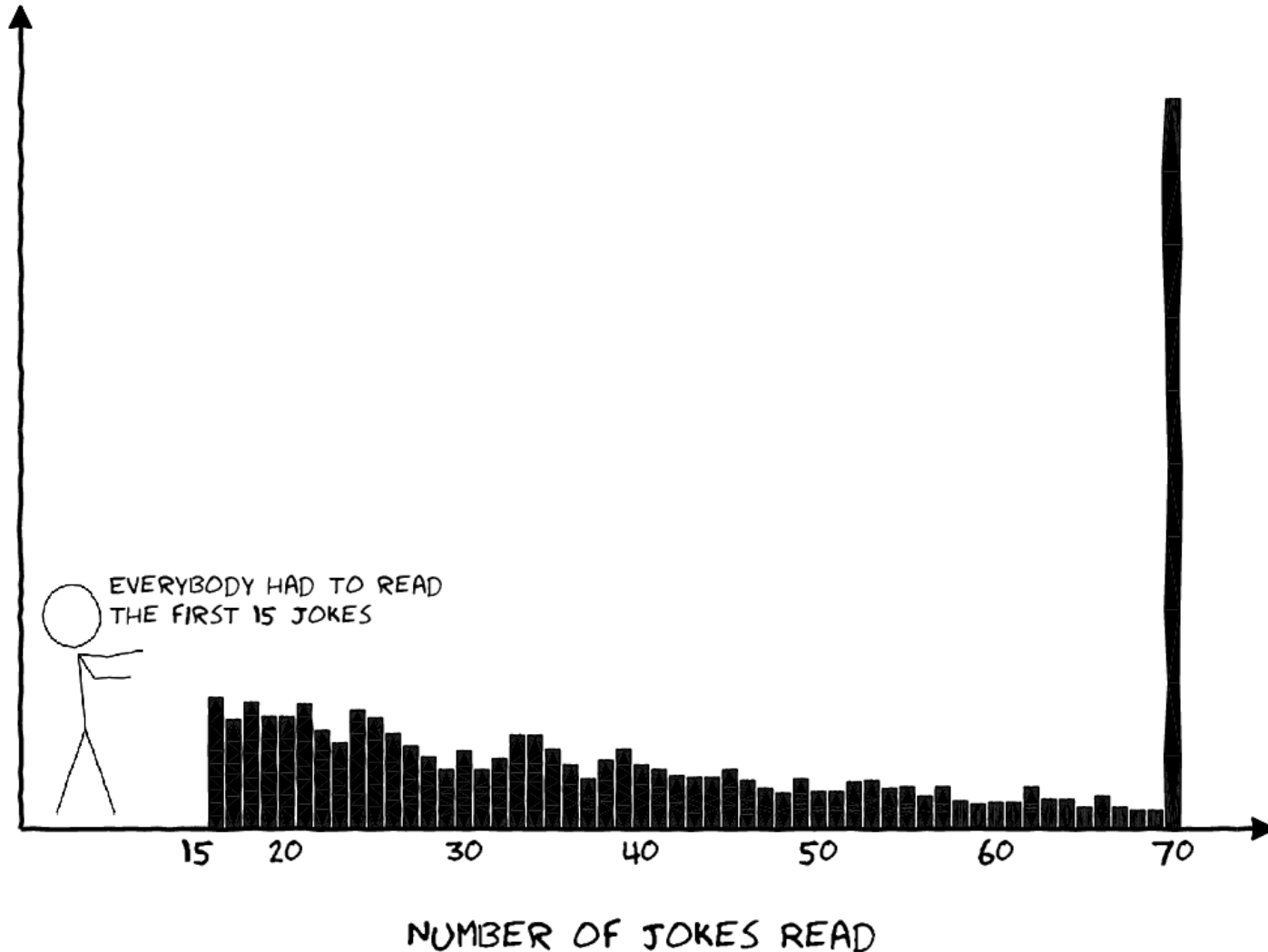


- One burning question for cognitive science is “when do people give up listening to jokes?”

- Holds the record of **40 hours and 8 minutes** for the longest stand-up comedy show by an individual, at the Diamond Jo Casino in Dubuque, Iowa, in late April 2013
- He was allowed only a 5 minute break each hour, was required not to repeat any joke within a 4 hour period, and needed to have at least 10 people in the audience at all times.

# Jester On-Line Joke System

- Distribution of jokes read by 2607 people, at a time when the system had 70 jokes (Goldberg et al, 2001)



# A Censored Geometric Model

- Suppose people have some probability of quitting after each joke, and the measurement of this geometric distribution is censored by the on-line joke system only having 70 jokes

$$\theta \sim \text{Uniform}(0, 1)$$

$$\alpha_i \sim 15 + \text{Geometric}(\theta)$$

$$y_i = \begin{cases} \alpha_i & \text{if } \alpha_i \leq 70 \\ 70 & \text{if } \alpha_i > 70. \end{cases}$$

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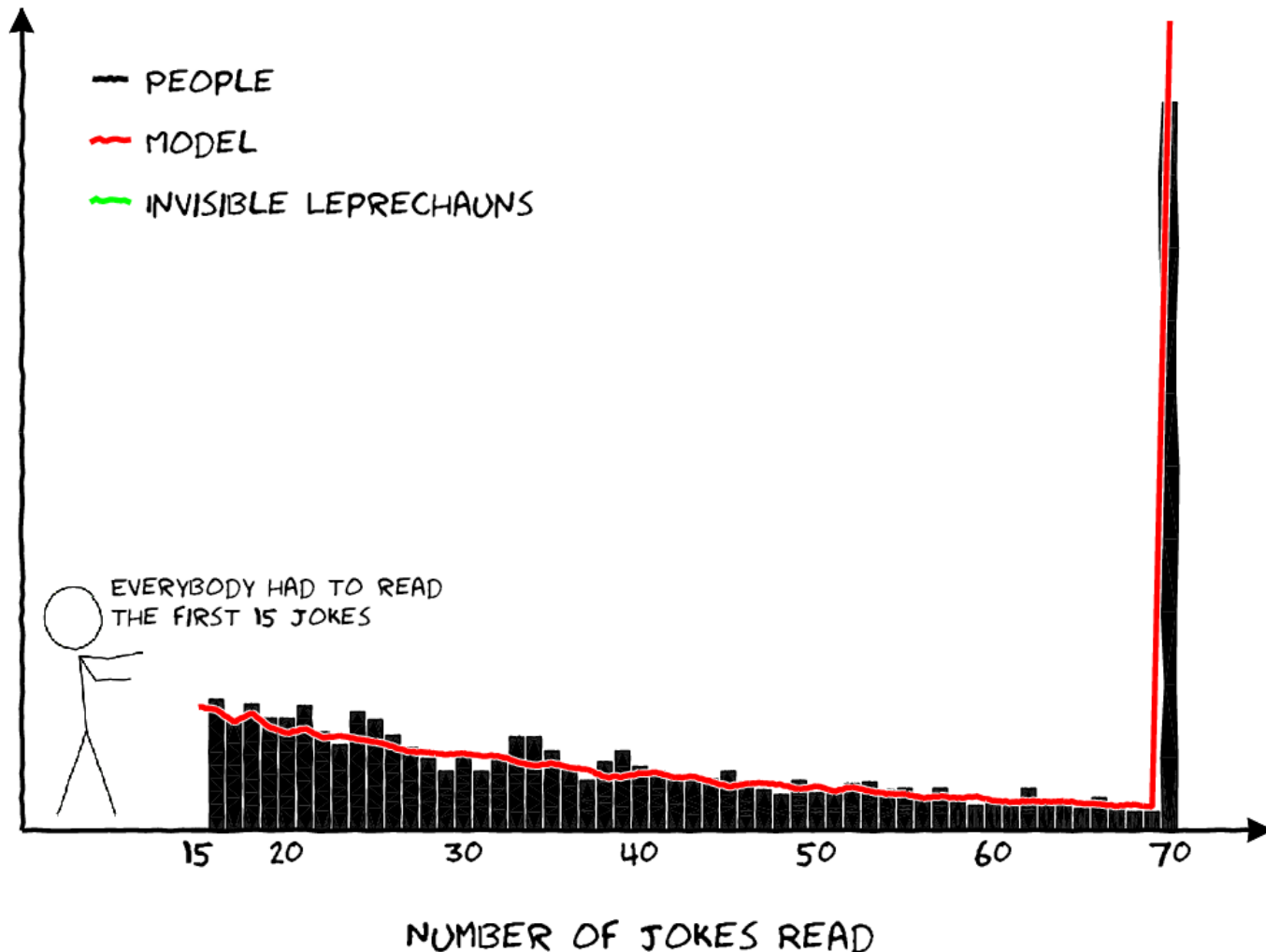
$$\alpha_i \sim 15 + \text{Geometric}(\theta)$$

$$y_i = \begin{cases} \alpha_i & \text{if } \alpha_i \leq 70 \\ 70 & \text{if } \alpha_i > 70 \end{cases}$$

```
model {  
  theta ~ duniform(0, 1)  
  for (i in 1:nPeople) {  
    alpha[i] ~ dnegbin(theta, 1)  
    # b is [1, 2, ..., 70]  
    y[i] ~ dinterval(alpha[i], b)  
  }  
}
```

# Posterior Predictive Agreement

- This simple model is descriptively adequate, with a termination probability around 3% for each person on each joke





# An Extended Model with Individual Differences

- Two sorts of individual differences

- people can have different termination probabilities

$$\mu \sim \text{Uniform}(0, 1)$$

$$\sigma \sim \text{Uniform}(0, 1)$$

$$\theta_i \sim \text{Gaussian}_{(0,1)}\left(\mu, \frac{1}{\sigma^2}\right)$$

- some people just always have to read all the jokes

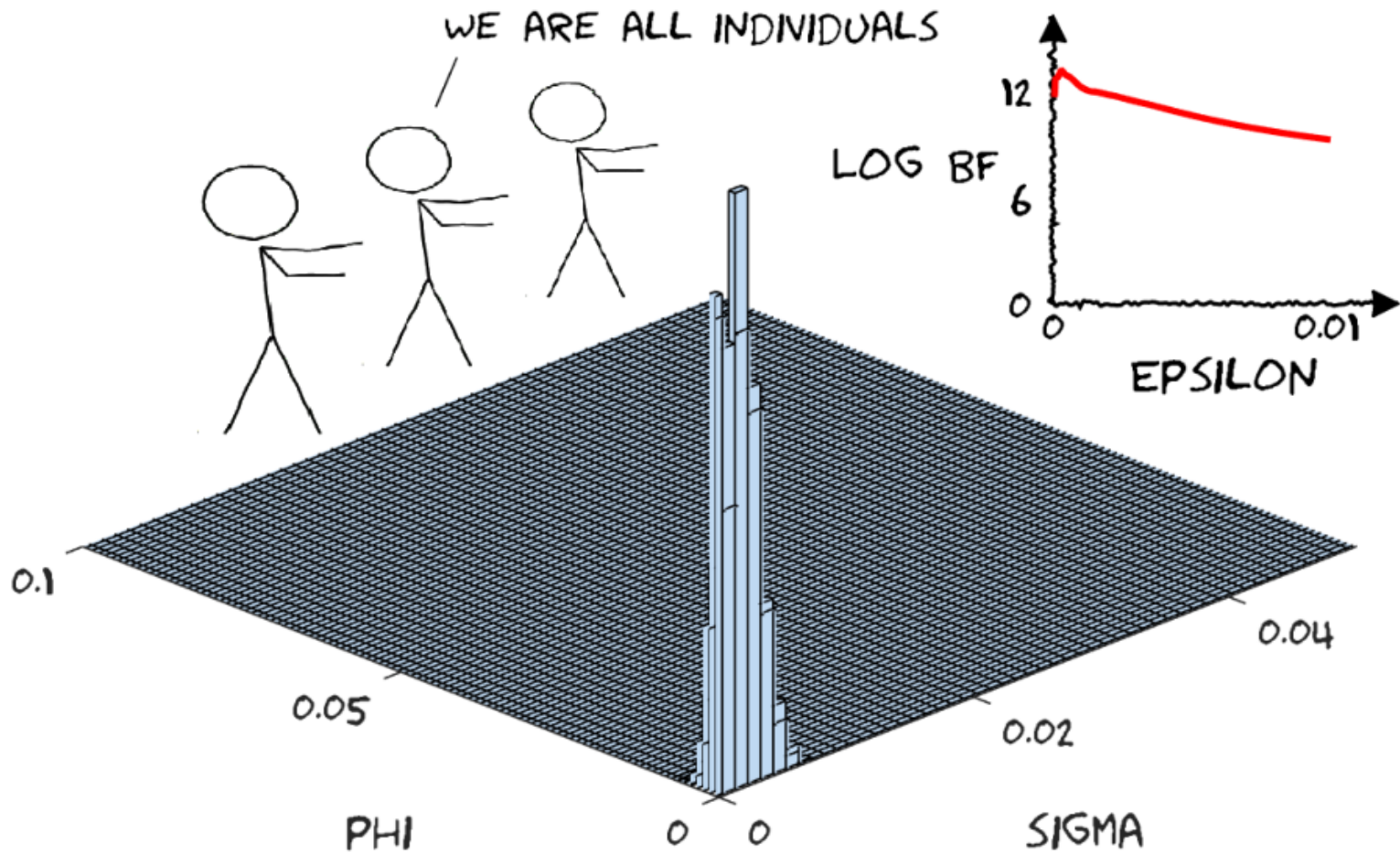
$$\phi \sim \text{Uniform}(0, 1)$$

$$z_i \sim \text{Bernoulli}(\phi)$$

$$y_i \sim \begin{cases} 15 + \text{Geometric}(\theta_i) & \text{if } z_i = 0 \text{ and } \alpha_i \leq 70 \\ 70 & \text{if } z_i = 0 \text{ and } \alpha_i > 70 \\ 70 & \text{if } z_i = 1. \end{cases}$$

# Are There Individual Differences?

- Bayes factor provides evidence for the original simple model, so these data provide no evidence for the individual differences



# **Bayesian Statistics and Cognitive Science**

# Bayesian methods

- Bayesian methods let you infer parameters, evaluate models, and understand and make predictions about data
- Three types of application in psychology

# Bayesian methods

- Bayesian methods let you infer parameters, evaluate models, and understand and make predictions about data
- Three types of application in psychology
  - **Bayes in the head:** Use Bayes as a theoretical metaphor, assuming that when people make inferences they apply Bayesian methods (at some level)



Josh Tenenbaum



Tom Griffiths



Mark Steyvers



Charles Kemp

# Bayesian methods

- Bayesian methods let you infer parameters, evaluate models, and understand and make predictions about data
- Three types of application in psychology
  - **Bayes in the head**
  - **Bayes for data analysis:** Instead of using frequentist estimation, confidence intervals, null hypothesis testing, and so on, use Bayesian inference to analyze data



EJ Wagenmakers



Jeff Rouder



Joachim  
Vandekerckhove



Richard Morey



John Kruschke



# Bayesian methods

- Bayesian methods let you infer parameters, evaluate models, and understand and make predictions about data
- Three types of application in psychology
  - **Bayes in the head**
  - **Bayes for data analysis**
  - **Bayes for cognitive modeling:** Use Bayesian inference to relate models of psychological processes to behavioral data



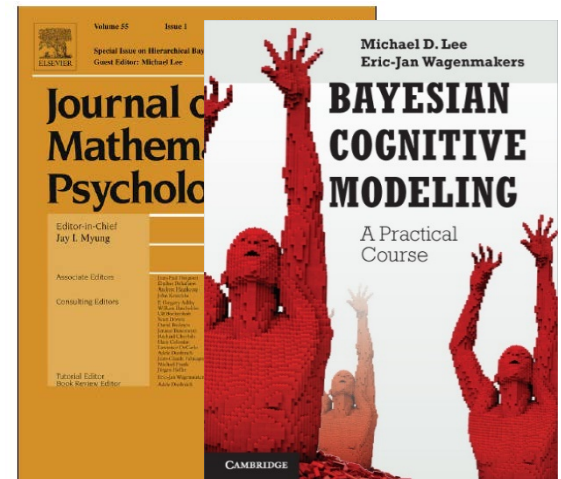
EJ Wagenmakers



Jeff Rouder

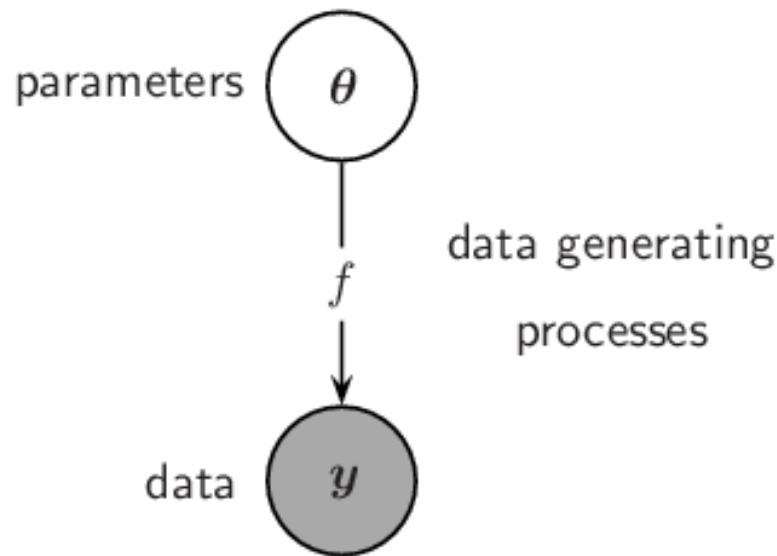


Wolf Vanpaemel



# Cognitive Models as Generative Statistical Models

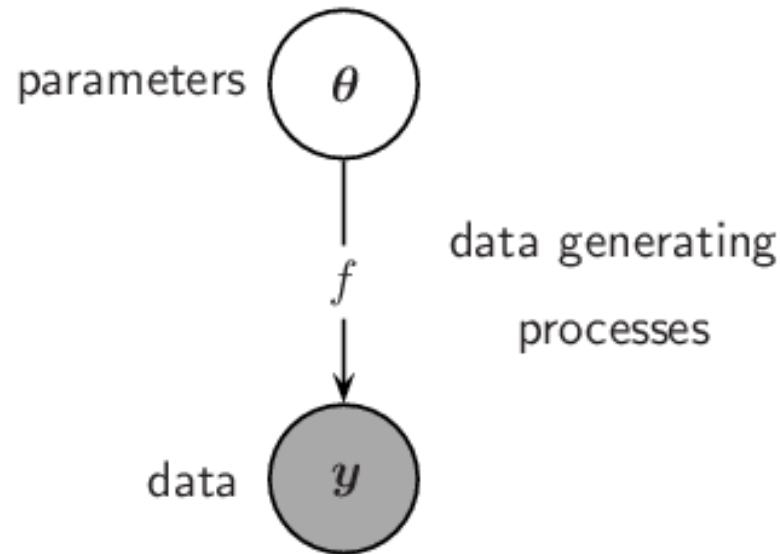
- Most cognitive models can be thought of as data generating processes, combining
  - Psychological processes, formalized by the likelihood
  - Psychological variables, formalized by parameters





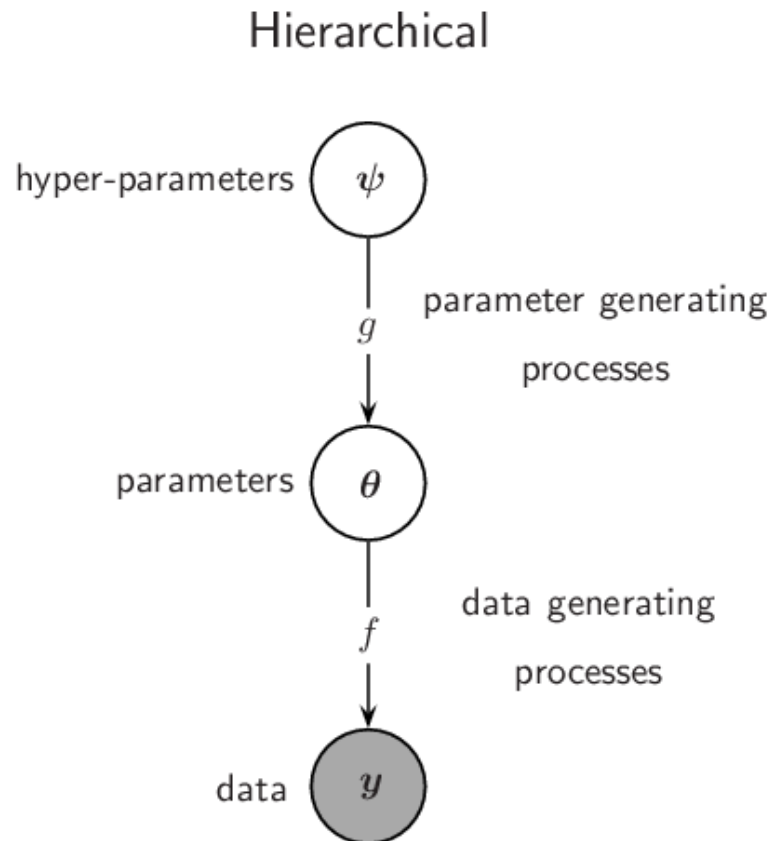
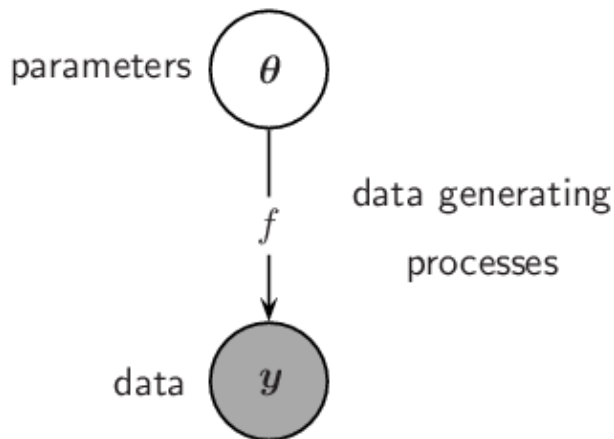
# Attraction of Bayesian methods

- Beyond the conceptual coherence and completeness, the great advantage of Bayesian methods is they allow cognitive that are more complicated than the standard one to be considered



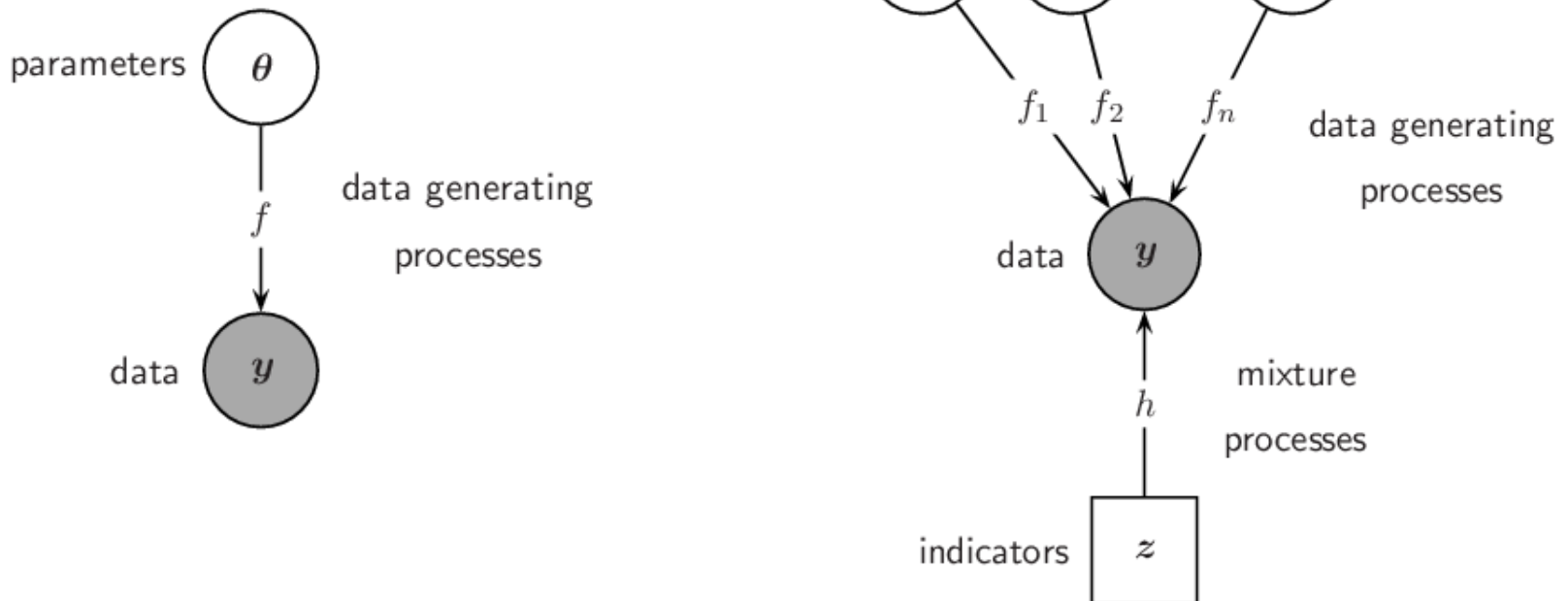
# Hierarchical modeling

- Hierarchical models extend the standard approach by including a modeling account of how the basic model parameters themselves are generated



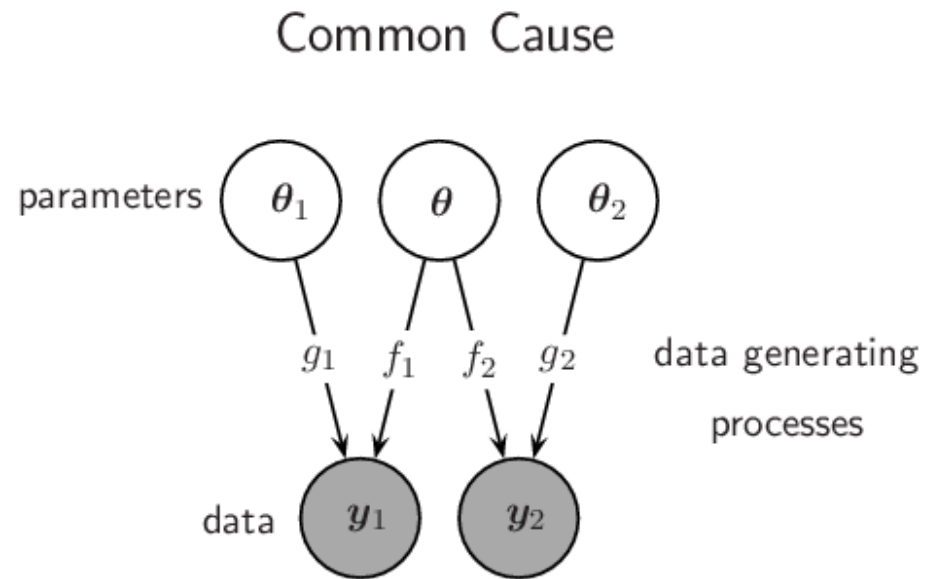
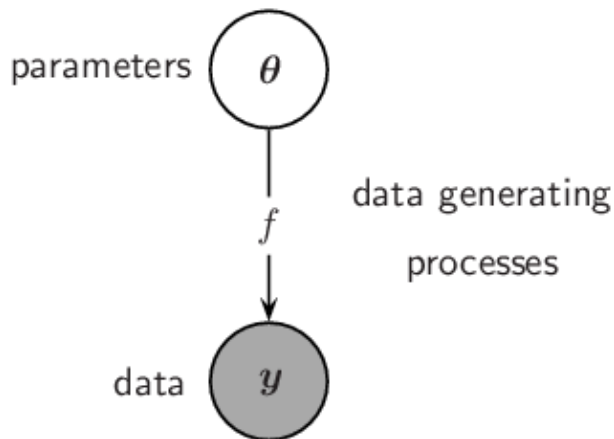
# Latent-mixture modeling

- Latent-mixture models extend the standard approach by allowing behavioral data to be generated as a mixture of multiple different processes and controlling parameters



# Common-cause modeling

- Common-cause models extend the standard approach by allowing the same psychological variables to influence multiple sorts of observed behavior



# **Wisdom of the Crowd for Ranking Data**

RANK SITE MONTHLY PEOPLE

1  google.com 235,702,560

2  youtube.com 200,078,640

3  facebook.com 142,452,240

4  msn.com 122,279,512

5  yahoo.com 115,347,760

6  bing.com 104,640,000

7  amazon.com 101,915,368

---

37  foxnews.com 22,965,512

38  blogger.com 22,904,332

39  Hidden profile —

40  ranker.com 21,974,350

41  Hidden profile —

42  answers.com 20,358,508

43  craigslist.org 20,114,412





# The Most Anticipated 2013 Films

**List Criteria:** Only films with a release date in 2013.

List of the most eagerly anticipated movies opening in 2013. Before the year gets going, it's people are most excited about seeing in the coming year. Will these movies live up to expectations and dominate people's thinking about 2013 films, with the most conventionally exciting titles even on a list like this in advance can potentially answer these questions down the road. (Don't [Anticipated Films of 2012 here!](#))

2013 will continue many of the trends set in the preceding years. Remakes remain a staple. **SORRY, VOTING AND RERANKING FOR THIS LIST HAS BEEN CLOSED.**

SEE LIST RANKED BY

REGION AGE MEN WOMEN

26 RERANKS



THE LIST
 OPTIONS

**1**
 681
 252
 
*the people's #1*  
**Man of Steel**  
 Amy Adams, Diane Lane, Ke...

**The Hobbit: The Desolation of Smaug**  
 Cate Blanchett, Evangeline Lilly, Orlando Bloom,

**2**
 640
 225
 
*most listed*  
**The Hobbit: The**  
 Cate Blanchett, Evangeline L...

**The Lone Ranger**  
 Johnny Depp, Helena Bonham Carter, Armie

**3**
 873
 191
 
*ranked low on reranks*  
**Iron Man 3**  
 Gwyneth Paltrow, Robert Do...

**The Great Gatsby**  
 Leonardo DiCaprio, Isla Fisher, Carey Mulligan,

**4**
 606
 202
 
**Thor: The Dark**  
 Natalie Portman, Kat Dennin...

**Now You See Me**  
 Morgan Freeman, Isla Fisher, Michael Caine,

**5**
 692
 211
 
**Star Trek Into**  
 Zoe Saldana, Benedict Cum...

**Percy Jackson: Sea of Monsters**  
 Alexandra Daddario, Logan Lerman, Jake Abel,

Ranker List



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SEE LIST RANKED BY

REGION AGE MEN WOMEN

26 RERANKS



THE LIST OPTIONS

Actors

bonnie4arq

- 1**

681 24

*the people's #1*  
**Man of Steel**  
 Amy Adams, Diane Lane, ...
- 2**

640 22

*most listed*  
**The Hobbit: The Desolation of Smaug**  
 Cate Blanchett, Evangeline Lilly, ...
- 3**

873 19

*ranked low on reranks*  
**Iron Man 3**  
 Gwyneth Paltrow, Robert Downey Jr., ...
- 4**

606 20

*ranked low on reranks*  
**Thor: The Dark World**  
 Natalie Portman, Kat Dennings, ...
- 5**

692 21

*ranked low on reranks*  
**Star Trek Into Darkness**  
 Zoe Saldana, Benedict Cumberbatch, ...

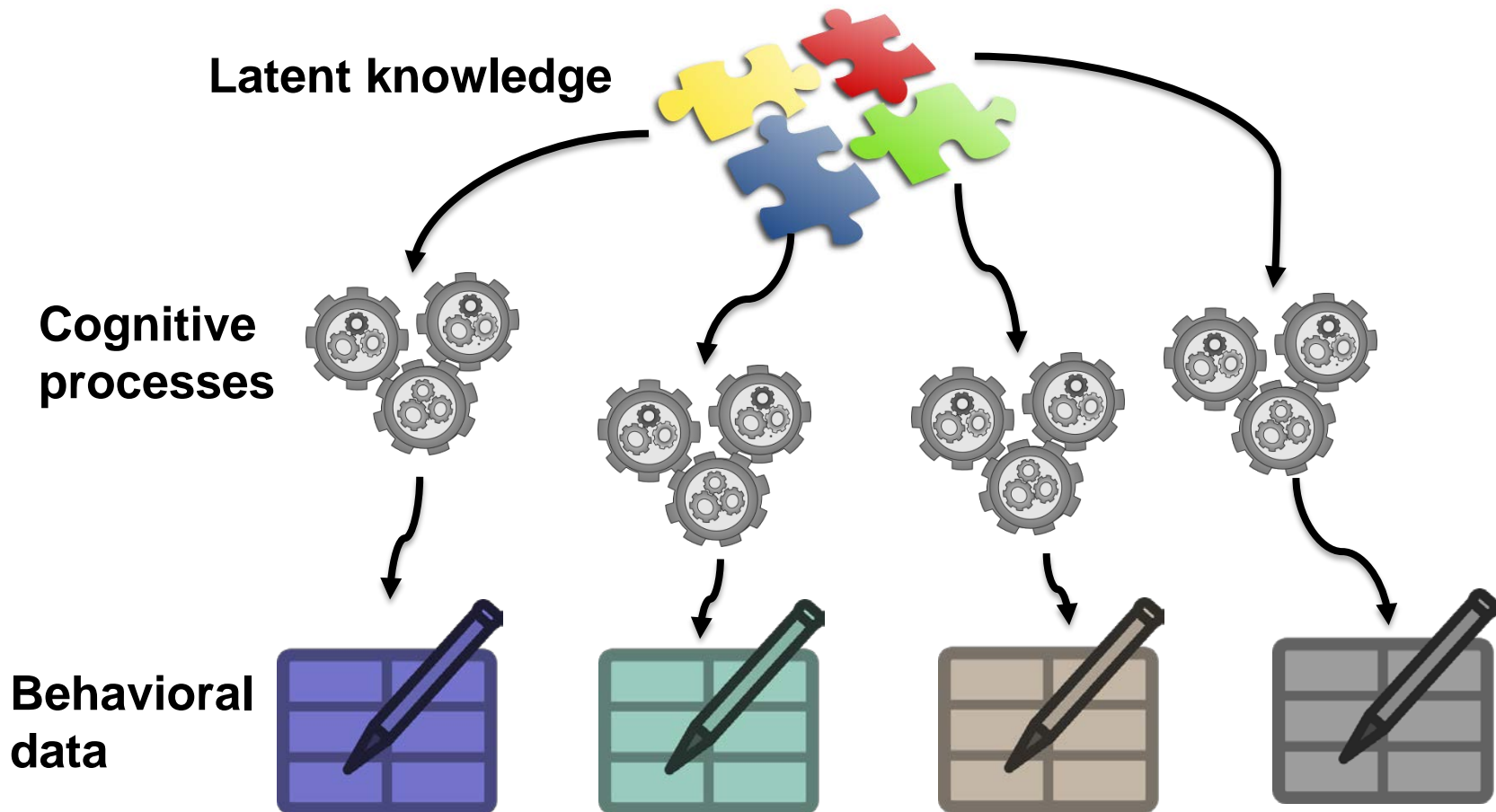
- The Hobbit: The Desolation of Smaug**  
 Cate Blanchett, Evangeline Lilly, Orlando Bloom, ...
- The Lone Ranger**  
 Johnny Depp, Helena Bonham Carter, Arnie Olden, ...
- The Great Gatsby**  
 Leonardo DiCaprio, Isla Fisher, Carey Mulligan, ...
- Now You See Me**  
 Morgan Freeman, Isla Fisher, Michael Caine, ...
- Percy Jackson: Sea of Monsters**  
 Alexandra Daddario, Logan Lerman, Jake Abel, ...

# Rankings

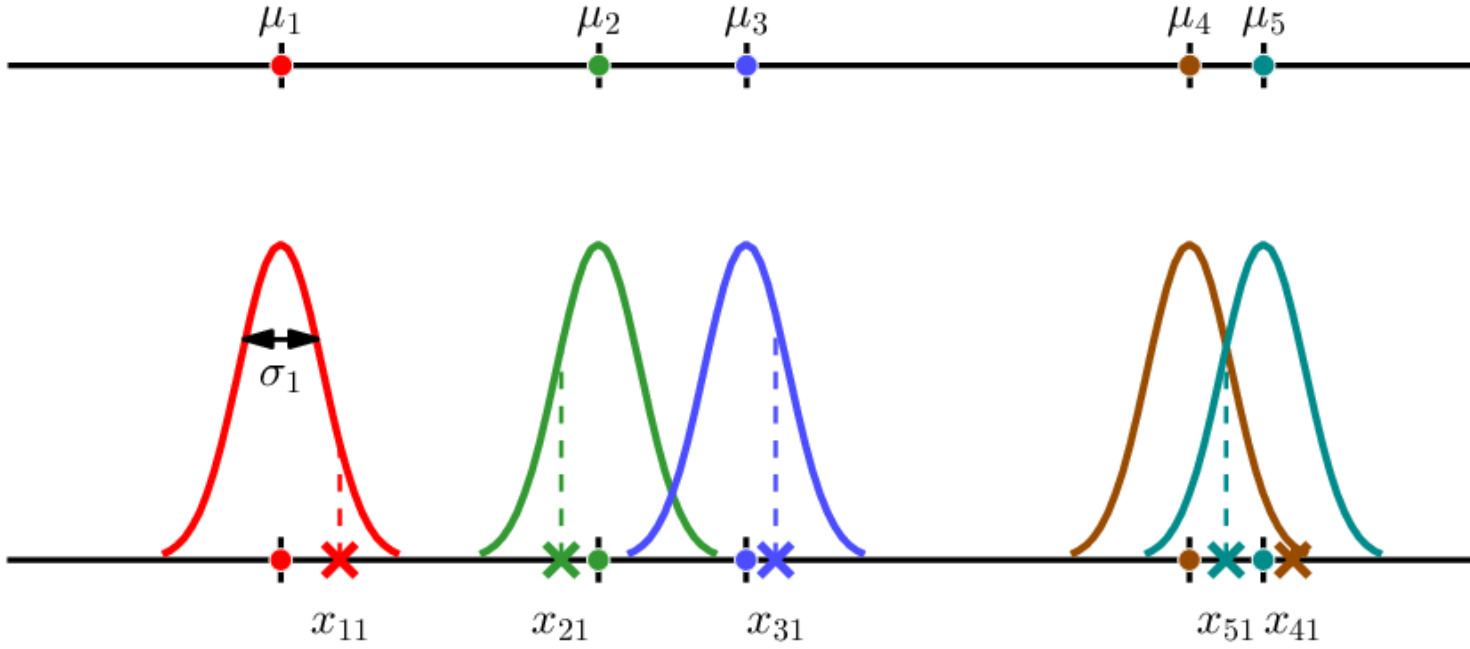


# Generative Cognitive Modeling Approach

- Use cognitive models of the processes that generate data, and the individual differences across people, as the bridge between the available data and the latent group knowledge



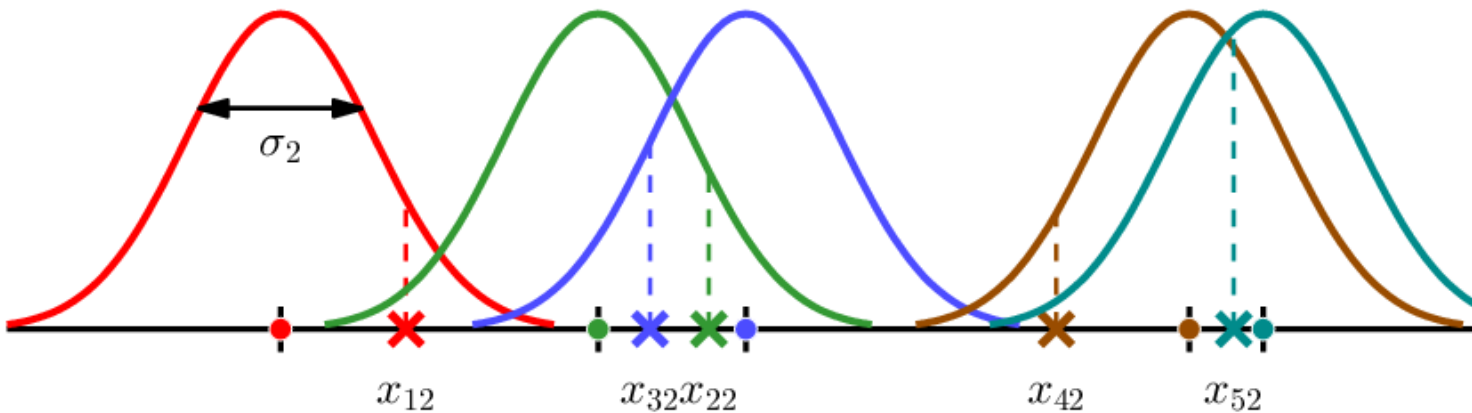
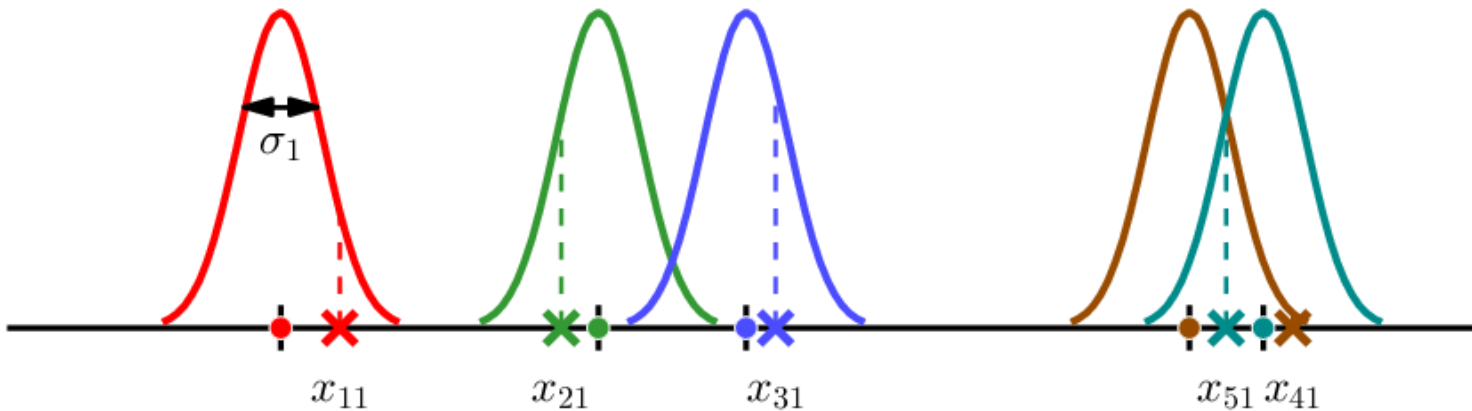
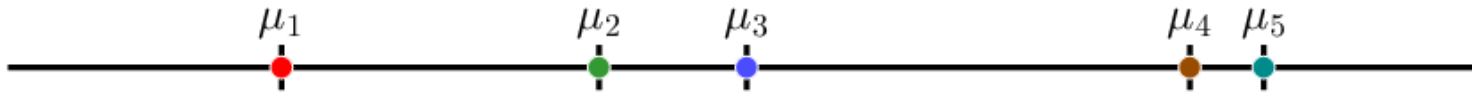
# Thurstone (1927) Model for Partial Lists



Observed Ranking

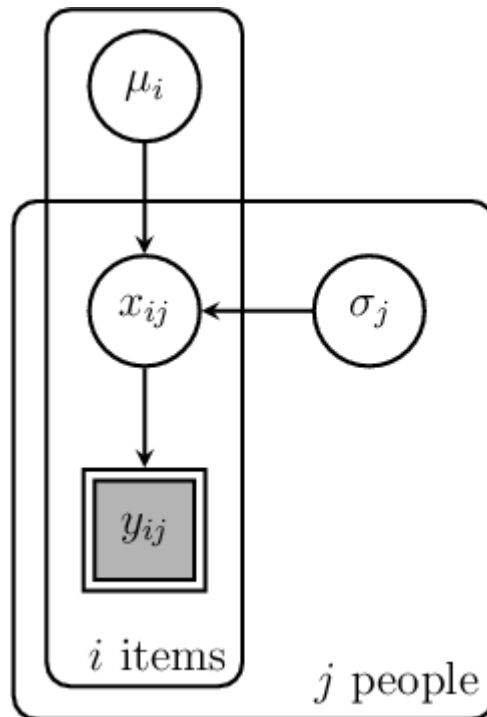
$$y_1 = (1, 2, 3, 5)$$

# Thurstone (1927) Model for Partial Lists



# Graphical Model

- Implement the Thurstonian model as a graphical model in JAGS, inferring two sorts of psychological parameters
  - latent item locations
  - expertise of each person

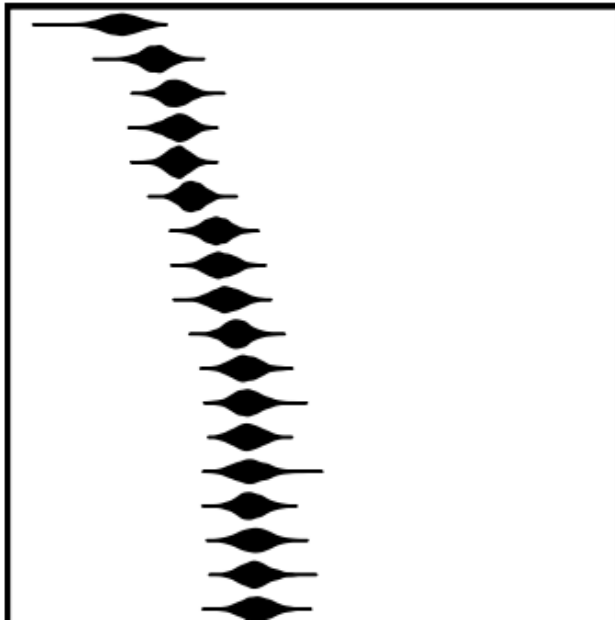


$$\begin{aligned}\mu_i &\sim \text{Uniform}(0, 1) \\ \sigma_j &\sim \text{Uniform}(0, 1) \\ x_{ij} &\sim \text{Gaussian}(\mu_i, 1/\sigma_j^2) \\ y_{ij} &= \text{Rank}(x_{ij})\end{aligned}$$

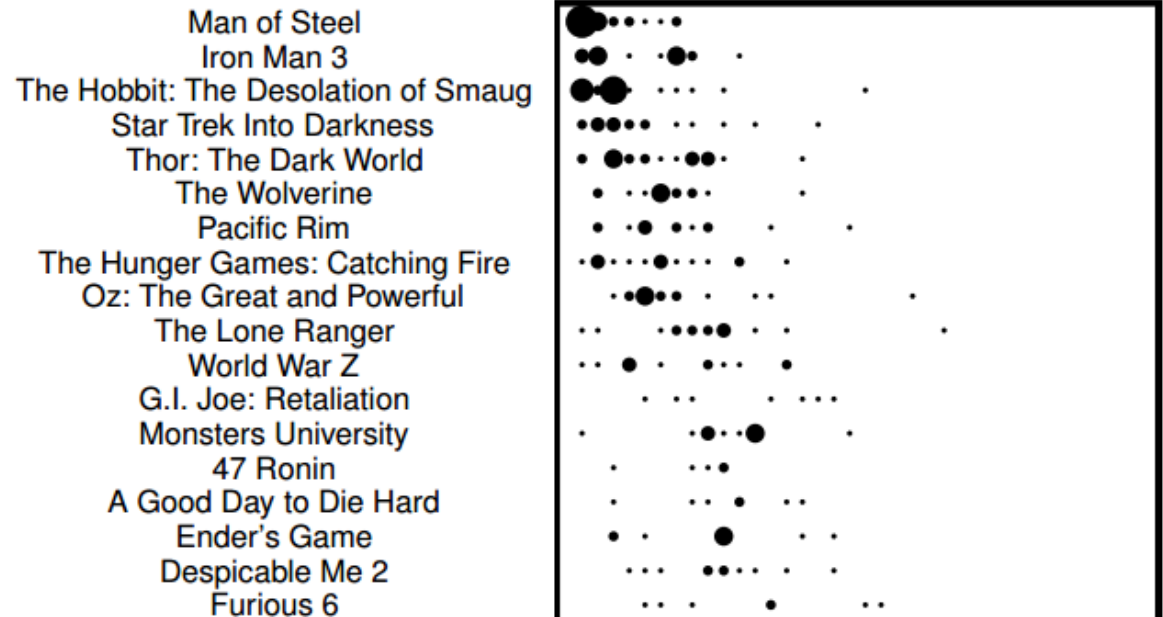
# Movie Inferences

- Order movies in terms of expected posterior of their inferred item location

Model Inferences

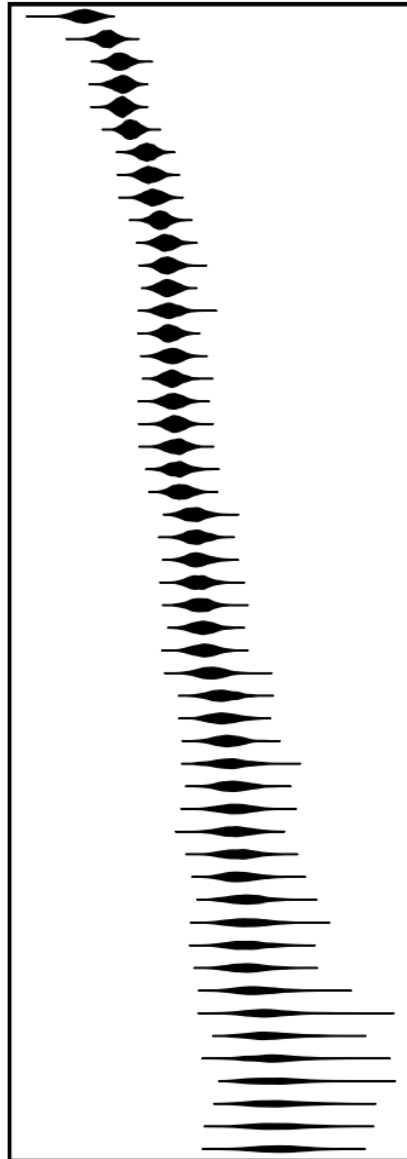


Observed Data



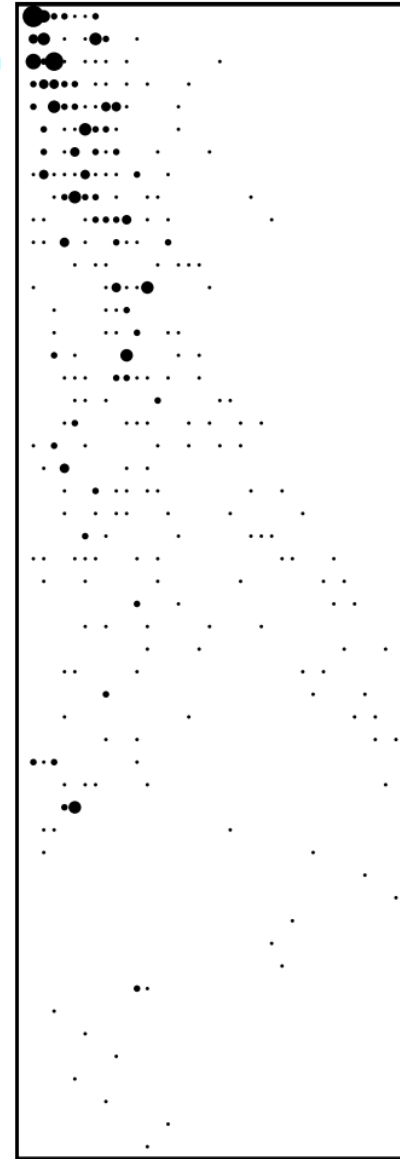
# Movie Inferences

Model Inferences



- Man of Steel
- Iron Man 3
- The Hobbit: The Desolation of Smaug
- Star Trek Into Darkness
- Thor: The Dark World
- The Wolverine
- Pacific Rim
- The Hunger Games: Catching Fire
- Oz: The Great and Powerful
- The Lone Ranger
- World War Z
- G.I. Joe: Retaliation
- Monsters University
- 47 Ronin
- A Good Day to Die Hard
- Ender's Game
- Despicable Me 2
- Furious 6
- The Hangover Part III
- Elysium
- Kick-Ass 2
- Oblivion
- Evil Dead
- Gangster Squad
- Jack the Giant Slayer
- After Earth
- R.I.P.D.
- Carrie
- Grown Ups 2
- Scary Movie 5
- Hansel & Gretel: Witch Hunters
- The Host
- The Croods
- The Great Gatsby
- Now You See Me
- Percy Jackson: Sea of Monsters
- Anchorman 2: The Legend Continues
- Gravity
- Warm Bodies
- Captain Phillips
- Side Effects
- Dark Skies
- The Wolf of Wall Street
- The Purge
- Only God Forgives
- Pieta
- Upstream Color
- Filth
- The Conjuring
- Simon Killer
- Alan Partridge: Alpha Papa

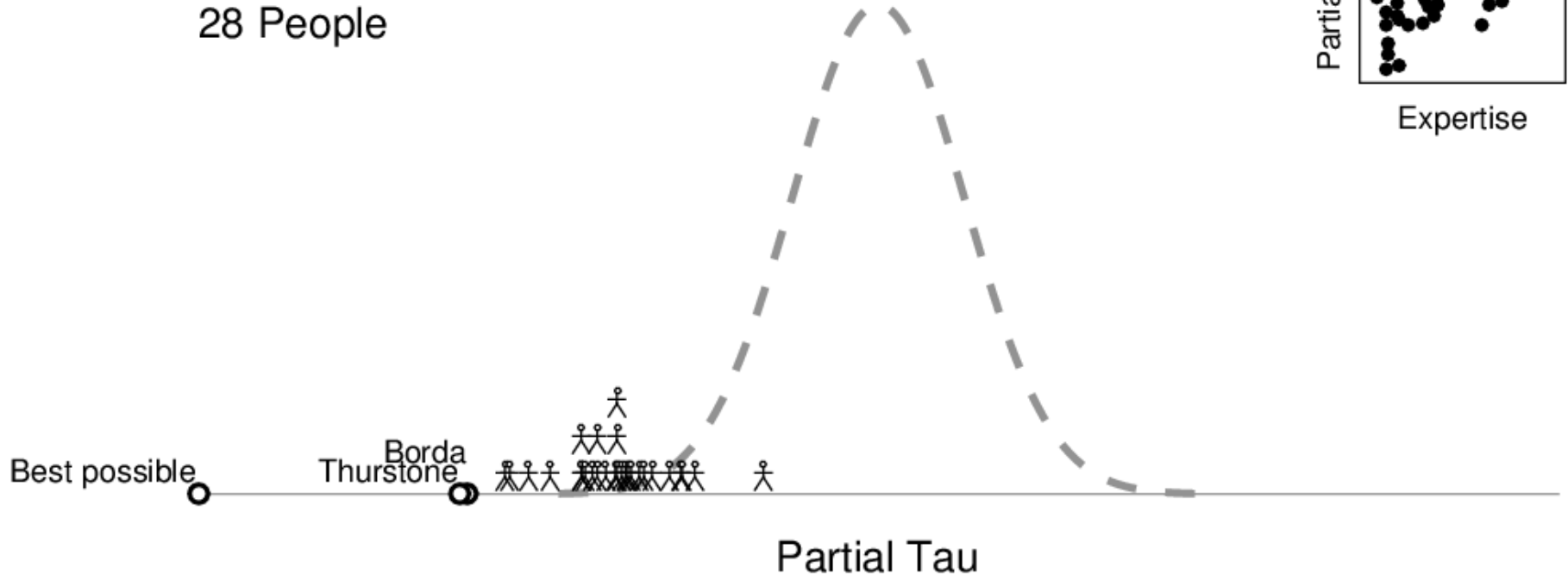
Observed Data



# Movie Performance

- We used ground truth box office takings from imdb.com, and partial tau to measure how well individuals, the model, and the Borda count statistical method performed

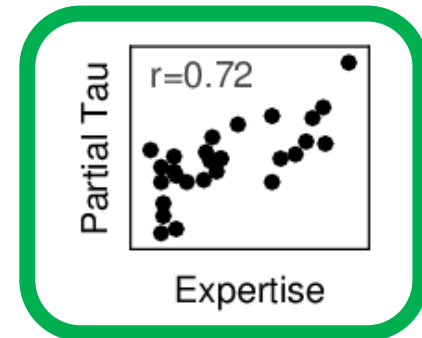
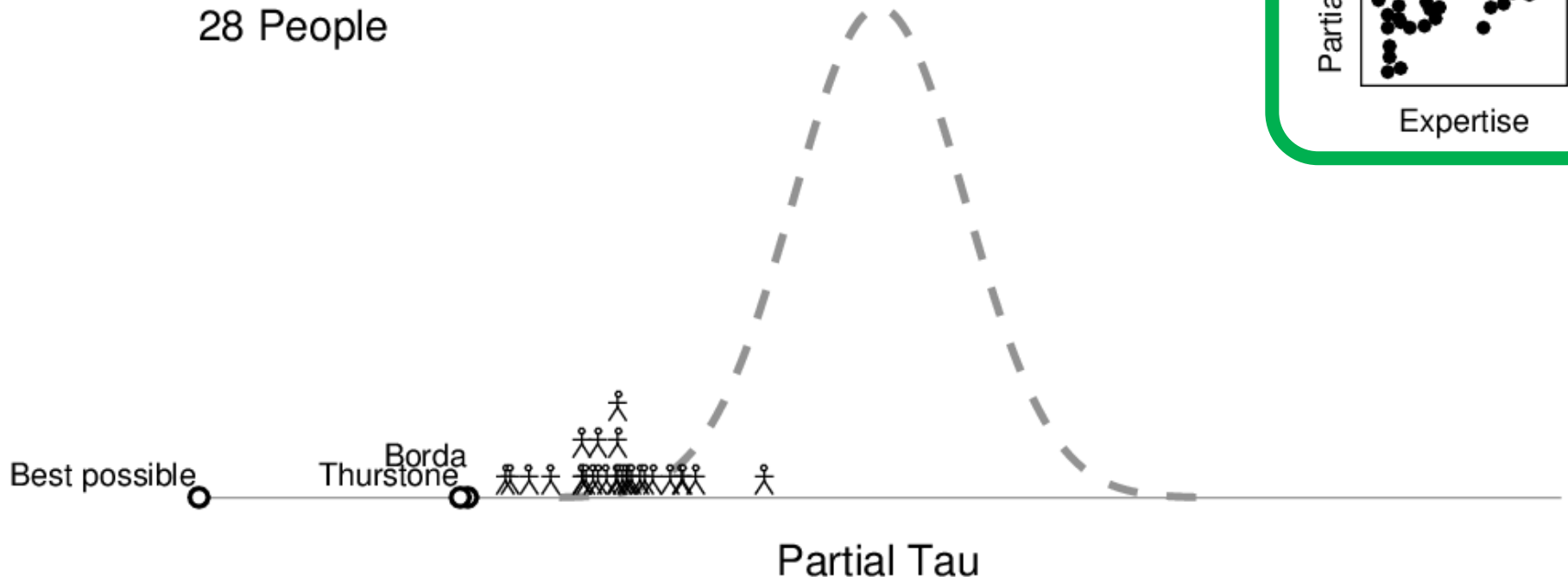
Movie Popularity  
59 Items  
28 People



# Movie Performance

- Both the model and Borda count make very good predictions, and are better than even the best individual
- The model also provides a good prediction of relative expertise

Movie Popularity  
59 Items  
28 People







Vote List

by Ranker Soccer ★ RANKER STAFF | tags | 15,784 votes | 2,083 voters | 43,207 views

# Who Will Win the 2014 World Cup

List Criteria: Only teams participating in the 2014 World Cup

A list of what country will win the 2014 World Cup. It is the most watched event in the world. The best footballing countries in the world aim to leave Brazil with the trophy in the world. The best footballing countries in the world aim to leave Brazil with the trophy, bringing prestige and honor to their respective countries, and coming home as champions (it won't be going anywhere).

Just ask Spain what it meant when their players took home the 2010 trophy, the first in its history. **SORRY, VOTING AND RERANKING FOR THIS LIST HAS BEEN CLOSED.**

SEE LIST RANKED BY REGION AGE MEN WOMEN 19 RERANKS

THE LIST OPTIONS Capital

Scandinavia14

1 Germany Berlin  
740 likes, 468 dislikes

Germany Berlin

2 Netherlands Amsterdam  
489 likes, 389 dislikes

Brazil Brasilia

3 Brazil Brasilia  
483 likes, 502 dislikes

Spain Madrid

4 Argentina Buenos Aires  
419 likes, 461 dislikes

Russia Moscow

5 France Paris  
227 likes, 357 dislikes

Portugal Lisbon

# Ranker List



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SEE LIST RANKED BY REGION AGE MEN WOMEN 19 RERANKS X

THE LIST OPTIONS Capital Scandinavia14

1		Germany Berlin		Germany Berlin
2		Netherlands Amsterdam		Brazil Brasília
3		Brazil Brasília		Spain Madrid
4		Argentina Buenos Aires		Russia Moscow
5		France Paris		Portugal Lisbon

# Rankings



# Who Will Win the 2014 World Cup

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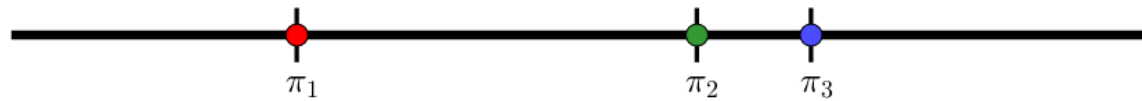
THE LIST	OPTIONS	Capital	Scandinavia14
1	740  468	<b>Germany</b> Berlin	<b>Germany</b> Berlin
2	489  389	<b>Netherlands</b> Amsterdam	<b>Brazil</b> Brasília
3	483  502	<b>Brazil</b> Brasília	<b>Spain</b> Madrid
4	419  461	<b>Argentina</b> Buenos Aires	<b>Russia</b> Moscow
5	227  357	<b>France</b> Paris	<b>Portugal</b> Lisbon

Voting

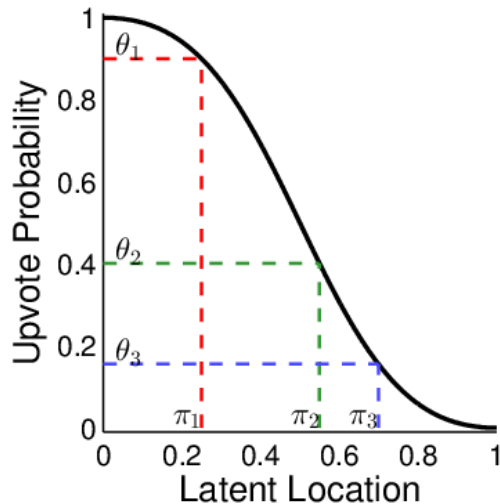
# Modeling Ranking and Voting Data

- Extend the Thurstone model of rank data generation to include a transfer function mapping latent item location to up-voting

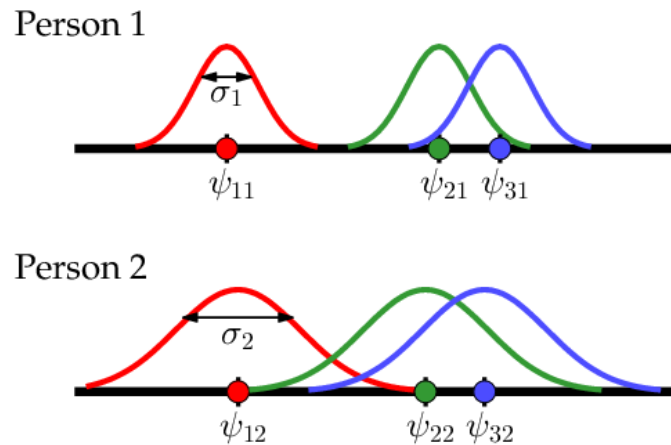
Latent Ground Truth



Voting Data



Ranking Data







# Bracket Prior Information





- The bracket for the World Cup is highly relevant prior information, which is natural for Bayesian analysis

## 2014 FIFA WORLD CUP GROUPS

### GROUP A

- 1  BRAZIL
- 2  CROATIA
- 3  MEXICO
- 4  CAMEROON





### GROUP B

- 1  SPAIN
- 2  NETHERLANDS
- 3  CHILE
- 4  AUSTRALIA





### GROUP C

- 1  COLOMBIA
- 2  GREECE
- 3  IVORY COAST
- 4  JAPAN





### GROUP D

- 1  URUGUAY
- 2  COSTA RICA
- 3  ENGLAND
- 4  ITALY

### GROUP E

- 1  SWITZERLAND
- 2  ECUADOR
- 3  FRANCE
- 4  HONDURAS





### GROUP F

- 1  ARGENTINA
- 2  BOSNIA & HERZ.
- 3  IRAN
- 4  NIGERIA

### GROUP G

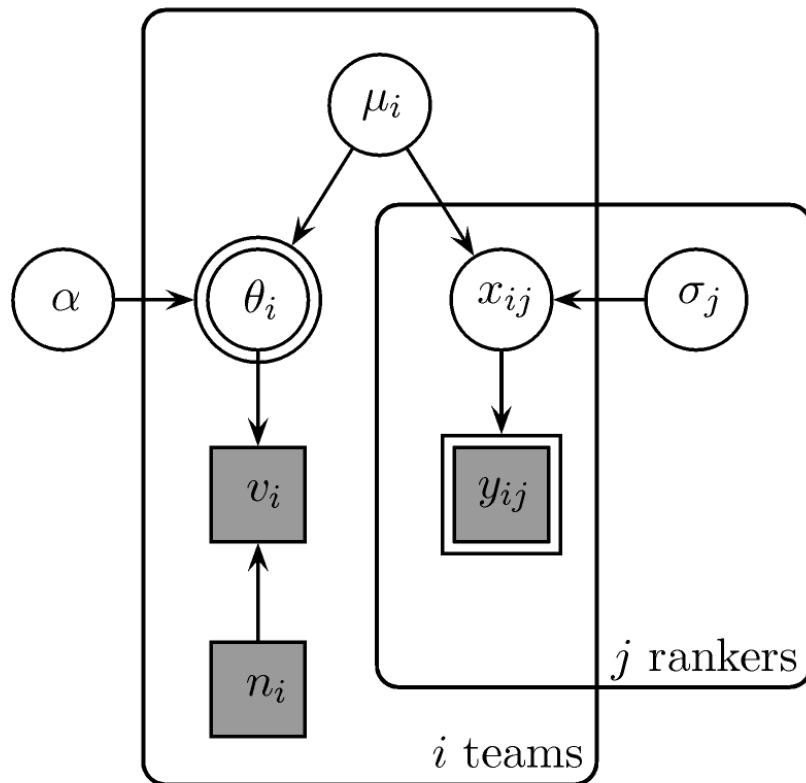
- 1  GERMANY
- 2  PORTUGAL
- 3  GHANA
- 4  USA

### GROUP H

- 1  BELGIUM
- 2  ALGERIA
- 3  RUSSIA
- 4  KOREA REP.

# Combining All Relevant Information

- Graphical model combines ranking data, voting data, and bracket structure prior information



$$\mu_i \stackrel{\mathcal{B}}{\sim} \text{Gaussian}(0, 0.001)$$

$$\sigma_j \sim \text{Uniform}(0, 20)$$

$$x_{ij} \sim \text{Gaussian}(\mu_i, \sigma_j^{-2})$$

$$y_{ij} \leftarrow \text{Rank}(x_{ij})$$

$$\alpha \sim \text{Gaussian}(0, 0.01)$$

$$\theta_i \leftarrow 1 / (1 + \exp\{-\alpha\mu_i\})$$

$$v_i \sim \text{Binomial}(\theta_i, n_i)$$

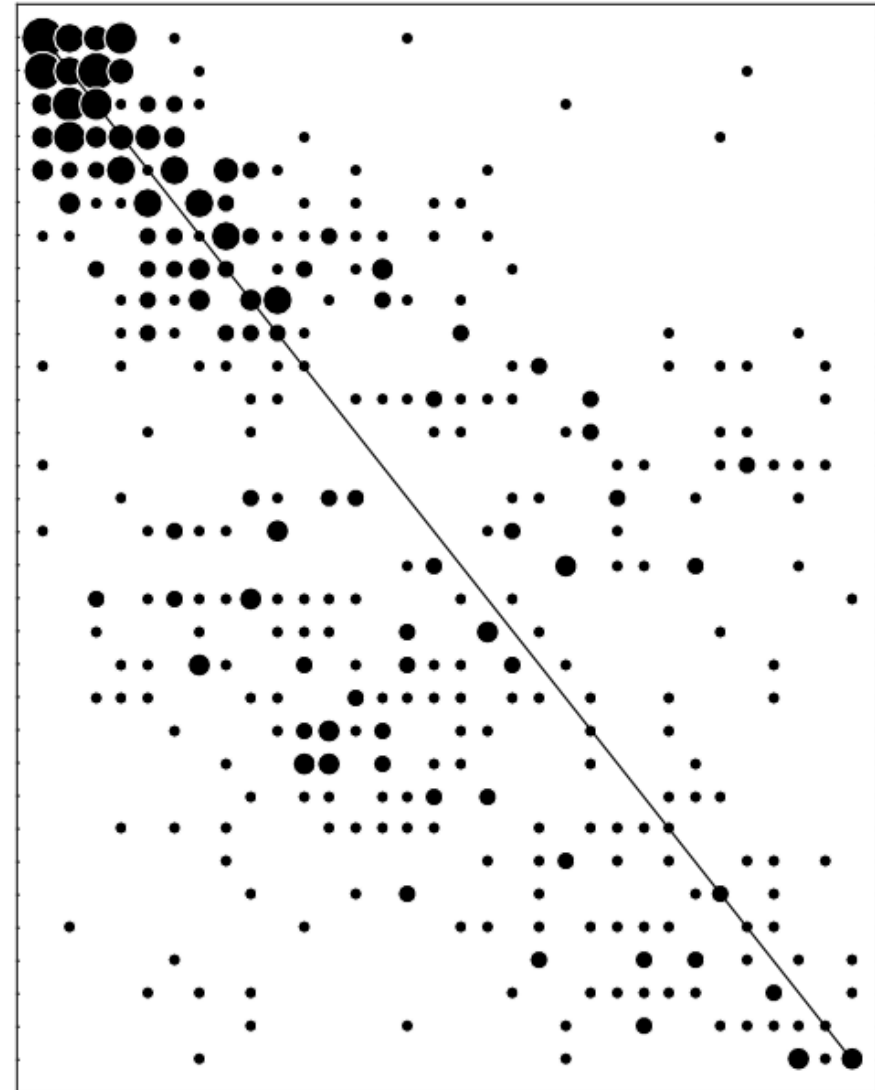
# Model World Cup Predictions

Voting Data



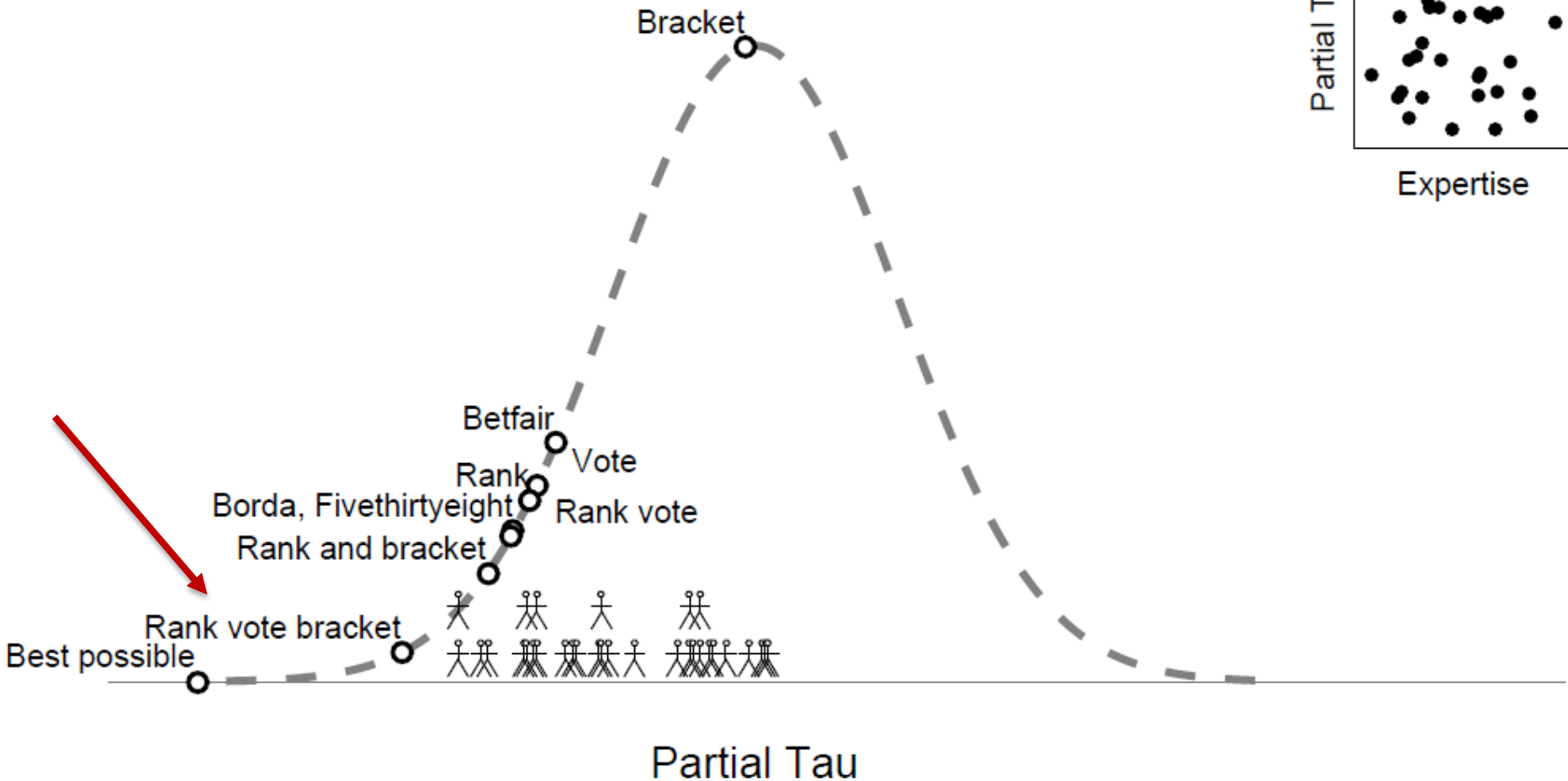
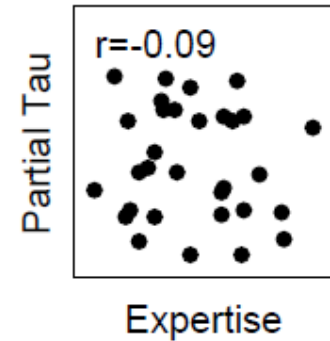
- Germany
- Brazil
- Spain
- Argentina
- Netherlands
- Italy
- England
- Belgium
- France
- Colombia
- Mexico
- Switzerland
- Nigeria
- Algeria
- United States of America
- Portugal
- Greece
- Uruguay
- Côte d'Ivoire
- Chile
- Ghana
- Croatia
- Japan
- Bosnia and Herzegovina
- Russia
- Ecuador
- South Korea
- Cameroon
- Australia
- Costa Rica
- Honduras
- Iran

Ranking Data



# World Cup Performance

- Combining all the information makes a better prediction than all people, and various betting and prediction sites

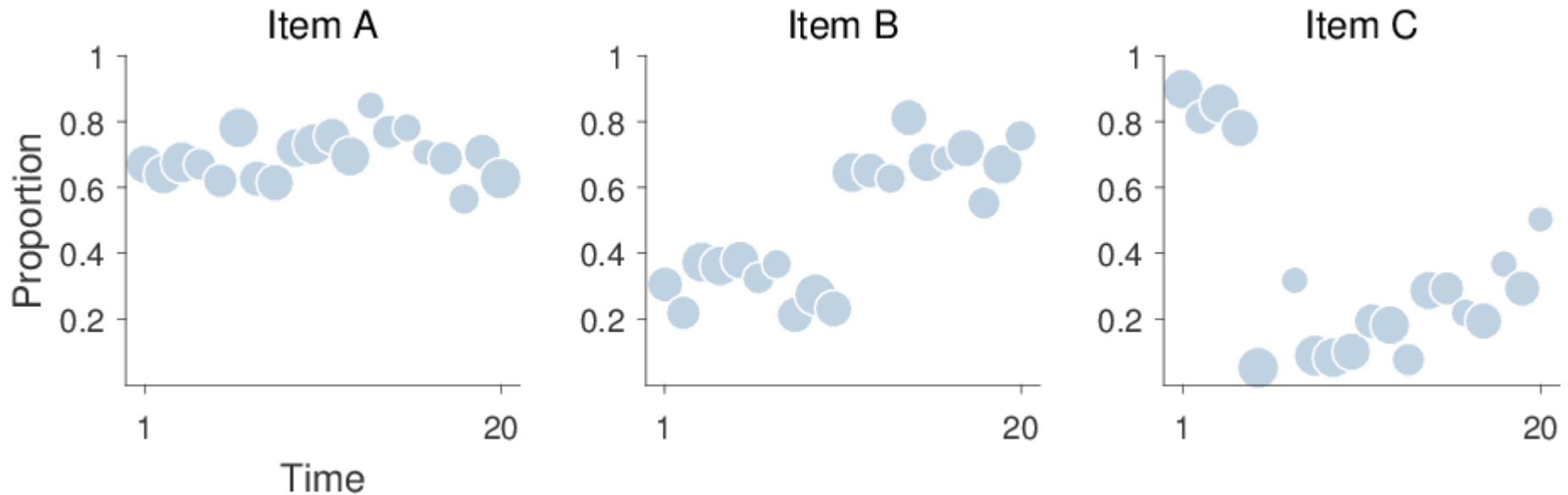




# **Detecting Step Changes in Cognition**

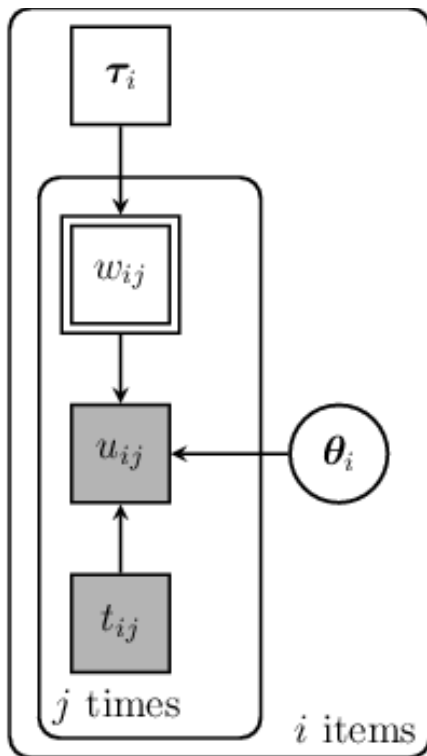
# Detecting Step Change

- Lots of cognitive phenomena and observed behaviors change suddenly
  - Large statistical literature on inferring how many changes there are, and where they occur (Barry & Hartigan 1993, Chib 1998, Fearnhead 2006, Adams and Mackay 2007)



# A Graphical Model of Detecting Change

- Key is the spike-and-slab prior on the (ordered) change points
  - half the prior mass to “no change”
  - other half distributed equally over meaningful possibilities



$$\theta_{ik} \sim \text{Uniform}(0, 1)$$

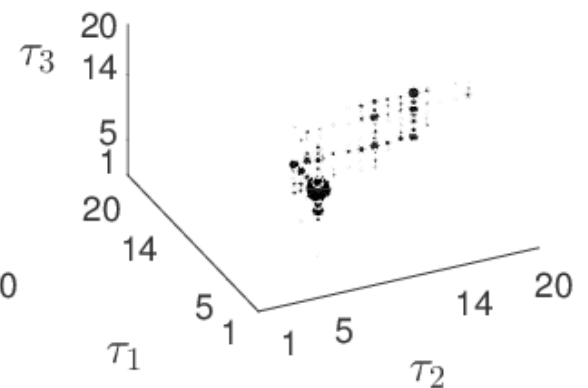
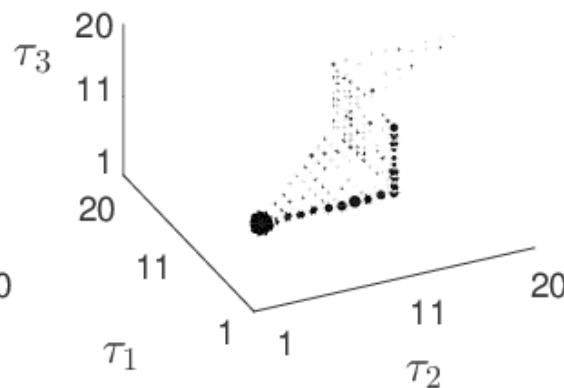
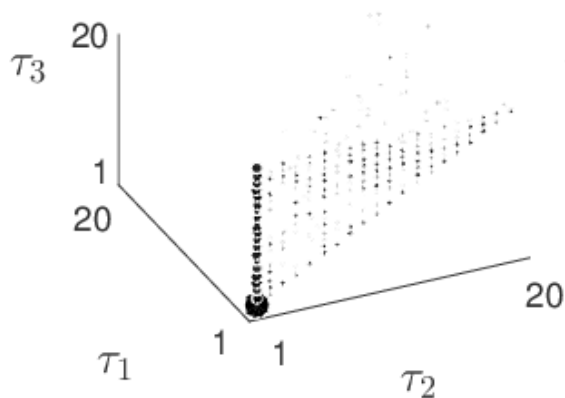
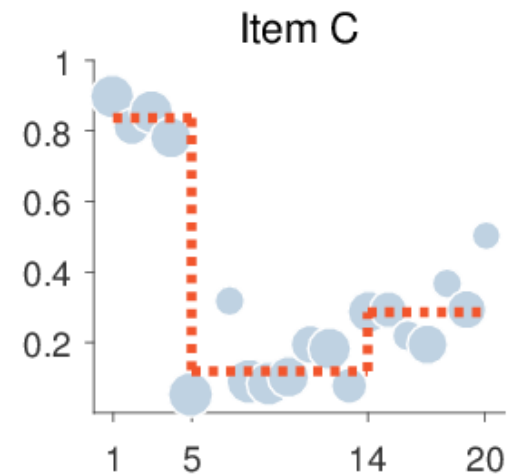
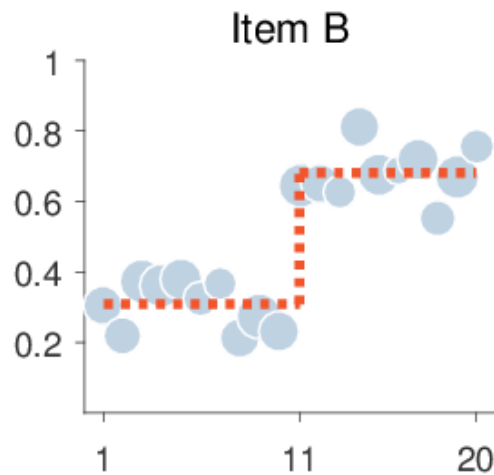
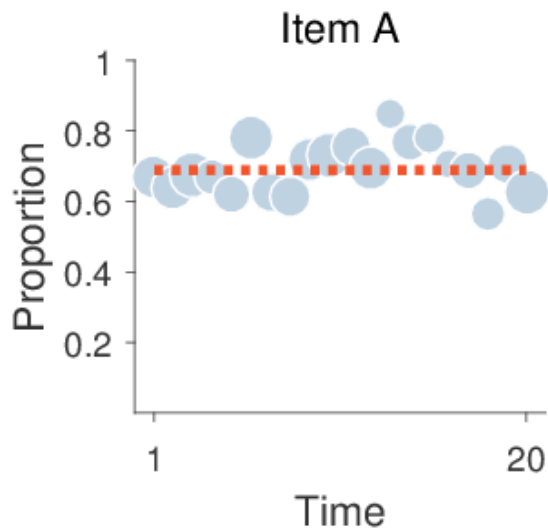
$$\tau_{ik} \sim \text{Categorical}\left(\overbrace{\left(\frac{n}{2n-1}, \frac{1}{2n-1}, \dots, \frac{1}{2n-1}\right)}^{n \text{ times}}\right); \tau_1 \leq \dots \leq \tau_\gamma$$

$$w_{ij} = \sum_k \mathcal{I}(j \geq \tau_{ik})$$

$$u_{ij} \sim \text{Binomial}(\theta_{i,w_{ij}}, t_{ij})$$

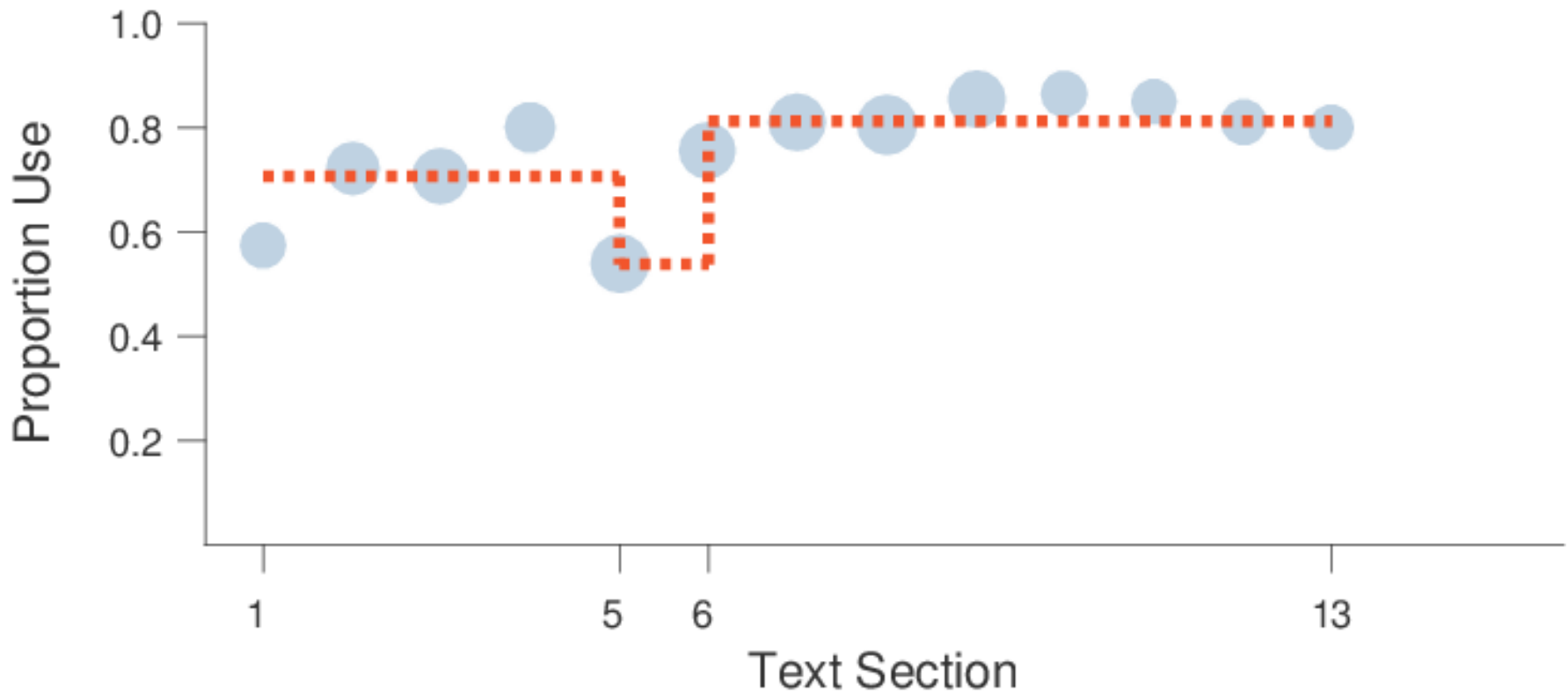
# Demonstration With Toy Data

- Take the posterior mode as point estimate of change points, and infer rates conditional on those points



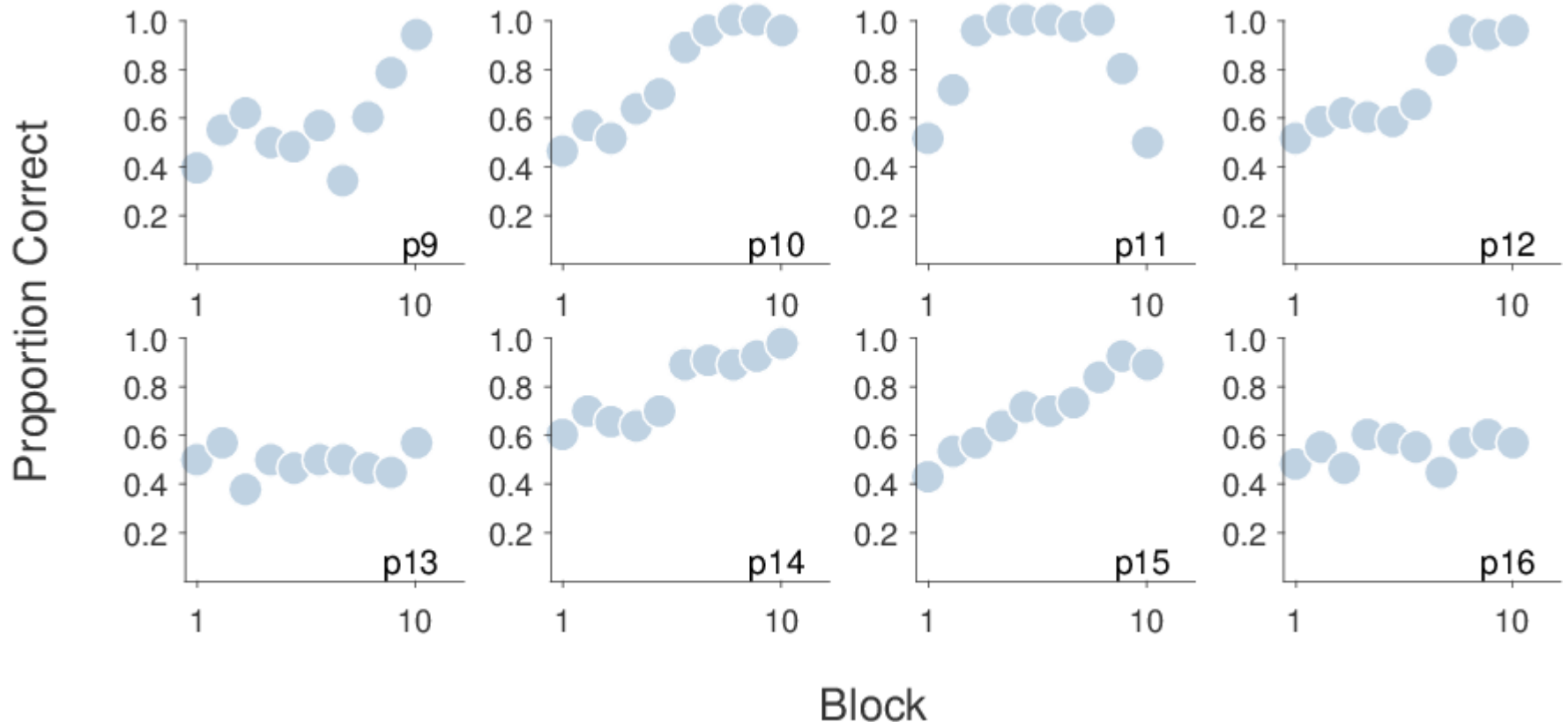
# Lindisfarne Scribes Problem

- Looks at the proportion of use of forms like “drives” vs “driveth” in 13 ordered gospels, as a basis for inferring changes in the scribes
- Our inferences match the analytic ones of Smith (1975)



# Category Learning Performance

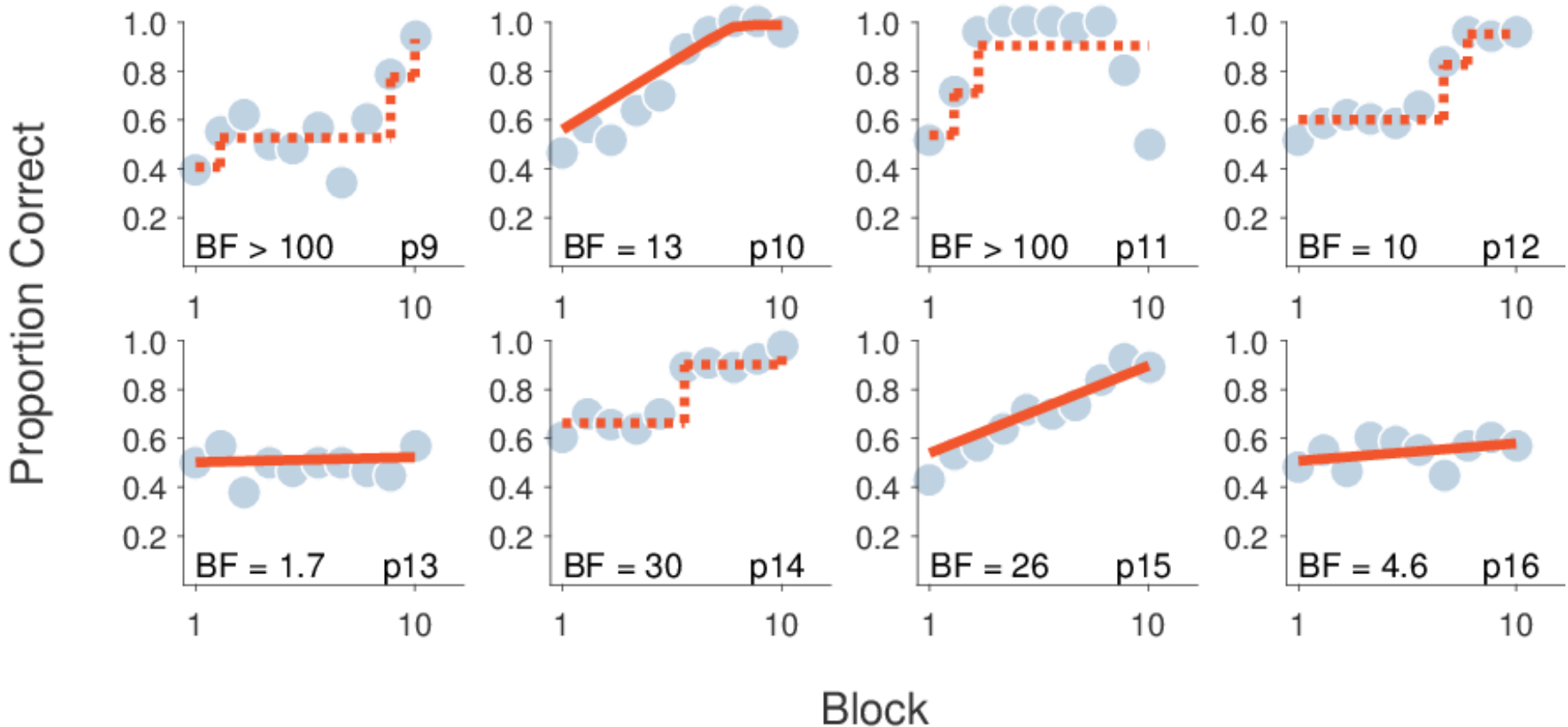
- Cognitive science is interested in whether people learn categories
  - Incrementally, by fine-tuning associations
  - Suddenly, by swapping in and out different candidate rules





# Results

- Find evidence for individual differences, with both incremental and step-change learning
  - can infer learning curves, and Bayes factors










# ranker.com NFL 2016-2017 MVP Votes

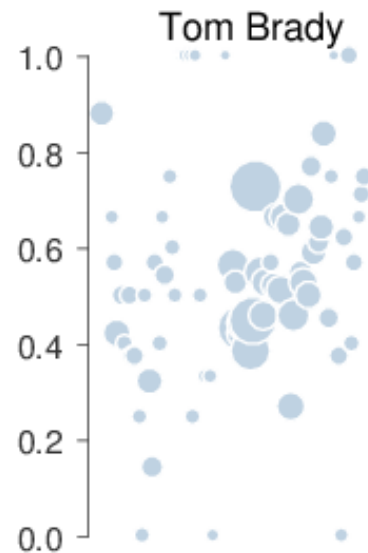
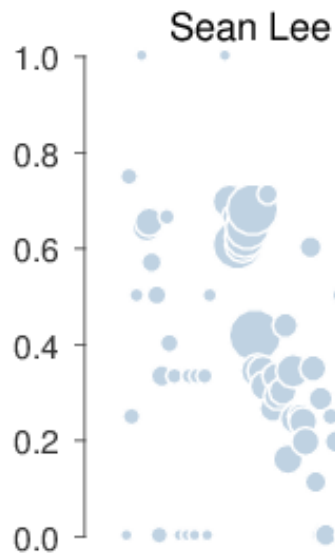
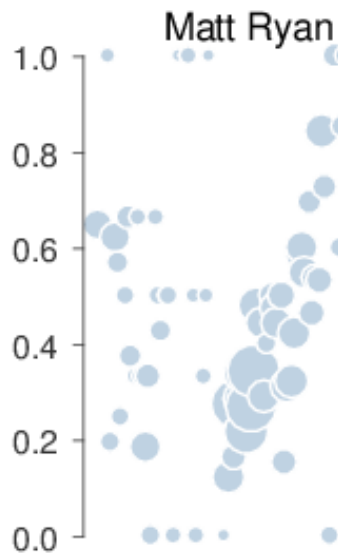
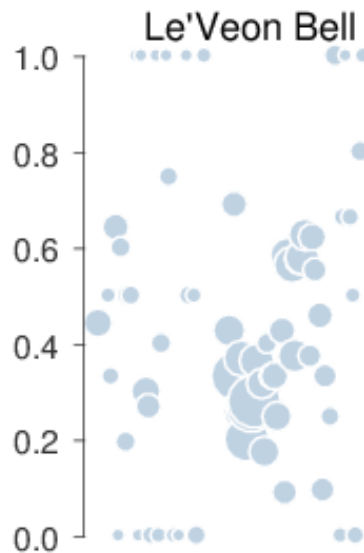
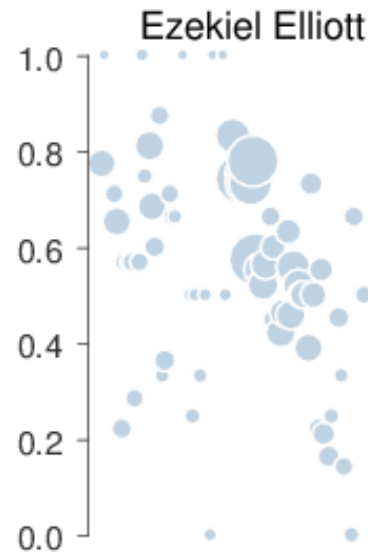
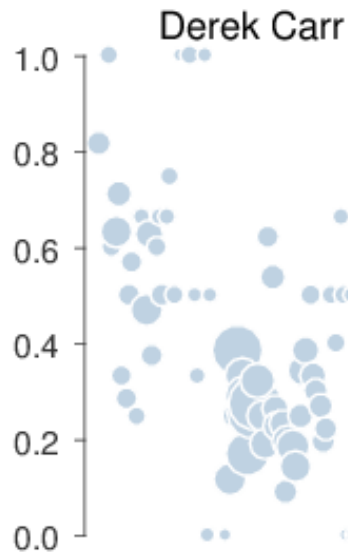
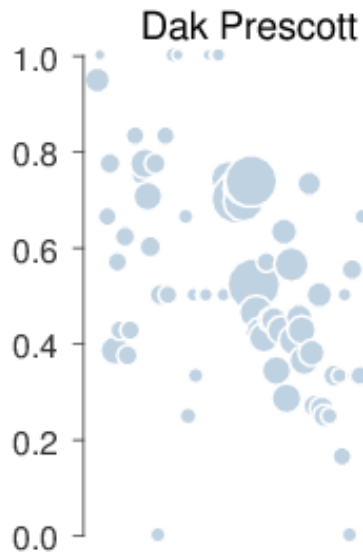
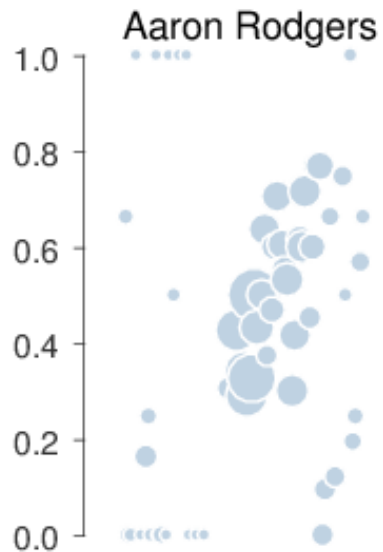
- List created in November 2016, and received 31,907 votes for 27 different players, up until Matt Ryan won on February 4, 2017

SEE LIST RANKED BY [WOMEN](#) [MEN](#) [AGE](#) [REGION](#) [5 RERANKS](#)

[OPTIONS](#) [COMMENTS](#) [EMBED](#) [rerank list](#)

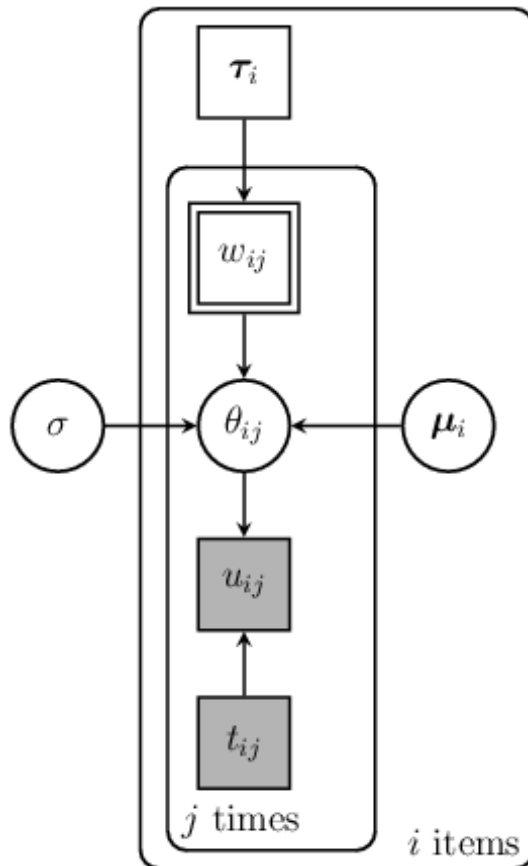
1	<a href="#">↑</a> 1,550	<a href="#">↓</a> 883		Ezekiel Elliott	▼
2	<a href="#">↑</a> 1,312	<a href="#">↓</a> 898		Dak Prescott age 22	▼
3	<a href="#">↑</a> 1,231	<a href="#">↓</a> 1,196		Tom Brady age 38	▼
4	<a href="#">↑</a> 700	<a href="#">↓</a> 1,013		<i>most listed</i> Matt Ryan age 30	▼
5	<a href="#">↑</a> 689	<a href="#">↓</a> 850		<i>added by Joshua Kennedy</i> Aaron Rodgers age 32	▼

# Voting Patterns for Eight Favorites



# Hierarchical Extension

- The rates within a stage come from an overarching Gaussian
  - assumed the variance is the same for all stages, coming from the different people in the crowd day-to-day



$$\mu_{ik} \sim \text{Uniform}(0, 1)$$

$$\sigma \sim \text{Gaussian}_{(0, \infty)}(\mu', 1/(\sigma')^2)$$

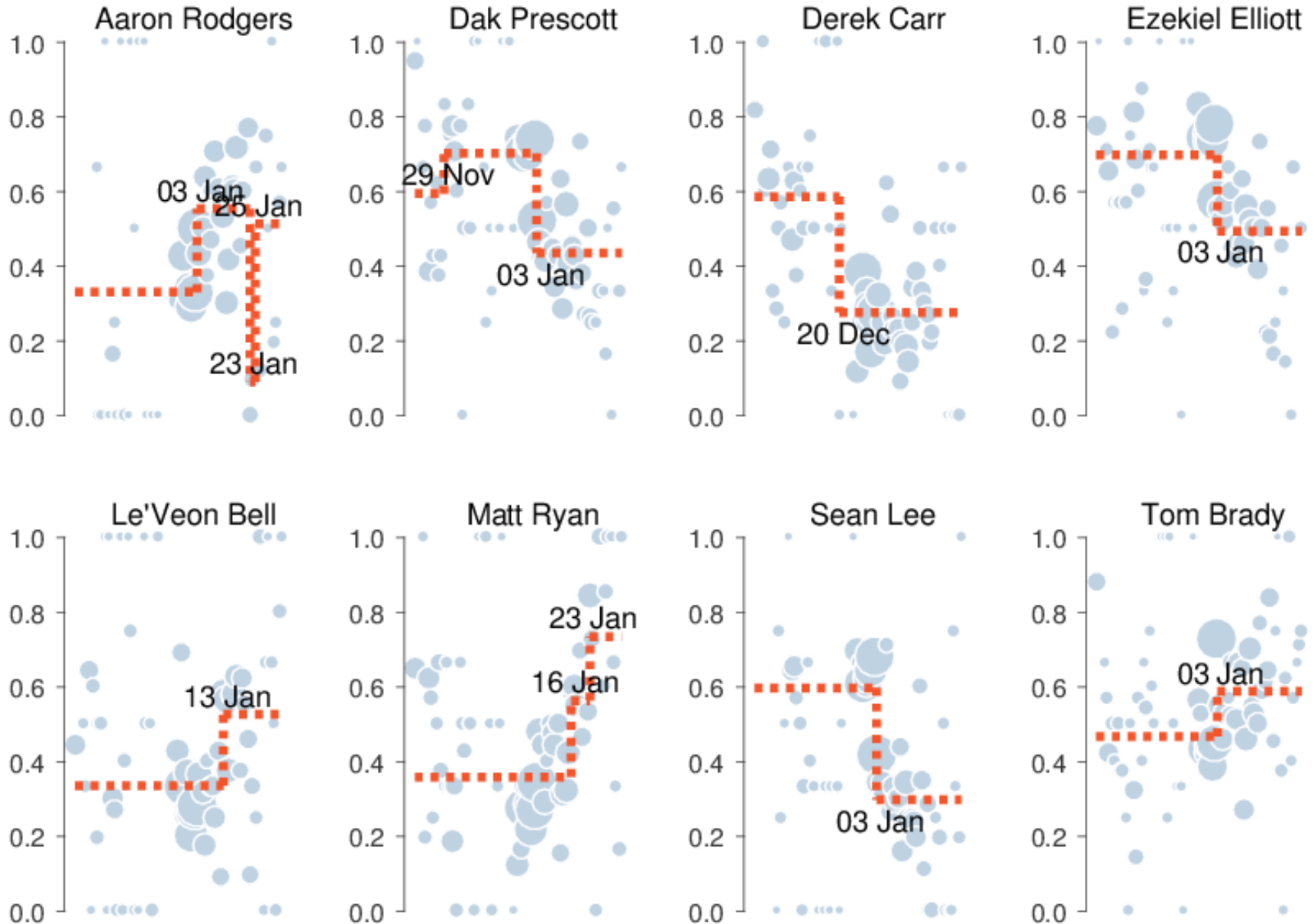
$$\tau_{ik} \sim \text{Categorical}\left(\overbrace{\frac{n}{2n-1}, \frac{1}{2n-1}, \dots, \frac{1}{2n-1}}^{n \text{ times}}\right); \tau_1 \leq \dots \leq \tau_\gamma$$

$$w_{ij} = \sum_k \mathcal{I}(j \geq \tau_{ik})$$

$$\theta_{ij} \sim \text{Gaussian}_{(0,1)}(\mu_i, w_{ij}, 1/\sigma^2)$$

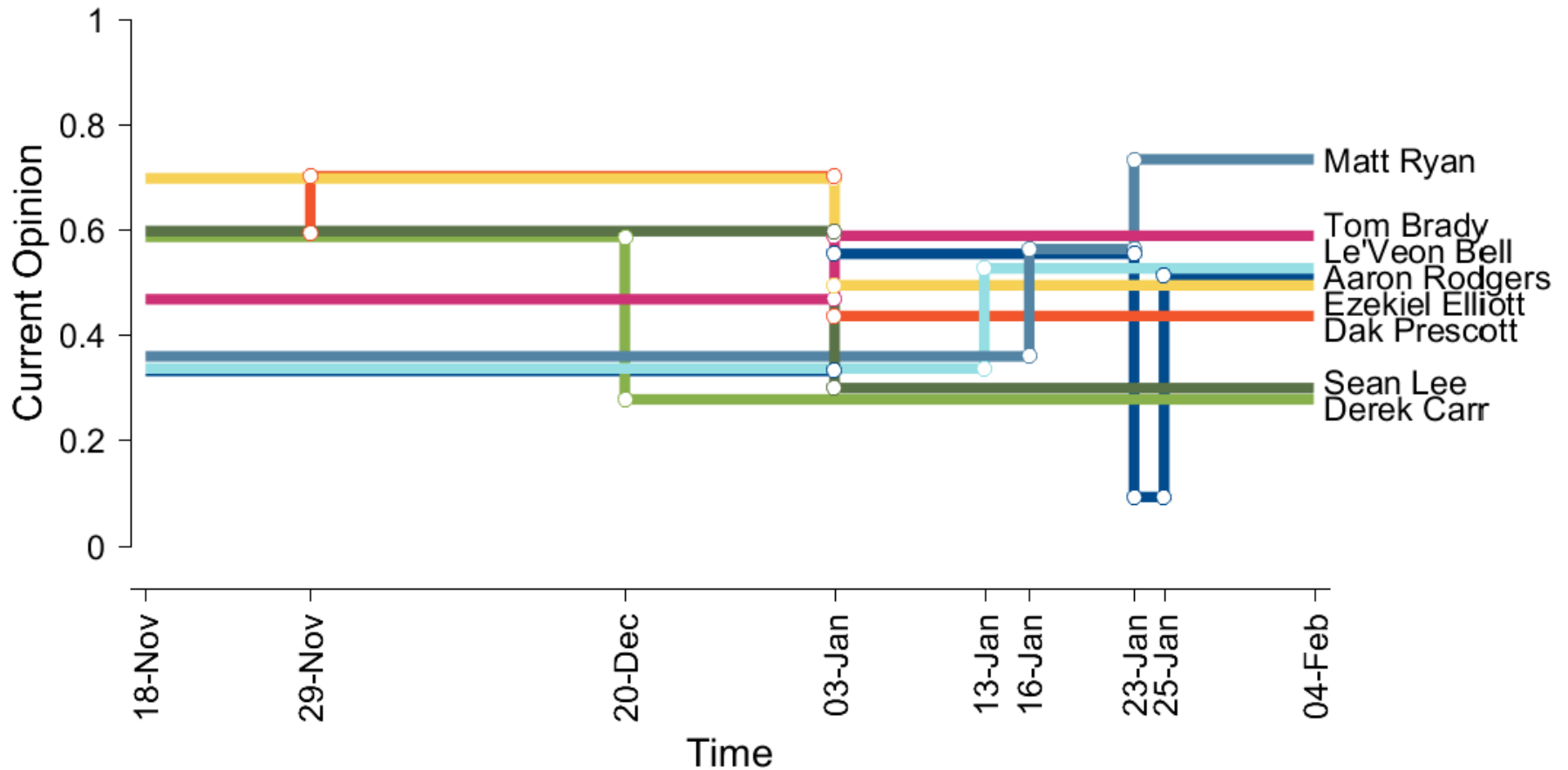
$$u_{ij} \sim \text{Binomial}(\theta_{ij}, t_{ij})$$

# Inferred Step Changes in Opinions



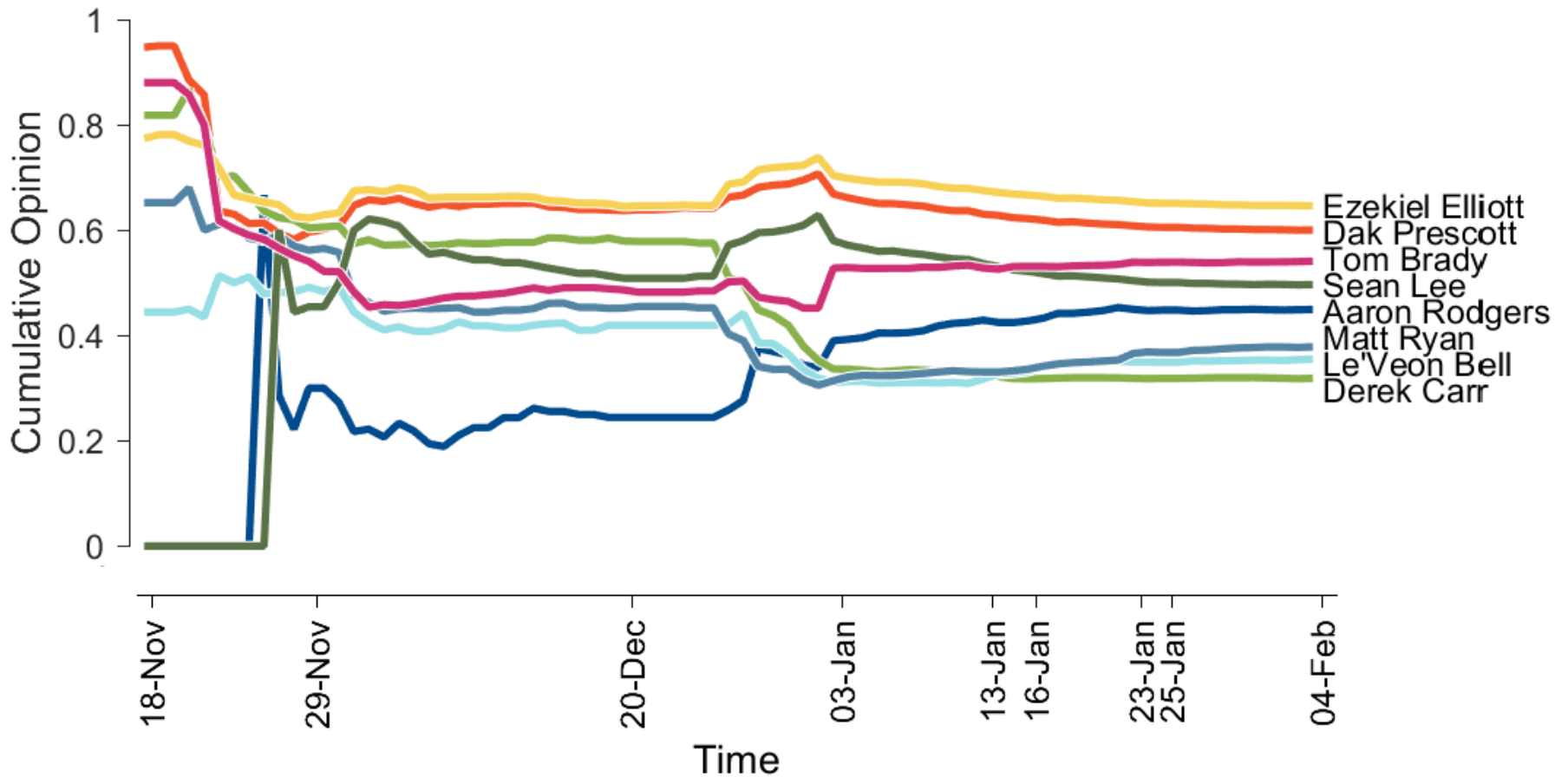
# Current Opinion

- At the time of voting, Matt Ryan has the best up-vote proportion, ahead of Tom Brady



# Cumulative Opinion

- Ezekiel Elliott had the greatest proportion of up-votes overall

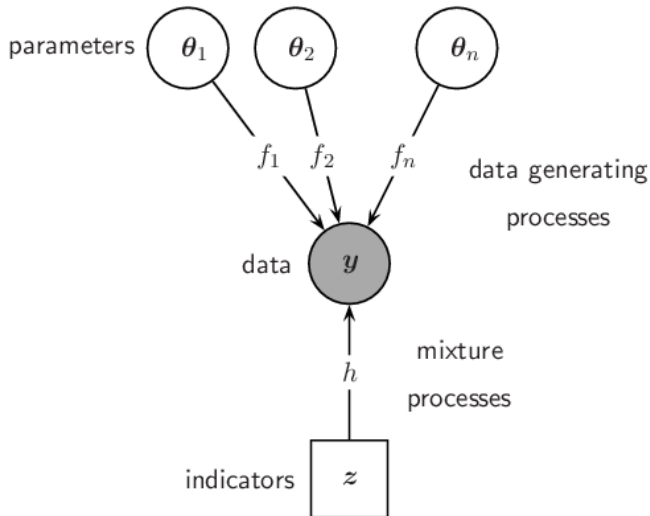


# Summary

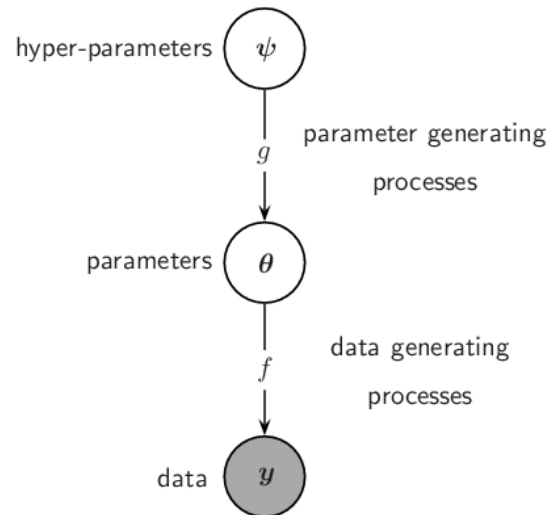
# Bayesian Benefits for Cognitive Modeling

- Bayesian methods allow theorists to develop, evaluate, and use richer generative models of how psychological variables and processes generate behavior
- Bayesian methods afford **theoretical freedom** with **rigorous assessment** and **flexible inferences**

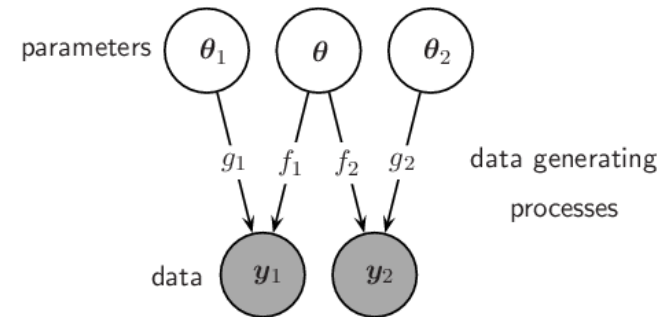
Latent Mixture



Hierarchical



Common Cause





# Collaborators



Ravi Iyer  
Ranker



Ravi Selker  
Jamovi



Irina Danileiko  
UC Irvine



Maime Guan  
UC Irvine



Joachim Van...  
UC Irvine



Kensuke Okada  
U Tokyo



Lucy Wu  
UC Irvine



Megan Lee  
UC Irvine