

# The Hierarchical Bayesian Inference Model

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# When do we need the hierarchical Bayesian inference model?

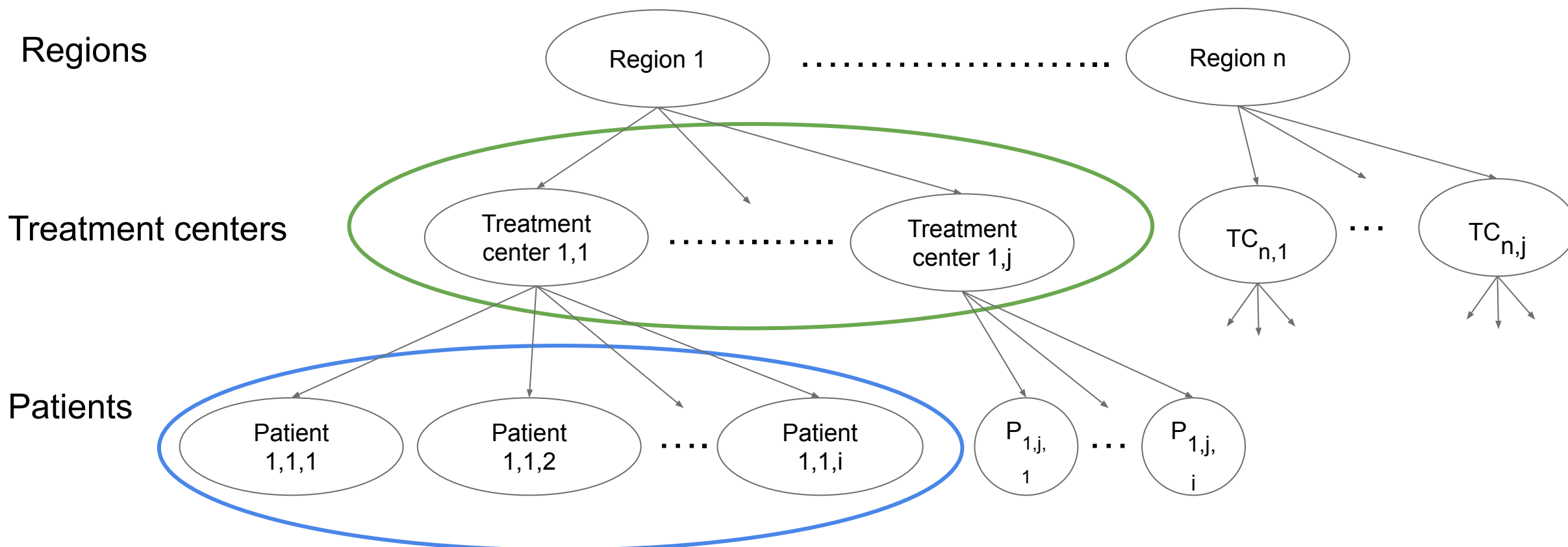
→ Hierarchical data set

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## CLINICAL TRIAL

Patients clustered within treatment centers nested within regions



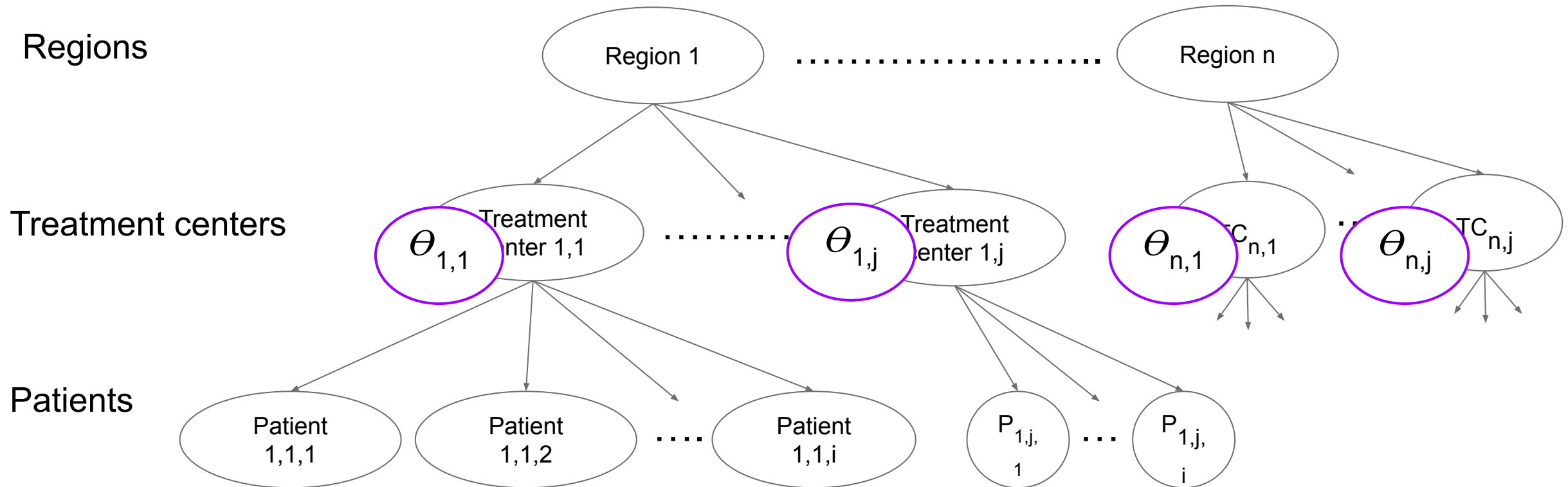
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→ Hierarchical data set

Parameter of interest:  
Survival probability  $\theta_{n,j}$

## CLINICAL TRIAL

Patients clustered within treatment centers nested within regions



# How does the HBI model differ from 'ordinary' Bayesian statistics?

## 'Ordinary' Bayes

$$\underbrace{P(\theta|y)}_{\text{Posterior}} \propto \underbrace{P(y|\theta)}_{\text{Likelihood}} \underbrace{P(\theta)}_{\text{Prior}}$$

$$y|\theta \sim P(y|\theta)$$

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## HBI model

$$y|\theta, \phi \sim P(y|\theta, \phi)$$

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$$P(\theta|y) \propto P(y|\theta)P(\theta)$$

⏟     ⏟     ⏟  
 Posterior   Likelihood   Prior

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## HBI model

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⏟     ⏟     ⏟     ⏟  
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# Testing the HBI model

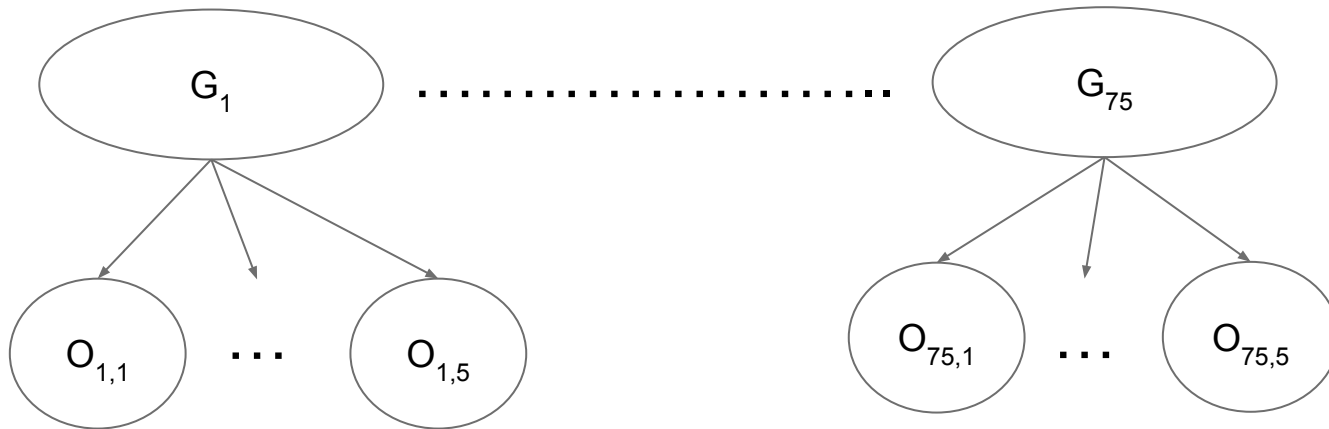
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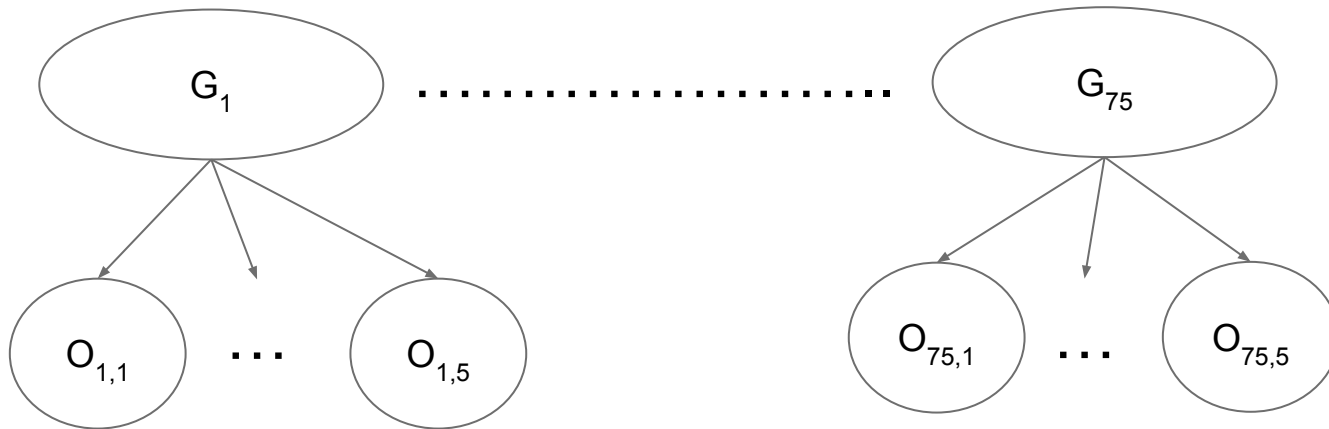
- 75 groups ( $j \in \{1, 75\}$ )
- 5 observations ( $i \in \{1, 5\}$ ) in each group



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Data

$$y_{ij} | \alpha_j, \beta_j \sim N(\alpha_j + \beta_j x_{ij}, \sigma^2)$$

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$$\sigma \sim U(0, 10)$$

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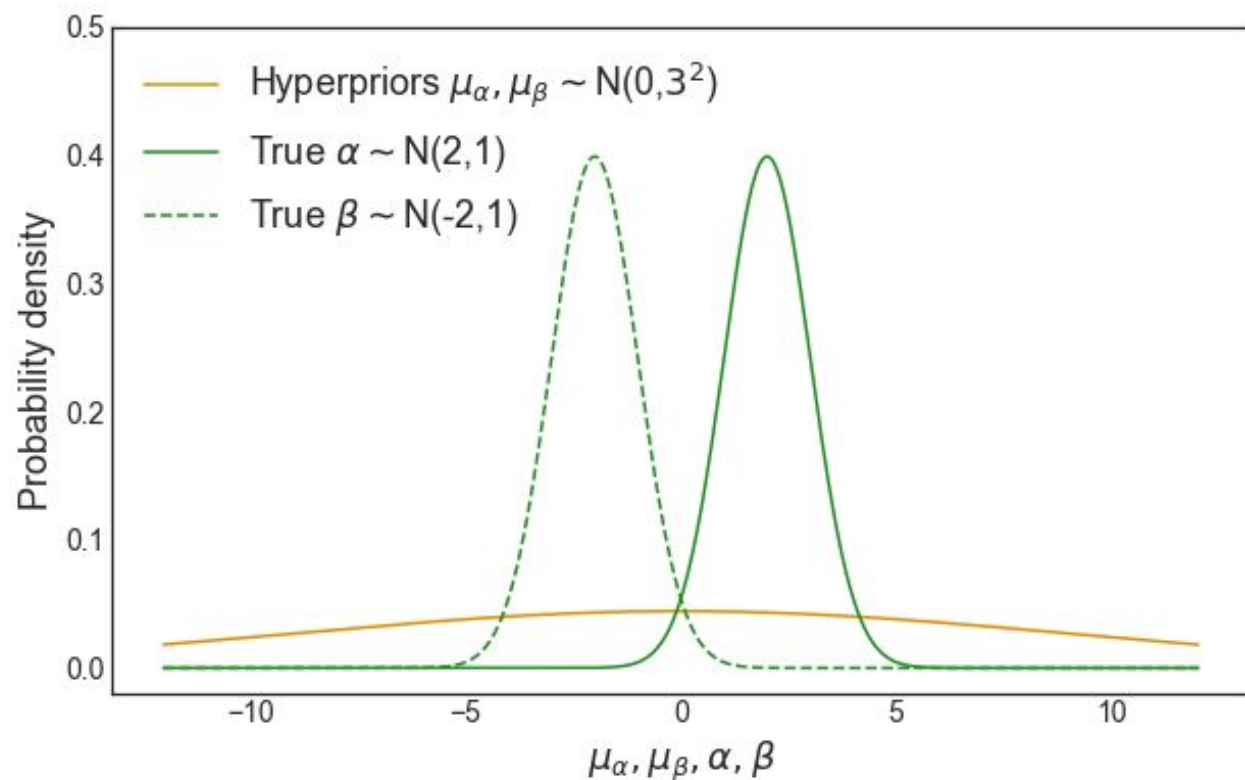
Hyperparameters

$$\mu_\alpha, \mu_\beta \sim N(0, 3^2)$$

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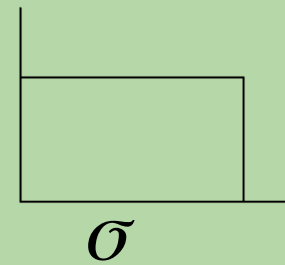
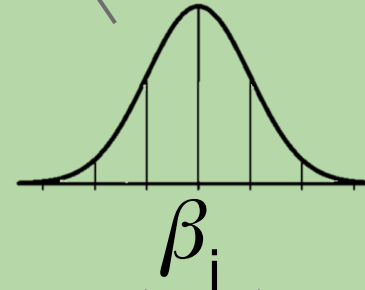
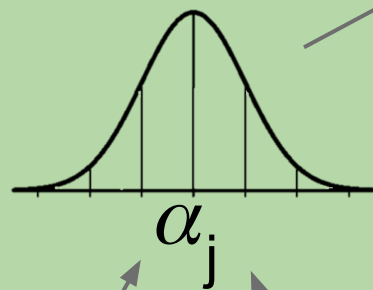
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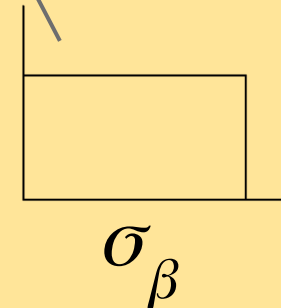
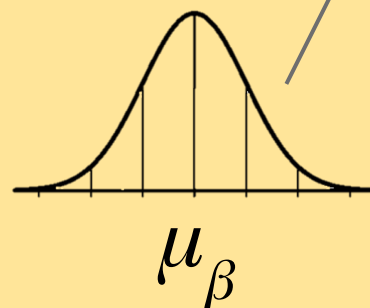
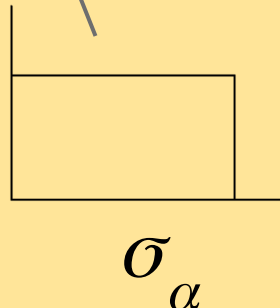
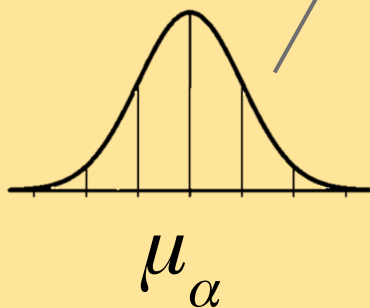
Data

$$y_{ij} = N(\alpha_j + \beta_j x_{ij}, \sigma^2)$$

Parameters



Hyperparameters



# Comparing the HBI model to other models

→ Overview of alternative models

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## 1) Completely pooling



Assume that  $\alpha$  and  $\beta$  are equal for all groups.



Corresponds to running a single regression on the whole data set



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→ Overview of alternative models

## 1) Completely pooling



Assume that  $\alpha$  and  $\square$  are equal for all groups.



Corresponds to running a single regression on the whole data set

## 2) No pooling



Assign each  $\alpha$  and  $\square$  their own wide, flat prior.



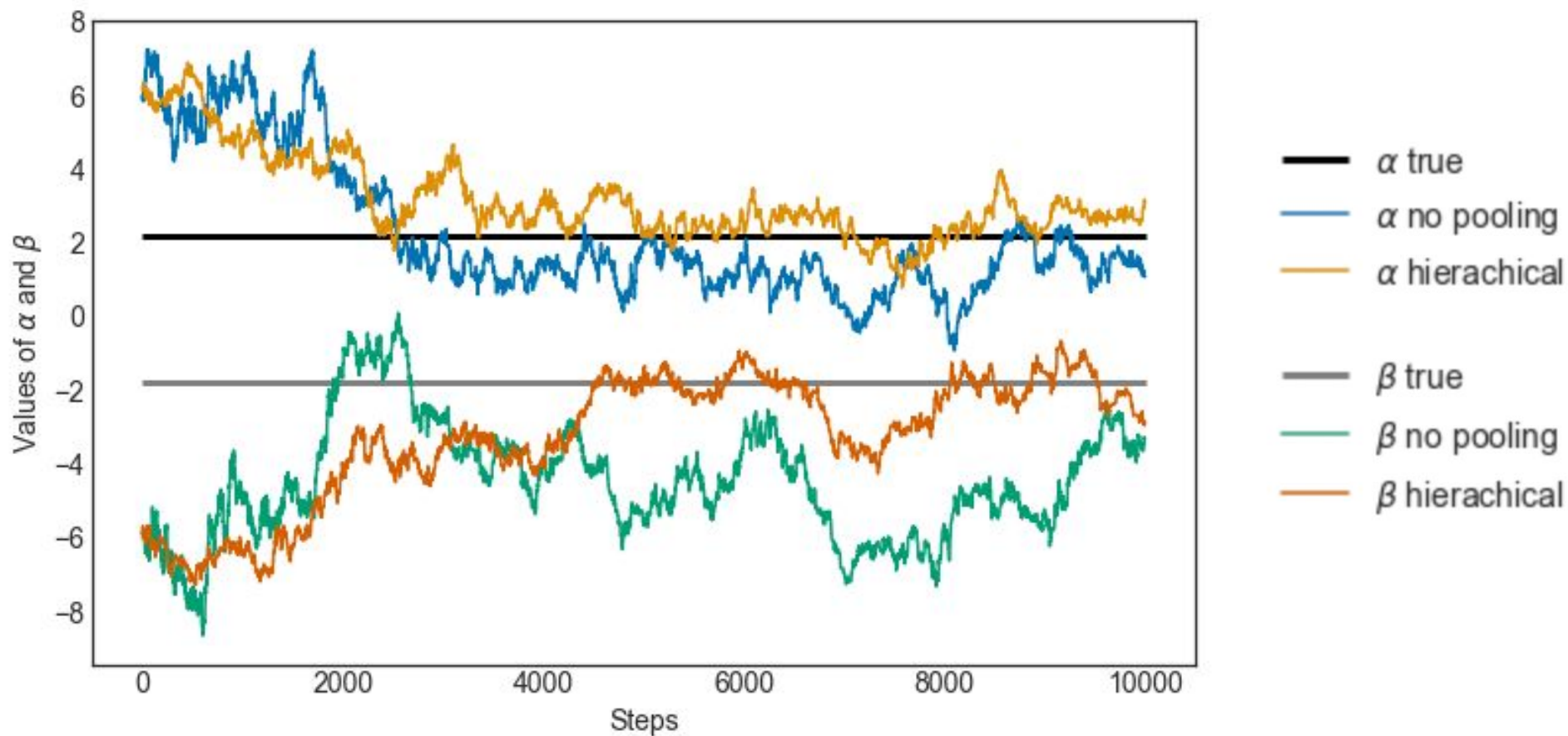
Corresponds to running a single regression on each group.

## Comparing the HBI model to other models

→ Markov Chain Metropolis-Hastings sampler: Obtaining estimates of  $\alpha$  and  $\square$

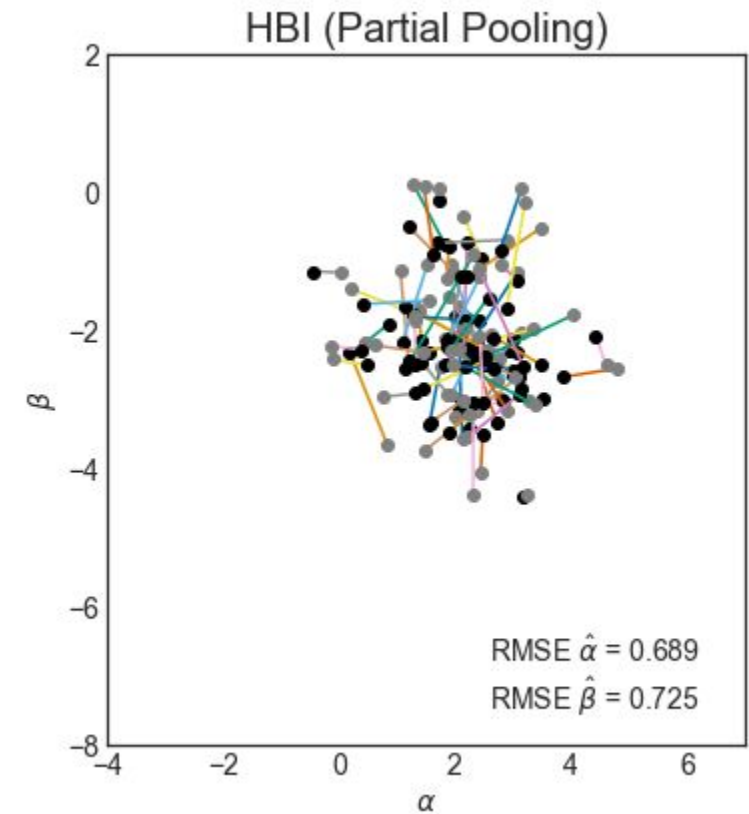
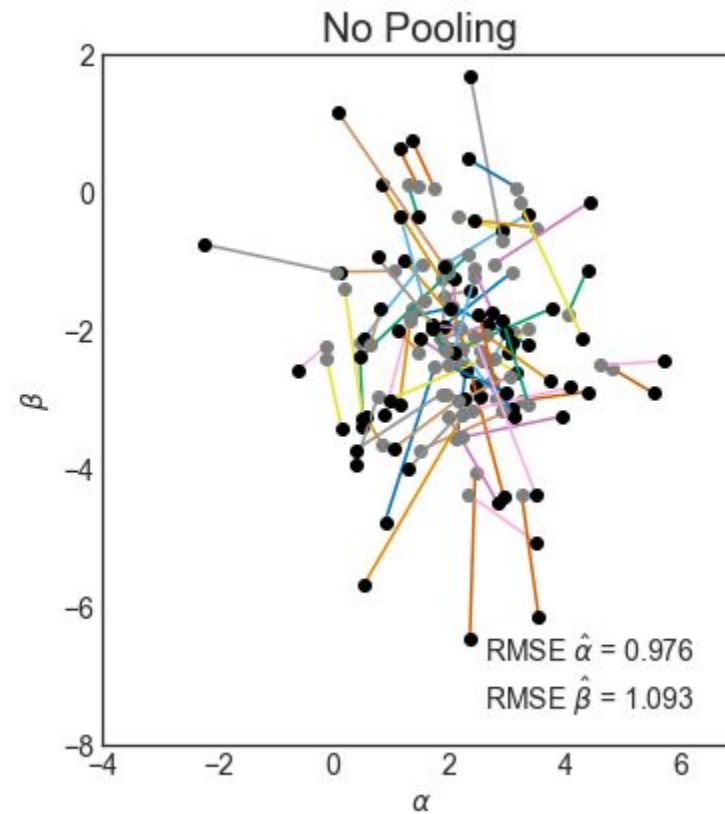
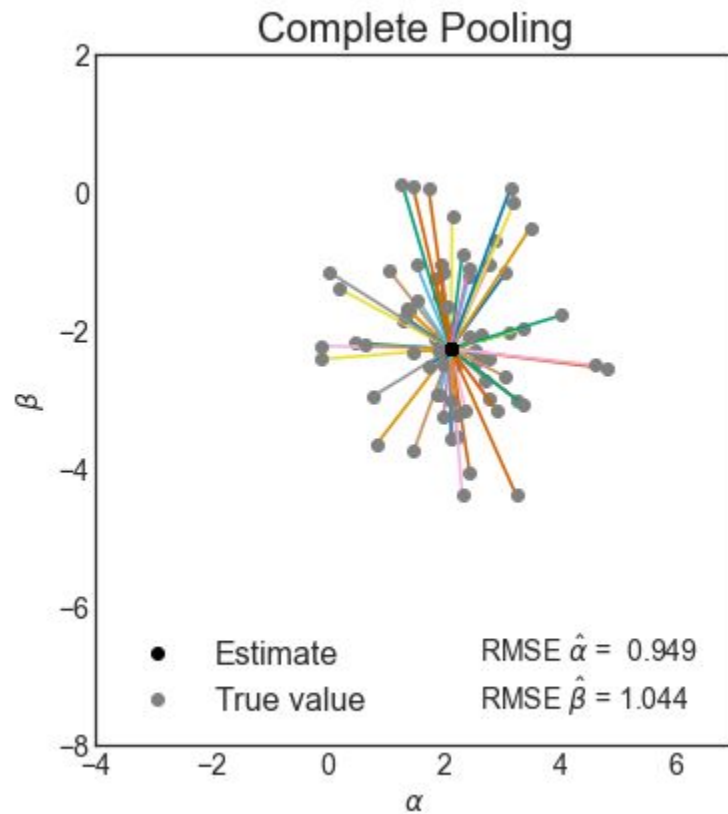
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→ Markov Chain Metropolis-Hastings sampler: Obtaining estimates of  $\alpha$  and  $\beta$



# Comparing the HBI model to other models

→ Comparing estimates of  $\alpha$  and  $\beta$  to the true values



## Summarizing: Hierarchical Bayesian Inference Model

*The HBI model provides a flexible framework for statistical modeling that can capture **variability across groups** and **improve the accuracy and precision of the parameter estimates***

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*The HBI model provides a flexible framework for statistical modeling that can capture **variability across groups** and **improve the accuracy and precision of the parameter estimates***

- Allows sharing of statistical strengths between the groups of data by assuming parameters come from common distributions
- Lets the data shape the prior itself by introducing the hyperprior

## Summarizing: Hierarchical Bayesian Inference Model

*The HBI model provides a flexible framework for statistical modeling that can capture **variability across groups** and **improve the accuracy and precision of the parameter estimates***

- Allows sharing of statistical strengths between the groups of data by assuming parameters come from common distributions
- The prior is affected by the data itself by introducing the hyperprior
- Estimates are less sensitive to noise as the prior structure pulls the estimates towards the population distribution (shrinkage)



**Thank you for you attention :)**