

# Hierarchical Walk-through

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Tim Frasier

# Data

- Milk composition and brain characteristics for a range of primates.

clade	kcal.per.g
Strepsirrhine (Strep)	0.49
New World Monkey (NWM)	0.80
Old World Monkey (OWM)	0.73
Ape	0.48



# Model

```
model {  
  // Definitions  
  vector[N] mu;  
  
  // Likelihood  
  for (i in 1:N) {  
    mu[i] = b1[x[i]];  
    y[i] ~ normal(mu[i], sigma[x[i]]);  
  }  
  
  // Priors  
  for (j in 1:nxLevels) {  
    b1[j] ~ normal(shopMean, shopMeanSD);  
    sigma[j] ~ cauchy(1, 1);  
  }  
  
  // Hyperpriors  
  shopMean ~ normal(0, 1);  
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Stan starts by selecting values, from the prior,  
for parameters with defined priors  
(the hyperpriors).

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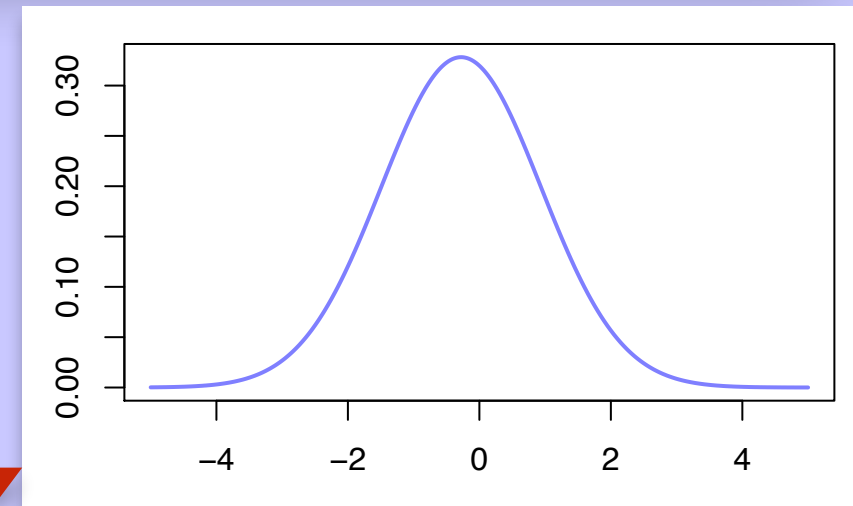
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shopMean = -0.2784659  
shopMeanSD = 1.215533

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These values are then used to create a prior value for the mean effect of each level in our nominal variable.

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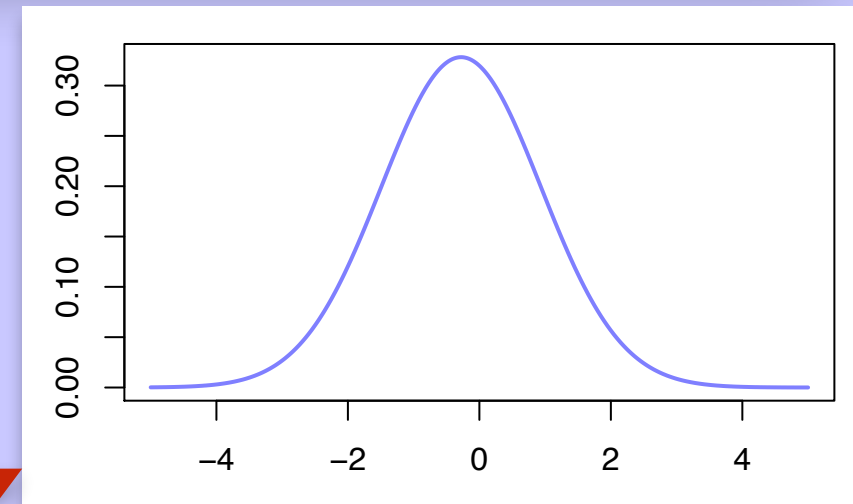


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<code>b1[Strep]</code>	<code>= -0.6652136</code>
<code>b1[NWM]</code>	<code>= -0.786263</code>
<code>b1[OWM]</code>	<code>= -2.408034</code>
<code>b1[Ape]</code>	<code>= -2.666315</code>



# Model

Sigma values for each level drawn from defined prior.

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sigma[Strep]	=	1.321518
sigma[NWM]	=	0.1529242
sigma[OWM]	=	0.1786416
sigma[Ape]	=	17.08555

b1[Strep]	=	-0.6652136
b1[NWM]	=	-0.786263
b1[OWM]	=	-2.408034
b1[Ape]	=	-2.666315

# Model

Calculate the likelihood of the data given these values for each parameter.

```
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  vector[N] mu;  
  
  // Likelihood  
  for (i in 1:N) {  
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# Model

- Propose new values for the parameters in the same manner
- Calculate the likelihood of the data with these new values
  - If higher, accept new values
  - If not, follow some rules to decide whether or not to accept new values



# Model

Will result in posterior probability distributions for the mean and s.d. for the effects of being in each group...

```
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  // Likelihood  
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    y[i] ~ normal(mu[i], sigma[x[i]]);  
  }  
  
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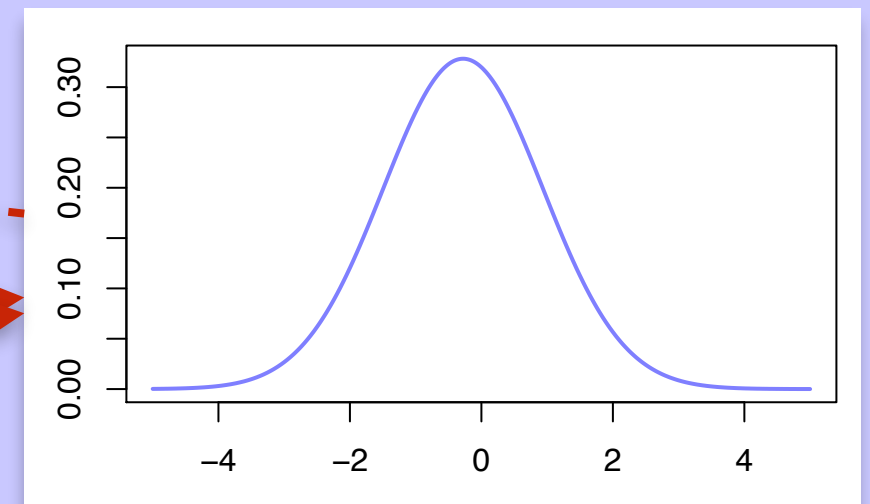
b1[Strep]  
b1[NWM]  
b1[OWM]  
b1[Ape]

sigma[Strep]  
sigma[NWM]  
sigma[OWM]  
sigma[Ape]

# Model

...as well as the mean and s.d. describing the normal distribution from which all effects (across all levels) arose.

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**Does that help?**