# Statistical Rethinking: Chapter 4 - Linear Models

## Chris Hayduk

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## 1 Easy

#### Problem 4E1.

In the model definition below, which line is the likelihood?

- 1.  $y_i \sim Normal(\mu, \sigma)$
- 2.  $\mu \sim \text{Normal}(0, 10)$
- 3.  $\sigma \sim \text{Uniform}(0, 10)$

Line 1 represents the likelihood.

#### Problem 4E2.

In the model definition just above, how many parameters are in the posterior distribution? There are two parameters:  $\mu$  and  $\sigma$ .

#### Problem 4E3.

Using the model definition above, write down the appropriate form of Bayes' theorem that includes the proper likelihood and priors.

$$Pr(\mu,\,\sigma|y) = \ \frac{\Pi_i Normal(y_i|\mu,\,\sigma)\ Normal(\mu|0,\,10)\ Uniform(\sigma|0,\,10)}{\int\!\!\int\!\!\Pi_i Normal(y_i|\mu,\,\sigma)\ Normal(\mu|0,\,10)\ Uniform(\sigma|0,\,10)\ d\mu d\sigma}$$

#### Problem 4E4.

In the model definition below, which line is the linear model?

- 1.  $y_i \sim Normal(\mu, \sigma)$
- 2.  $\mu_i = \alpha + \beta x_i$
- 3.  $\alpha \sim \text{Normal}(0, 10)$
- 4.  $\beta \sim \text{Normal}(0, 10)$
- 5.  $\sigma \sim \text{Uniform}(0, 10)$

Line 2 represents the linear model.

#### Problem 4E5.

In the model definition just above, how many parameters are in the posterior distribution?

There are three parameters:  $\alpha$ ,  $\beta$ , and  $\sigma$ .

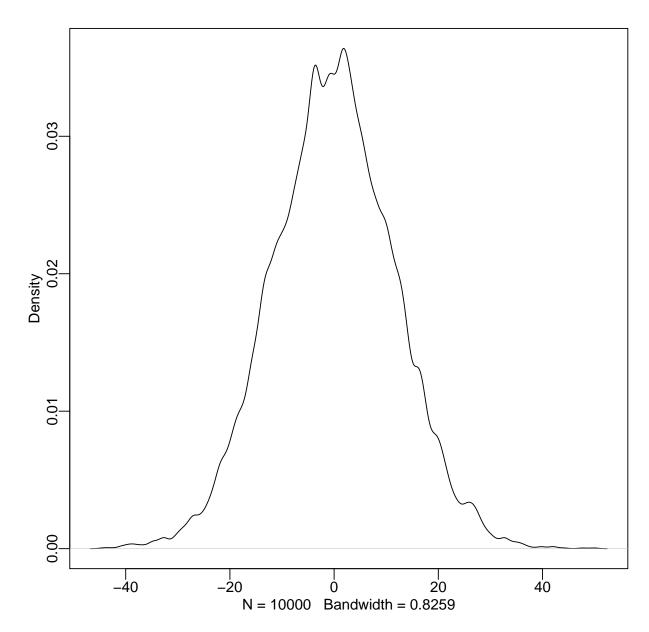
## 2 Medium

### Problem 4M1.

For the model definition below, simulate the observed heights from the prior (not the posterior).

```
\begin{aligned} y_i &\sim Normal(\mu,\,\sigma) \\ \mu &\sim Normal(0,\,10) \\ \sigma &\sim Uniform(0,\,10) \end{aligned}
```

```
sample_mu <- rnorm(1e4, 0, 10)
sample_sigma <- runif(1e4, 0, 10)
prior_h <- rnorm(1e4, sample_mu, sample_sigma)
dens(prior_h)</pre>
```



## Problem 4M2.

Translate the model just above into a map formula.

```
flist <- alist(
   y ~ dnorm(mu, sigma),
   mu ~ dnorm(0, 10),
   sigma ~ dunif(0, 10)
)</pre>
```

## Problem 4M3.

Translate the map model formula below into a mathematical model definition.

```
flist <- alist(
   y ~ dnorm(mu, sigma),
   mu <- a +b*x,
   a ~ dnorm(0, 50),
   b ~ dunif(0, 10),
   sigma ~ dunif(0, 50)
)</pre>
```

Model:

$$y_i \sim \text{Normal}(\mu, \sigma)$$
  
 $\mu_i = \alpha + \beta x_i$   
 $\alpha \sim \text{Normal}(0, 50)$   
 $\beta \sim \text{Uniform}(0, 10)$   
 $\sigma \sim \text{Uniform}(0, 50)$ 

#### Problem 4M4.

A sample of students is measured for height each year for 3 years. After the third year, you want to fit a linear regression predicting height using year as a predictor. Write down the mathematical model for this regression, using any variable names and priors you choose. Be prepared to defend your choice of priors.

Model:

$$y_i \sim \text{Normal}(\mu, \sigma)$$
  
 $\mu_i = \alpha + \beta x_i$   
 $\alpha \sim \text{Normal}(152, 25)$   
 $\beta \sim \text{Normal}(6, 3)$   
 $\sigma \sim \text{Uniform}(0, 50)$ 

The priors for  $\alpha$  represent an average of about 5 feet (152 cm) and a standard deviation of about 10 inches (25 cm). The priors for  $\beta$  represent an average increase of about 2.4 inches per year (6 cm - chosen using average growth rate for children) and a standard deviation of about 1.19 inches (3 cm). The prior for  $\sigma$  is a uniform prior between 0 and 50.

#### Problem 4M5.

Now suppose I tell you that the average height in the first year was 120 cm and that every student got taller each year and every student got taller each year. Does this information lead you to change your choice of prior?

With this no information, we should adjust  $\alpha$ . We now have:

$$y_i \sim \text{Normal}(\mu, \sigma)$$
  
 $\mu_i = \alpha + \beta x_i$   
 $\alpha \sim \text{Normal}(120, 25)$   
 $\beta \sim \text{Normal}(6, 3)$   
 $\sigma \sim \text{Uniform}(0, 50)$ 

The priors for  $\alpha$  have been adjusted to account for the new information we have about the average height in the first year. We already chose  $\beta$  as a positive number with a relatively small standard deviation, so the information about students growing taller each year does not effect our choice of prior.

#### Problem 4M6.

Now suppose I tell you that the variance among heights for students of the same age is never more than 64 cm. How does this lead you to revise your priors?

 $\sigma$  is just the square root of the variance. Thus, we would like  $\beta$ 's standard deviation to be less than  $\sqrt{64} = 8$ . Since we already chose a standard deviation of 3 cm, we do not need to revise this choice.