

STATS 205P HW3

Chuqi Wang 79167724

2024-05-20

Q1:

```
library(readr)
getwd()
```

```
## [1] "/Users/chuqiwan/Desktop/UCI/STATS205P/hw3"
```

```
setwd("/Users/chuqiwan/Desktop/UCI/STATS205P/hw3")
data = read.csv("kid-iq.csv")
```

(a)

```
# Load necessary libraries
library(rstan)
```

```
## Loading required package: StanHeaders
```

```
##
```

```
## rstan version 2.32.6 (Stan version 2.32.2)
```

```
## For execution on a local, multicore CPU with excess RAM we recommend calling
```

```
## options(mc.cores = parallel::detectCores()).
```

```
## To avoid recompilation of unchanged Stan programs, we recommend calling
```

```
## rstan_options(auto_write = TRUE)
```

```
## For within-chain threading using `reduce_sum()` or `map_rect()` Stan functions,
```

```
## change `threads_per_chain` option:
```

```
## rstan_options(threads_per_chain = 1)
```

```
# Define the Stan model for Model 1: kidScore ~ momHs + momIq
```

```
stan_model1 <- "
```

```
data {
  int<lower=0> N;           // number of observations
  vector[N] kidScore;      // dependent variable
  vector[N] momHs;         // independent variable momHs
  vector[N] momIq;         // independent variable momIq
}
```

```
parameters {
  real alpha;              // intercept
  real beta_momHs;         // coefficient for momHs
  real beta_momIq;         // coefficient for momIq
  real<lower=0> sigma;      // error scale
}
```

```
model {
  // Priors
  alpha ~ normal(0, 1000);
```

```

beta_momHs ~ normal(0, 1000);
beta_momIq ~ normal(0, 1000);
sigma ~ scaled_inv_chi_square(1, 0.05);

// Likelihood
kidScore ~ normal(alpha + beta_momHs * momHs + beta_momIq * momIq, sigma);
}
"

# Define the Stan model for Model 2: kidScore ~ momHs + momIq + momAge
stan_model2 <- "
data {
  int<lower=0> N;          // number of observations
  vector[N] kidScore;     // dependent variable
  vector[N] momHs;        // independent variable momHs
  vector[N] momIq;        // independent variable momIq
  vector[N] momAge;       // independent variable momAge
}
parameters {
  real alpha;             // intercept
  real beta_momHs;        // coefficient for momHs
  real beta_momIq;        // coefficient for momIq
  real beta_momAge;       // coefficient for momAge
  real<lower=0> sigma;    // error scale
}
model {
  // Priors
  alpha ~ normal(0, 1000);
  beta_momHs ~ normal(0, 1000);
  beta_momIq ~ normal(0, 1000);
  beta_momAge ~ normal(0, 1000);
  sigma ~ scaled_inv_chi_square(1, 0.05);

  // Likelihood
  kidScore ~ normal(alpha + beta_momHs * momHs + beta_momIq * momIq + beta_momAge * momAge, sigma);
}
"

# Prepare the data for Stan
stan_data1 <- list(
  N = nrow(data),
  kidScore = data$kidScore,
  momHs = data$momHs,
  momIq = data$momIq
)

stan_data2 <- list(
  N = nrow(data),
  kidScore = data$kidScore,
  momHs = data$momHs,
  momIq = data$momIq,
  momAge = data$momAge
)

```



```

## Chain 2: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 2: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 2: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 2: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 2: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 2: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 2: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 2: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 2:
## Chain 2: Elapsed Time: 0.721 seconds (Warm-up)
## Chain 2: 0.357 seconds (Sampling)
## Chain 2: 1.078 seconds (Total)
## Chain 2:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 3).
## Chain 3:
## Chain 3: Gradient evaluation took 1.2e-05 seconds
## Chain 3: 1000 transitions using 10 leapfrog steps per transition would take 0.12 seconds.
## Chain 3: Adjust your expectations accordingly!
## Chain 3:
## Chain 3:
## Chain 3: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 3: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 3: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 3: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 3: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 3: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 3: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 3: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 3: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 3: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 3: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 3: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 3:
## Chain 3: Elapsed Time: 0.622 seconds (Warm-up)
## Chain 3: 0.414 seconds (Sampling)
## Chain 3: 1.036 seconds (Total)
## Chain 3:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 4).
## Chain 4:
## Chain 4: Gradient evaluation took 1.2e-05 seconds
## Chain 4: 1000 transitions using 10 leapfrog steps per transition would take 0.12 seconds.
## Chain 4: Adjust your expectations accordingly!
## Chain 4:
## Chain 4:
## Chain 4: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 4: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 4: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 4: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 4: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 4: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 4: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 4: Iteration: 1200 / 2000 [ 60%] (Sampling)

```



```

## Chain 2: Gradient evaluation took 1.8e-05 seconds
## Chain 2: 1000 transitions using 10 leapfrog steps per transition would take 0.18 seconds.
## Chain 2: Adjust your expectations accordingly!
## Chain 2:
## Chain 2:
## Chain 2: Iteration:    1 / 2000 [  0%] (Warmup)
## Chain 2: Iteration:   200 / 2000 [ 10%] (Warmup)
## Chain 2: Iteration:   400 / 2000 [ 20%] (Warmup)
## Chain 2: Iteration:   600 / 2000 [ 30%] (Warmup)
## Chain 2: Iteration:   800 / 2000 [ 40%] (Warmup)
## Chain 2: Iteration:  1000 / 2000 [ 50%] (Warmup)
## Chain 2: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 2: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 2: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 2: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 2: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 2: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 2:
## Chain 2: Elapsed Time: 1.264 seconds (Warm-up)
## Chain 2:                0.908 seconds (Sampling)
## Chain 2:                2.172 seconds (Total)
## Chain 2:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 3).
## Chain 3:
## Chain 3: Gradient evaluation took 1.7e-05 seconds
## Chain 3: 1000 transitions using 10 leapfrog steps per transition would take 0.17 seconds.
## Chain 3: Adjust your expectations accordingly!
## Chain 3:
## Chain 3:
## Chain 3: Iteration:    1 / 2000 [  0%] (Warmup)
## Chain 3: Iteration:   200 / 2000 [ 10%] (Warmup)
## Chain 3: Iteration:   400 / 2000 [ 20%] (Warmup)
## Chain 3: Iteration:   600 / 2000 [ 30%] (Warmup)
## Chain 3: Iteration:   800 / 2000 [ 40%] (Warmup)
## Chain 3: Iteration:  1000 / 2000 [ 50%] (Warmup)
## Chain 3: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 3: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 3: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 3: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 3: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 3: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 3:
## Chain 3: Elapsed Time: 1.249 seconds (Warm-up)
## Chain 3:                0.916 seconds (Sampling)
## Chain 3:                2.165 seconds (Total)
## Chain 3:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 4).
## Chain 4:
## Chain 4: Gradient evaluation took 1.7e-05 seconds
## Chain 4: 1000 transitions using 10 leapfrog steps per transition would take 0.17 seconds.
## Chain 4: Adjust your expectations accordingly!
## Chain 4:

```

```

## Chain 4:
## Chain 4: Iteration:    1 / 2000 [  0%] (Warmup)
## Chain 4: Iteration:   200 / 2000 [ 10%] (Warmup)
## Chain 4: Iteration:   400 / 2000 [ 20%] (Warmup)
## Chain 4: Iteration:   600 / 2000 [ 30%] (Warmup)
## Chain 4: Iteration:   800 / 2000 [ 40%] (Warmup)
## Chain 4: Iteration:  1000 / 2000 [ 50%] (Warmup)
## Chain 4: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 4: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 4: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 4: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 4: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 4: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 4:
## Chain 4: Elapsed Time: 1.159 seconds (Warm-up)
## Chain 4:                0.915 seconds (Sampling)
## Chain 4:                2.074 seconds (Total)
## Chain 4:

```

```
fit1
```

```

## Inference for Stan model: anon_model.
## 4 chains, each with iter=2000; warmup=1000; thin=1;
## post-warmup draws per chain=1000, total post-warmup draws=4000.
##
##               mean se_mean   sd      2.5%      25%      50%      75%      97.5%
## alpha          25.64     0.15 5.74     14.20     21.86     25.80     29.54     36.57
## beta_momHs       5.93     0.04 2.15       1.68       4.48       5.92       7.38     10.20
## beta_momIq       0.57     0.00 0.06       0.45       0.53       0.56       0.60       0.68
## sigma          18.14     0.01 0.61     17.04     17.71     18.11     18.56     19.37
## lp__          -1476.41     0.03 1.34    -1479.74    -1477.08    -1476.13    -1475.41    -1474.69
##               n_eff Rhat
## alpha          1461    1
## beta_momHs     2519    1
## beta_momIq     1453    1
## sigma          2668    1
## lp__           1592    1
##
## Samples were drawn using NUTS(diag_e) at Tue May 21 19:48:08 2024.
## For each parameter, n_eff is a crude measure of effective sample size,
## and Rhat is the potential scale reduction factor on split chains (at
## convergence, Rhat=1).

```

```
fit2
```

```

## Inference for Stan model: anon_model.
## 4 chains, each with iter=2000; warmup=1000; thin=1;
## post-warmup draws per chain=1000, total post-warmup draws=4000.
##
##               mean se_mean   sd      2.5%      25%      50%      75%      97.5%
## alpha          20.65     0.20 9.18       2.81     14.41     20.74     26.93     38.27
## beta_momHs       5.62     0.04 2.30       1.20       4.02       5.58       7.21     10.23
## beta_momIq       0.56     0.00 0.06       0.44       0.52       0.56       0.60       0.68
## beta_momAge       0.24     0.01 0.33      -0.40       0.02       0.25       0.46       0.90
## sigma          18.16     0.01 0.64     16.95     17.72     18.14     18.57     19.46
## lp__          -1476.80     0.04 1.60    -1480.63    -1477.60    -1476.47    -1475.62    -1474.69

```

```
##           n_eff Rhat
## alpha      2158    1
## beta_momHs 3230    1
## beta_momIq 2533    1
## beta_momAge 2453    1
## sigma      3397    1
## lp__       1632    1
##
## Samples were drawn using NUTS(diag_e) at Tue May 21 19:48:37 2024.
## For each parameter, n_eff is a crude measure of effective sample size,
## and Rhat is the potential scale reduction factor on split chains (at
## convergence, Rhat=1).
```

For model 1 and 2 we set priors to be as follow:

$$\sigma^2 \sim \text{Inv} - \chi^2(1, 0.5)$$

$$\beta_j \sim N(0, 1000)$$

By applying MCMC, for M1, we found that the estimated model is given by:

$$\widehat{\text{kidScore}} = 25.93 + 5.96\text{momHs} + 0.56\text{momIq}$$

The posterior distribution of M1 are given by:

$$\beta_0 \sim N(25.93, 6.04^2)$$

$$\beta_1 \sim N(5.96, 2.21^2)$$

$$\beta_2 \sim N(0.56, 0.06^2)$$

$$\sigma^2 \sim \text{Inv} - \chi^2(18.17, 0.62^2)$$

For M2, we found that the estimated model is given by:

$$\widehat{\text{kidScore}} = 20.92 + 5.59\text{momHs} + 0.56\text{momIq} + 0.23\text{momAge}$$

The posterior distribution of M2 are given by:

$$\beta_0 \sim N(20.92, 6.04^2)$$

$$\beta_1 \sim N(5.59, 2.31^2)$$

$$\beta_2 \sim N(0.56, 0.06^2)$$

$$\beta_3 \sim N(0.23, 0.34^2)$$

$$\sigma^2 \sim \text{Inv} - \chi^2(18.16, 0.62^2)$$

(b)

```
stan_model1 <- "
data {
  int<lower=0> N;
  vector[N] kidScore;
  vector[N] momHs;
  vector[N] momIq;
}
parameters {
  real alpha;
  real beta_momHs;
```



```

    real beta_momIq;
    real<lower=0> sigma;
  }
  model {
    alpha ~ normal(0, 1000);
    beta_momHs ~ normal(0, 1000);
    beta_momIq ~ normal(0, 1000);
    sigma ~ scaled_inv_chi_square(1, 0.05);

    kidScore ~ normal(alpha + beta_momHs * momHs + beta_momIq * momIq, sigma);
  }
  generated quantities {
    vector[N] y_rep;
    for (n in 1:N)
      y_rep[n] = normal_rng(alpha + beta_momHs * momHs[n] + beta_momIq * momIq[n], sigma);
  }
  "

stan_model2 <- "
data {
  int<lower=0> N;
  vector[N] kidScore;
  vector[N] momHs;
  vector[N] momIq;
  vector[N] momAge;
}
parameters {
  real alpha;
  real beta_momHs;
  real beta_momIq;
  real beta_momAge;
  real<lower=0> sigma;
}
model {
  alpha ~ normal(0, 1000);
  beta_momHs ~ normal(0, 1000);
  beta_momIq ~ normal(0, 1000);
  beta_momAge ~ normal(0, 1000);
  sigma ~ scaled_inv_chi_square(1, 0.05);

  kidScore ~ normal(alpha + beta_momHs * momHs + beta_momIq * momIq + beta_momAge * momAge, sigma);
}
generated quantities {
  vector[N] y_rep;
  for (n in 1:N)
    y_rep[n] = normal_rng(alpha + beta_momHs * momHs[n] + beta_momIq * momIq[n] + beta_momAge * momAge[n], sigma);
}
"

fit1 <- stan(model_code = stan_model1, data = stan_data1)

```

```
## Trying to compile a simple C file
```

```
## Running /Library/Frameworks/R.framework/Resources/bin/R CMD SHLIB foo.c
```

```
## using C compiler: 'Apple clang version 14.0.3 (clang-1403.0.22.14.1)'
```

```

## using SDK: 'MacOSX13.3.sdk'
## clang -arch arm64 -I"/Library/Frameworks/R.framework/Resources/include" -DNDEBUG -I"/Library/Frameworks/R.framework/Resources/include"
## In file included from <built-in>:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/library/StanHeaders/include/StanHeaders/StanHeaders.h:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/library/RcppEigen/include/Eigen/src/Core/Matrix.h:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/library/RcppEigen/include/Eigen/src/Core/MatrixBase.h:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/library/RcppEigen/include/Eigen/src/Core/Matrix.h:1:
## #include <cmath>
## ~~~~~
## 1 error generated.
## make: *** [foo.o] Error 1
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 1).
## Chain 1:
## Chain 1: Gradient evaluation took 5.1e-05 seconds
## Chain 1: 1000 transitions using 10 leapfrog steps per transition would take 0.51 seconds.
## Chain 1: Adjust your expectations accordingly!
## Chain 1:
## Chain 1:
## Chain 1: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 1: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 1: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 1: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 1: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 1: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 1: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 1: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 1: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 1: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 1: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 1: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 1:
## Chain 1: Elapsed Time: 0.648 seconds (Warm-up)
## Chain 1: 0.392 seconds (Sampling)
## Chain 1: 1.04 seconds (Total)
## Chain 1:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 2).
## Chain 2:
## Chain 2: Gradient evaluation took 1.3e-05 seconds
## Chain 2: 1000 transitions using 10 leapfrog steps per transition would take 0.13 seconds.
## Chain 2: Adjust your expectations accordingly!
## Chain 2:
## Chain 2:
## Chain 2: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 2: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 2: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 2: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 2: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 2: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 2: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 2: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 2: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 2: Iteration: 1600 / 2000 [ 80%] (Sampling)

```

```

## Chain 2: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 2: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 2:
## Chain 2: Elapsed Time: 0.679 seconds (Warm-up)
## Chain 2: 0.431 seconds (Sampling)
## Chain 2: 1.11 seconds (Total)
## Chain 2:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 3).
## Chain 3:
## Chain 3: Gradient evaluation took 1.7e-05 seconds
## Chain 3: 1000 transitions using 10 leapfrog steps per transition would take 0.17 seconds.
## Chain 3: Adjust your expectations accordingly!
## Chain 3:
## Chain 3:
## Chain 3: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 3: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 3: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 3: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 3: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 3: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 3: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 3: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 3: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 3: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 3: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 3: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 3:
## Chain 3: Elapsed Time: 0.674 seconds (Warm-up)
## Chain 3: 0.426 seconds (Sampling)
## Chain 3: 1.1 seconds (Total)
## Chain 3:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 4).
## Chain 4:
## Chain 4: Gradient evaluation took 1.3e-05 seconds
## Chain 4: 1000 transitions using 10 leapfrog steps per transition would take 0.13 seconds.
## Chain 4: Adjust your expectations accordingly!
## Chain 4:
## Chain 4:
## Chain 4: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 4: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 4: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 4: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 4: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 4: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 4: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 4: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 4: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 4: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 4: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 4: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 4:
## Chain 4: Elapsed Time: 0.689 seconds (Warm-up)

```



```

## Chain 2: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 2: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 2: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 2: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 2: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 2: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 2: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 2: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 2: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 2: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 2: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 2:
## Chain 2: Elapsed Time: 1.229 seconds (Warm-up)
## Chain 2: 0.922 seconds (Sampling)
## Chain 2: 2.151 seconds (Total)
## Chain 2:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 3).
## Chain 3:
## Chain 3: Gradient evaluation took 2.2e-05 seconds
## Chain 3: 1000 transitions using 10 leapfrog steps per transition would take 0.22 seconds.
## Chain 3: Adjust your expectations accordingly!
## Chain 3:
## Chain 3:
## Chain 3: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 3: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 3: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 3: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 3: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 3: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 3: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 3: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 3: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 3: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 3: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 3: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 3:
## Chain 3: Elapsed Time: 1.244 seconds (Warm-up)
## Chain 3: 0.928 seconds (Sampling)
## Chain 3: 2.172 seconds (Total)
## Chain 3:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 4).
## Chain 4:
## Chain 4: Gradient evaluation took 2e-05 seconds
## Chain 4: 1000 transitions using 10 leapfrog steps per transition would take 0.2 seconds.
## Chain 4: Adjust your expectations accordingly!
## Chain 4:
## Chain 4:
## Chain 4: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 4: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 4: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 4: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 4: Iteration: 800 / 2000 [ 40%] (Warmup)

```

```
## Chain 4: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 4: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 4: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 4: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 4: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 4: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 4: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 4:
## Chain 4: Elapsed Time: 1.238 seconds (Warm-up)
## Chain 4:           0.917 seconds (Sampling)
## Chain 4:           2.155 seconds (Total)
## Chain 4:
```

```
y_rep1 <- extract(fit1)$y_rep
y_rep2 <- extract(fit2)$y_rep
```

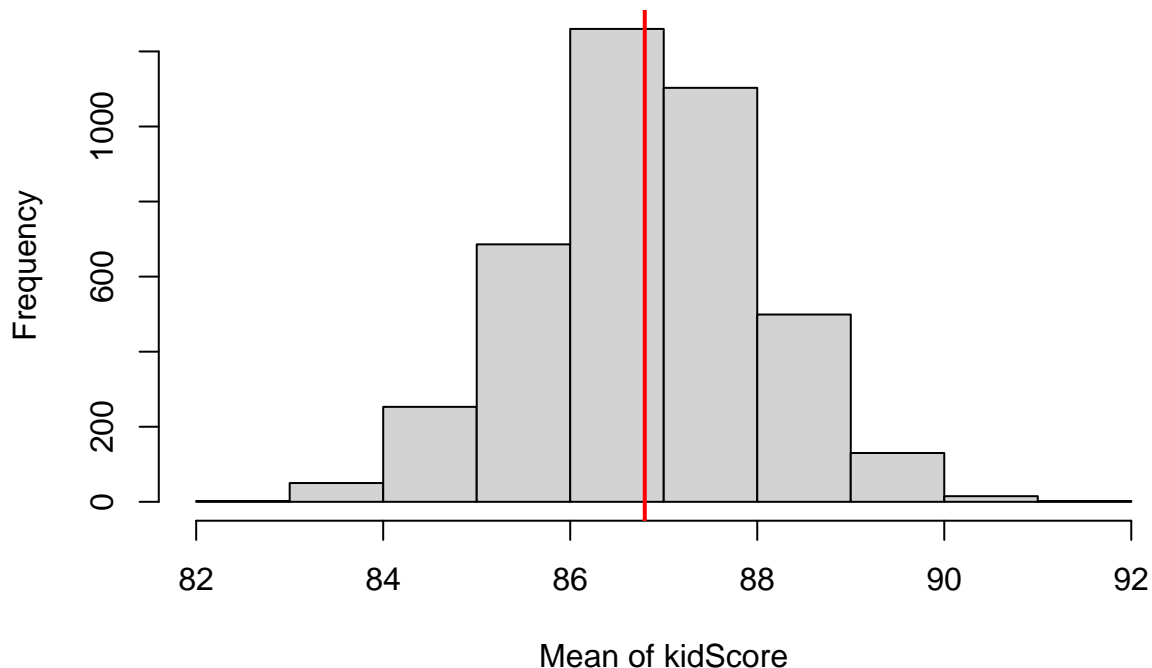
```
means1 <- apply(y_rep1, 1, mean)
means2 <- apply(y_rep2, 1, mean)
```

```
observed_mean <- mean(data$kidScore)
print(observed_mean)
```

```
## [1] 86.79724
```

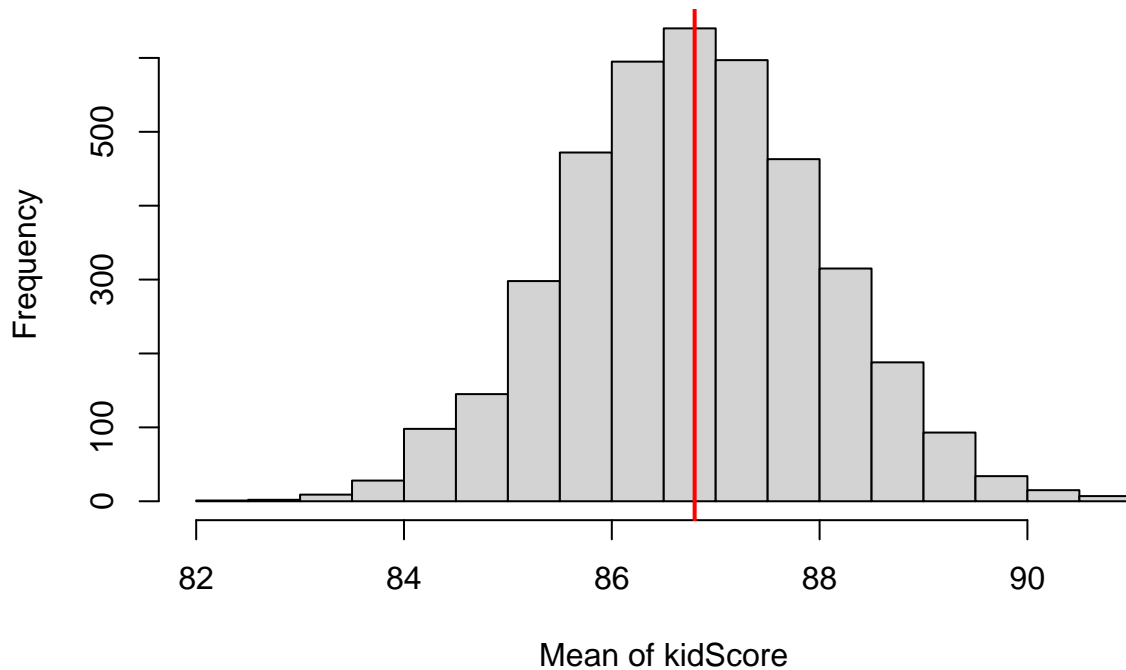
```
hist(means1, main="Model 1: Distribution of Simulated Means", xlab="Mean of kidScore")
abline(v=observed_mean, col="red", lwd=2)
```

Model 1: Distribution of Simulated Means



```
hist(means2, main="Model 2: Distribution of Simulated Means", xlab="Mean of kidScore")
abline(v=observed_mean, col="red", lwd=2)
```

Model 2: Distribution of Simulated Means



The observed mean of kidScore is 86.79724. For each model, by simulating 100 datasets, the distribution of simulated means for both M1 and M2 are shown above.

(c)

```
# Split the data into training and testing sets
train_data <- data[1:300, ]
test_data <- data[301:434, ]

stan_data1_train <- list(
  N = nrow(train_data),
  kidScore = train_data$kidScore,
  momHs = train_data$momHs,
  momIq = train_data$momIq
)

stan_data2_train <- list(
  N = nrow(train_data),
  kidScore = train_data$kidScore,
  momHs = train_data$momHs,
  momIq = train_data$momIq,
  momAge = train_data$momAge
)

# Define the Stan model for Model 1: kidScore ~ momHs + momIq
stan_model1 <- "
data {
  int<lower=0> N;           // number of observations
  vector[N] kidScore;      // dependent variable
  vector[N] momHs;         // independent variable momHs
  vector[N] momIq;         // independent variable momIq
```

```

}
parameters {
  real alpha;           // intercept
  real beta_momHs;      // coefficient for momHs
  real beta_momIq;      // coefficient for momIq
  real<lower=0> sigma;   // error scale
}
model {
  // Priors
  alpha ~ normal(0, 1000);
  beta_momHs ~ normal(0, 1000);
  beta_momIq ~ normal(0, 1000);
  sigma ~ scaled_inv_chi_square(1, 0.05);

  // Likelihood
  kidScore ~ normal(alpha + beta_momHs * momHs + beta_momIq * momIq, sigma);
}
"

# Define the Stan model for Model 2: kidScore ~ momHs + momIq + momAge
stan_model2 <- "
data {
  int<lower=0> N;           // number of observations
  vector[N] kidScore;     // dependent variable
  vector[N] momHs;        // independent variable momHs
  vector[N] momIq;        // independent variable momIq
  vector[N] momAge;       // independent variable momAge
}
parameters {
  real alpha;           // intercept
  real beta_momHs;      // coefficient for momHs
  real beta_momIq;      // coefficient for momIq
  real beta_momAge;     // coefficient for momAge
  real<lower=0> sigma;   // error scale
}
model {
  // Priors
  alpha ~ normal(0, 1000);
  beta_momHs ~ normal(0, 1000);
  beta_momIq ~ normal(0, 1000);
  beta_momAge ~ normal(0, 1000);
  sigma ~ scaled_inv_chi_square(1, 0.05);

  // Likelihood
  kidScore ~ normal(alpha + beta_momHs * momHs + beta_momIq * momIq + beta_momAge * momAge, sigma);
}
"

fit1 <- stan(model_code = stan_model1, data = stan_data1_train)

## Trying to compile a simple C file

## Running /Library/Frameworks/R.framework/Resources/bin/R CMD SHLIB foo.c
## using C compiler: 'Apple clang version 14.0.3 (clang-1403.0.22.14.1)'

```



```

## using SDK: 'MacOSX13.3.sdk'
## clang -arch arm64 -I"/Library/Frameworks/R.framework/Resources/include" -DNDEBUG -I"/Library/Frameworks/R.framework/Resources/include"
## In file included from <built-in>:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/library/StanHeaders/include/stan/math/prim/fun/abs.hpp:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/library/RcppEigen/include/Eigen/src/Core/Matrix.h:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/library/RcppEigen/include/Eigen/src/Core/MatrixBase.h:1:
## In file included from /Library/Frameworks/R.framework/Versions/4.3-arm64/Resources/library/RcppEigen/include/Eigen/src/Core/Matrix.h:1:
## #include <cmath>
## ~~~~~
## 1 error generated.
## make: *** [foo.o] Error 1
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 1).
## Chain 1:
## Chain 1: Gradient evaluation took 4.4e-05 seconds
## Chain 1: 1000 transitions using 10 leapfrog steps per transition would take 0.44 seconds.
## Chain 1: Adjust your expectations accordingly!
## Chain 1:
## Chain 1:
## Chain 1: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 1: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 1: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 1: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 1: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 1: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 1: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 1: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 1: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 1: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 1: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 1: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 1:
## Chain 1: Elapsed Time: 0.502 seconds (Warm-up)
## Chain 1: 0.274 seconds (Sampling)
## Chain 1: 0.776 seconds (Total)
## Chain 1:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 2).
## Chain 2:
## Chain 2: Gradient evaluation took 9e-06 seconds
## Chain 2: 1000 transitions using 10 leapfrog steps per transition would take 0.09 seconds.
## Chain 2: Adjust your expectations accordingly!
## Chain 2:
## Chain 2:
## Chain 2: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 2: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 2: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 2: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 2: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 2: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 2: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 2: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 2: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 2: Iteration: 1600 / 2000 [ 80%] (Sampling)

```

```

## Chain 2: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 2: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 2:
## Chain 2: Elapsed Time: 0.426 seconds (Warm-up)
## Chain 2: 0.258 seconds (Sampling)
## Chain 2: 0.684 seconds (Total)
## Chain 2:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 3).
## Chain 3:
## Chain 3: Gradient evaluation took 9e-06 seconds
## Chain 3: 1000 transitions using 10 leapfrog steps per transition would take 0.09 seconds.
## Chain 3: Adjust your expectations accordingly!
## Chain 3:
## Chain 3:
## Chain 3: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 3: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 3: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 3: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 3: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 3: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 3: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 3: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 3: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 3: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 3: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 3: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 3:
## Chain 3: Elapsed Time: 0.457 seconds (Warm-up)
## Chain 3: 0.249 seconds (Sampling)
## Chain 3: 0.706 seconds (Total)
## Chain 3:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 4).
## Chain 4:
## Chain 4: Gradient evaluation took 9e-06 seconds
## Chain 4: 1000 transitions using 10 leapfrog steps per transition would take 0.09 seconds.
## Chain 4: Adjust your expectations accordingly!
## Chain 4:
## Chain 4:
## Chain 4: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 4: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 4: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 4: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 4: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 4: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 4: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 4: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 4: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 4: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 4: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 4: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 4:
## Chain 4: Elapsed Time: 0.583 seconds (Warm-up)

```



```

## Chain 2: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 2: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 2: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 2: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 2: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 2: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 2: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 2: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 2: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 2: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 2: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 2:
## Chain 2: Elapsed Time: 0.886 seconds (Warm-up)
## Chain 2: 0.617 seconds (Sampling)
## Chain 2: 1.503 seconds (Total)
## Chain 2:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 3).
## Chain 3:
## Chain 3: Gradient evaluation took 1.3e-05 seconds
## Chain 3: 1000 transitions using 10 leapfrog steps per transition would take 0.13 seconds.
## Chain 3: Adjust your expectations accordingly!
## Chain 3:
## Chain 3:
## Chain 3: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 3: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 3: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 3: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 3: Iteration: 800 / 2000 [ 40%] (Warmup)
## Chain 3: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 3: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 3: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 3: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 3: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 3: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 3: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 3:
## Chain 3: Elapsed Time: 0.917 seconds (Warm-up)
## Chain 3: 0.689 seconds (Sampling)
## Chain 3: 1.606 seconds (Total)
## Chain 3:
##
## SAMPLING FOR MODEL 'anon_model' NOW (CHAIN 4).
## Chain 4:
## Chain 4: Gradient evaluation took 1.3e-05 seconds
## Chain 4: 1000 transitions using 10 leapfrog steps per transition would take 0.13 seconds.
## Chain 4: Adjust your expectations accordingly!
## Chain 4:
## Chain 4:
## Chain 4: Iteration: 1 / 2000 [ 0%] (Warmup)
## Chain 4: Iteration: 200 / 2000 [ 10%] (Warmup)
## Chain 4: Iteration: 400 / 2000 [ 20%] (Warmup)
## Chain 4: Iteration: 600 / 2000 [ 30%] (Warmup)
## Chain 4: Iteration: 800 / 2000 [ 40%] (Warmup)

```

```

## Chain 4: Iteration: 1000 / 2000 [ 50%] (Warmup)
## Chain 4: Iteration: 1001 / 2000 [ 50%] (Sampling)
## Chain 4: Iteration: 1200 / 2000 [ 60%] (Sampling)
## Chain 4: Iteration: 1400 / 2000 [ 70%] (Sampling)
## Chain 4: Iteration: 1600 / 2000 [ 80%] (Sampling)
## Chain 4: Iteration: 1800 / 2000 [ 90%] (Sampling)
## Chain 4: Iteration: 2000 / 2000 [100%] (Sampling)
## Chain 4:
## Chain 4: Elapsed Time: 0.812 seconds (Warm-up)
## Chain 4: 0.6 seconds (Sampling)
## Chain 4: 1.412 seconds (Total)
## Chain 4:

posterior1 <- extract(fit1)
posterior2 <- extract(fit2)

predictions1 <- mean(posterior1$alpha) + mean(posterior1$beta_momHs) * test_data$momHs +
  mean(posterior1$beta_momIq) * test_data$momIq
predictions2 <- mean(posterior2$alpha) + mean(posterior2$beta_momHs) * test_data$momHs +
  mean(posterior2$beta_momIq) * test_data$momIq + mean(posterior2$beta_momAge) * test_data$momAge

rmse1 <- sqrt(mean((predictions1 - test_data$kidScore)^2))
rmse2 <- sqrt(mean((predictions2 - test_data$kidScore)^2))

cat("RMSE for Model 1: ", rmse1, "\n")

## RMSE for Model 1: 19.48856
cat("RMSE for Model 2: ", rmse2, "\n")

## RMSE for Model 2: 19.48584

library(caret)

## Loading required package: ggplot2
## Loading required package: lattice
RMSE(predictions1, test_data$kidScore)

## [1] 19.48856
RMSE(predictions2, test_data$kidScore)

## [1] 19.48584

```

By splitting the data into 300 training data and 134 test data for two models, we found that the root mean squared error for model 1 is 19.48887 and for model 2 is 19.48533.