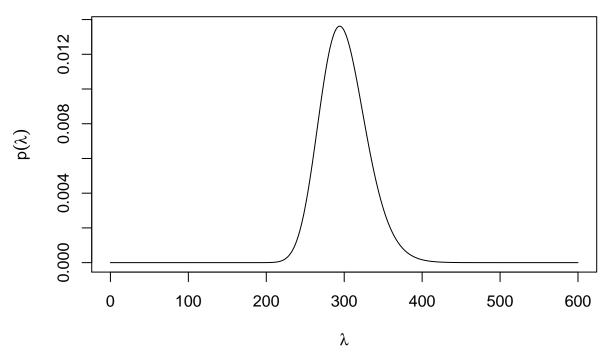
HW07 Report

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Problem 2b

```
library(MCMCpack)
## Loading required package: coda
## Loading required package: MASS
## ## Markov Chain Monte Carlo Package (MCMCpack)
## ## Copyright (C) 2003-2019 Andrew D. Martin, Kevin M. Quinn, and Jong Hee Park
## ##
## ## Support provided by the U.S. National Science Foundation
## ## (Grants SES-0350646 and SES-0350613)
## ##
## Specify the prior distribution: Inverse Gamma
m <- 300
         # Prior mean
v <- 900 # Prior variance
alpha = m^2 v + 2
beta = m * (alpha - 1)
alpha
## [1] 102
beta
## [1] 30300
## Plot the prior density
x \leftarrow seq(0.01, 600, length = 1000)
prior <- dinvgamma(x, shape = alpha, scale = beta) # the prior pdf
plot(x, prior, type = "l", xlab = expression(lambda),
    ylab = expression(p(lambda)), main = "Prior density")
```

Prior density



The choice of m and v is based on my current knowledge that, over a couple of years, there were around $\sim 4, 5$ earthquakes in Berkeley.

Problem 2c

```
## The data
load("BerkeleyEarthquakes.RData")
head(earthquakes)
##
                          Time
                                              Lon Depth Mag
                                    Lat
                                                                      Lag
        1969/04/14 04:22:36.21 37.8740 -122.2260 11.41 3.00
## 46
                                                                       NA
        1971/05/10 00:52:01.51 37.8863 -122.2432 9.51 3.15 755.8537650
                                                                3.0327751
        1971/05/13 01:39:13.28 37.8708 -122.2330 13.29 3.18
## 591
        1971/09/17 18:43:33.75 37.8692 -122.2345 11.36 3.05 127.7113480
## 722
## 723
        1971/09/17 21:34:35.44 37.8660 -122.2332 10.78 3.06
                                                                0.1187696
## 2184 1973/10/04 12:24:10.09 37.8643 -122.2335 9.85 3.10 747.6177622
data <- earthquakes$Lag[-1]</pre>
                                # First element is NA
n <- length(data)</pre>
                     # number of data points
s <- sum(data)
                  # sum of data points
## Update the parameters to get the posterior distribution
alpha_new <- alpha + n
beta_new <- beta + s
alpha_new
```

```
## [1] 168
beta_new
```

```
## [1] 44724.79

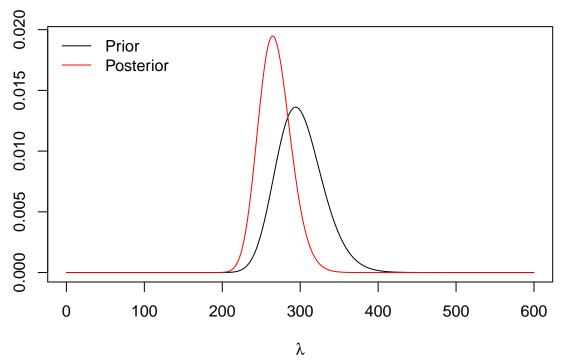
posterior <- dinvgamma(x, shape = alpha_new, scale = beta_new)

## Plot the prior and posterior densities

plot(x, prior, type = "l", xlab = expression(lambda), ylab = "",
        ylim = range(c(prior, posterior)))

lines(x, posterior, col = 2)

legend("topleft", lty = rep(1, 2), col = 1:2,
        legend = c("Prior", "Posterior"), bty = "n")</pre>
```



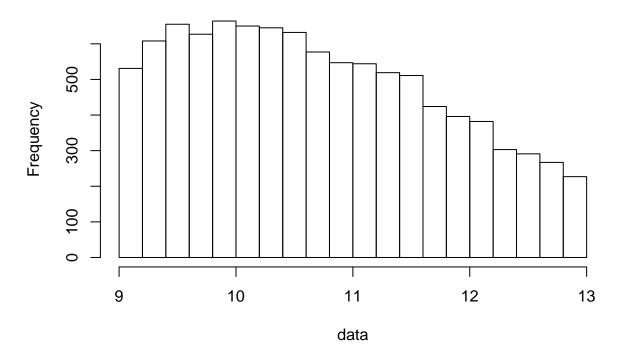
The data updates my knowledge towards reducing the mean and the variance because the data have smaller average time between earthquakes and smaller variance.

Problem 5c

```
N = 10000
mu = 10
sigma2 = 4
alpha = 9
beta = 13

data = rep(0, 1000)
i = 1
while (i <= N) {
    x = rnorm(1, mean = mu, sd = sigma2 ^ 0.5)
    if (x > alpha & x < beta) {
        data[i] = x
        i = i + 1
    }
}
hist(data, breaks = 20)</pre>
```

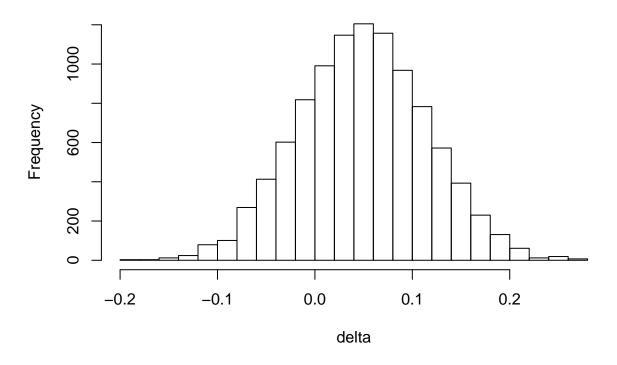
Histogram of data



Problem 6b

```
N = 10000
p1 = rbeta(N, shape1 = 34, shape2 = 68)
p2 = rbeta(N, shape1 = 39, shape2 = 63)
delta = p2 - p1
hist(delta, breaks = 20)
```

Histogram of delta



Problem 6c

```
# Posterior probability that delta > 0.04
p = sum(delta > 0.04) / N
p
```

[1] 0.5538