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## Problem 1

### Task a

Hand written solution

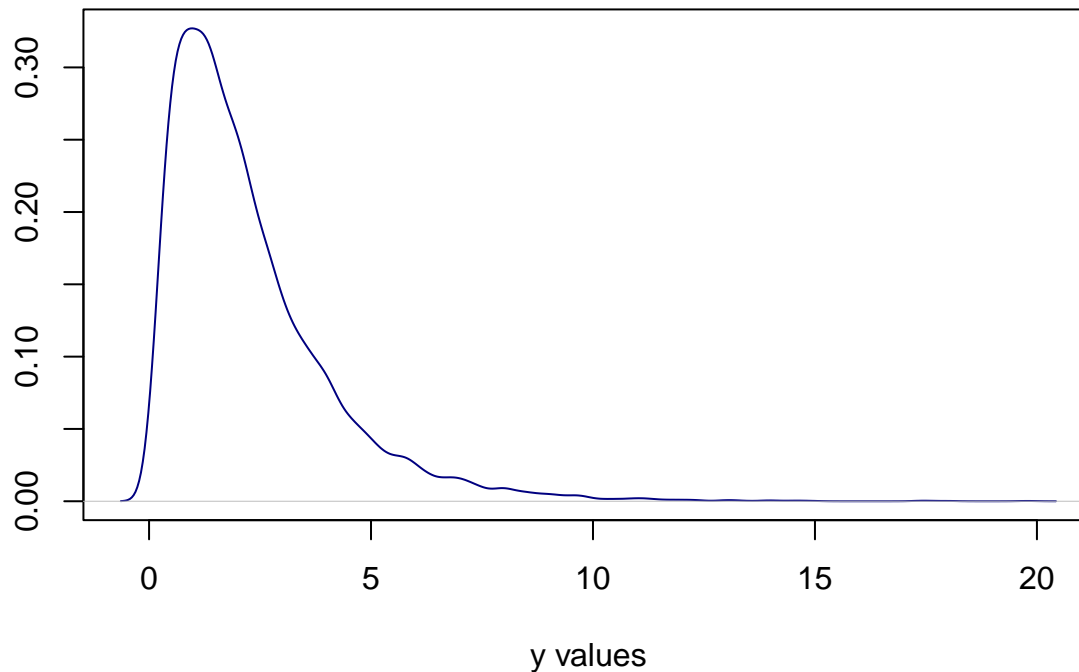
### Task b

```
obs <- c(2.32,1.82,2.4,2.08,2.13)
n <- length(obs)
Ndraws <- 10000

thetaDraws <- rgamma(n = Ndraws, shape = 2*n + 1, rate = 0.5 + sum(obs))
y <- rgamma(n = Ndraws, shape = 2, rate = thetaDraws)

plot(density(y), type = "l", col = "navy",
     main = "Posterior Distribution",
     xlab = "y values", ylab = "")
```

## Posterior Distribution



```
prob <- mean(y<1.9)
```

The  $Pr(Y_6 < 1.9|y_1, \dots, y_5)$  is 0.5313.

### Task c

```
Nweeks <- 30
weights <- matrix(0, Ndraws, Nweeks)

for (i in 1:Ndraws) {
  thetaDraws <- rgamma(n = Nweeks, shape = 2*n + 1, rate = 0.5 + sum(obs))
  weights[i,]<- rgamma(n = Nweeks, shape = 2, rate = thetaDraws)
}

ExceedWeights <- mean(rowSums(weights>2.4))
```

The expected number of weeks out of the future 30 weeks in which the maximal weight will exceed 2.4 thousands of kilos is approximately 10.5.

### Task d

```
loss_function <- function(a,weights){
  res <- a + mean(rowSums(weights>0.9*log(a)))
  return(res)
}

aGrid <- seq(0.01,10,0.01)

expected_loss <- matrix(0,length(aGrid),1)
```

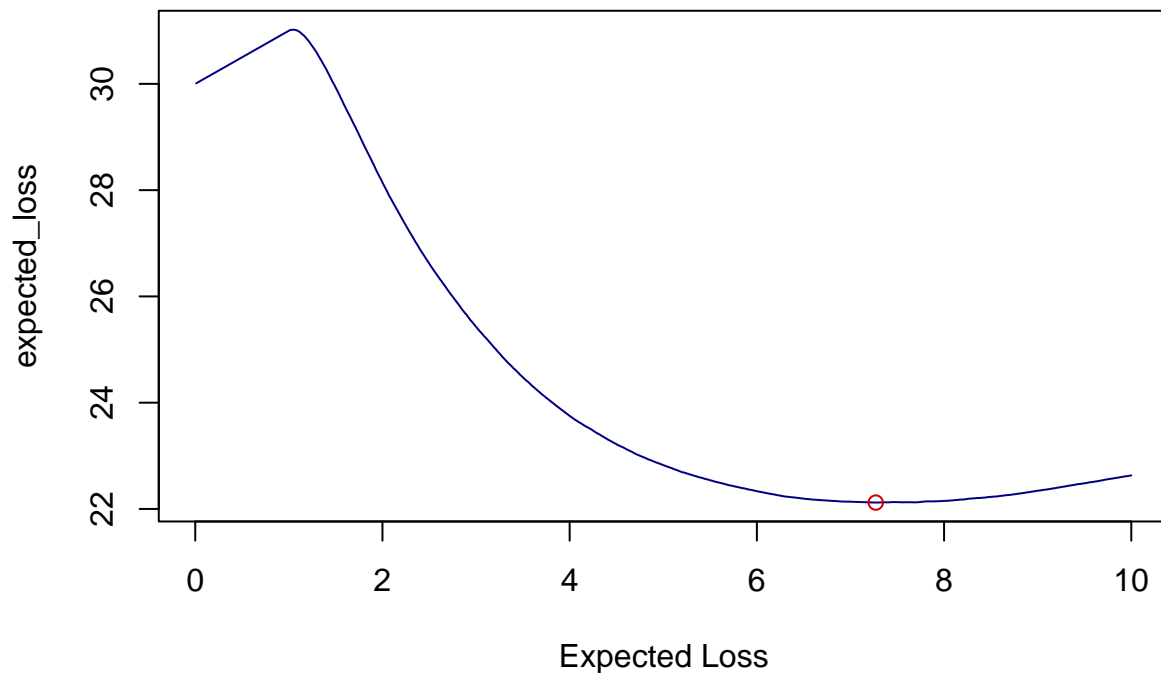
```

for (i in 1:length(aGrid)){
  expected_loss[i,] <- loss_function(aGrid[i],weights)
}

a0pt <- aGrid[which.min(expected_loss)]

plot(aGrid,expected_loss, type = "l", col = "navy",
      xlab = "Expected Loss")
points(a0pt, loss_function(a0pt,weights),col = "red3")

```



The optimal build cost is 7.39.

## Problem 2

```
source("ExamData.R")
```

### Task a

```

library(mvtnorm)

nIter <- 10000
mu_0 <- as.vector(rep(0,8))
Omega_0 <- 1/9 * diag(8)
v_0 <- 1
sigma2_0 <- 9
X <- as.matrix(X)

PostDraws <- BayesLinReg(y, X, mu_0, Omega_0, v_0, sigma2_0, nIter)

```

```

BetaDraws <- PostDraws$betaSample

interval <- quantile(BetaDraws[,2], probs = c(0.005,0.995))

interval <- data.frame(lower_bound = interval[1], upper_bound = interval[2])
colnames(interval) <- c("lower bound", "upper bound")
rownames(interval) <- c("99% Equal Tail Credible Interval")
knitr::kable(interval)

```

	lower bound	upper bound
99% Equal Tail Credible Interval	-0.350173	1.8344

It is 99 % posterior probability that  $\beta_1$  is on the interval (-0.33,1.84).

## Task b

```

mu <- BetaDraws[,1] + BetaDraws[,2] + BetaDraws[,3] + BetaDraws[,4] * 0.5 +
  BetaDraws[,6] + BetaDraws[,8]

CV <- sqrt(PostDraws$sigma2Sample)/mu

MedianCV <- median(CV)

```

The median of the coefficient of variation is approximately 1.83.

## Task c

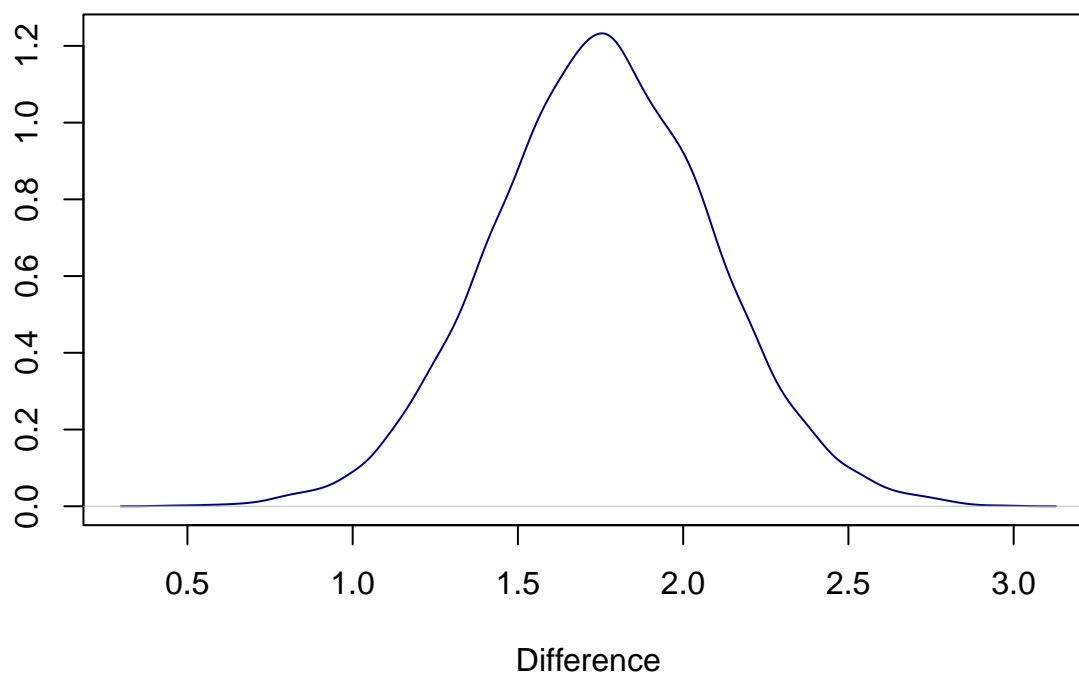
```

Effect_Inner <- BetaDraws[,1] + BetaDraws[,2] +
  BetaDraws[,5] + BetaDraws[,7]
Effect_South <- BetaDraws[,1] + BetaDraws[,2] +
  BetaDraws[,6] + BetaDraws[,8]

Diff <- Effect_Inner - Effect_South
plot(density(Diff), type = "l", col = "navy",
     main = "Inner vs South Appartments",
     xlab = "Difference", ylab = "")

```

## Inner vs South Apartments

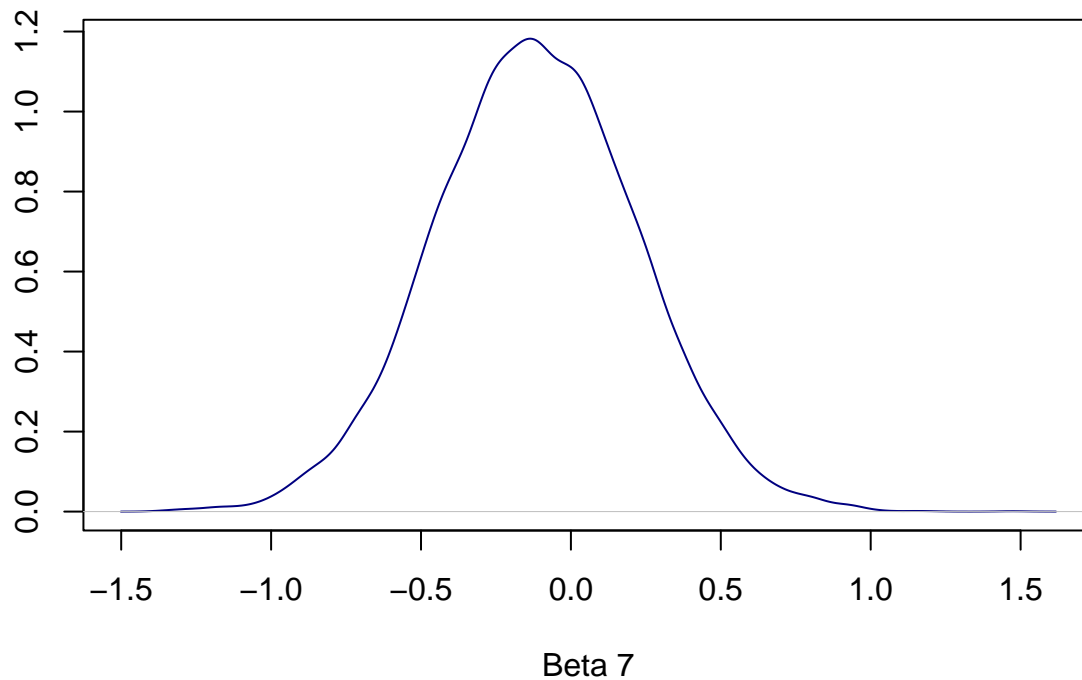


```
intervalDiff <- quantile(Diff, probs = c(0.025,0.975))
intervalDiff <- data.frame(lower_bound = intervalDiff[1],
                           upper_bound = intervalDiff[2])
colnames(intervalDiff) <- c("lower bound", "upper bound")
rownames(intervalDiff) <- c("95% Equal Tail Credible Interval")
knitr::kable(intervalDiff)
```

	lower bound	upper bound
95% Equal Tail Credible Interval	1.101313	2.413338

There is a high probability that the apartments in the inner city have a higher price than the apartments on the south side of the city. The 95% equal tail credible interval has positive bounds which strengthens the assumption.

```
plot(density(BetaDraws[,8]), type = "l", col = "navy",
     main = "", xlab = "Beta 7", ylab = "")
```



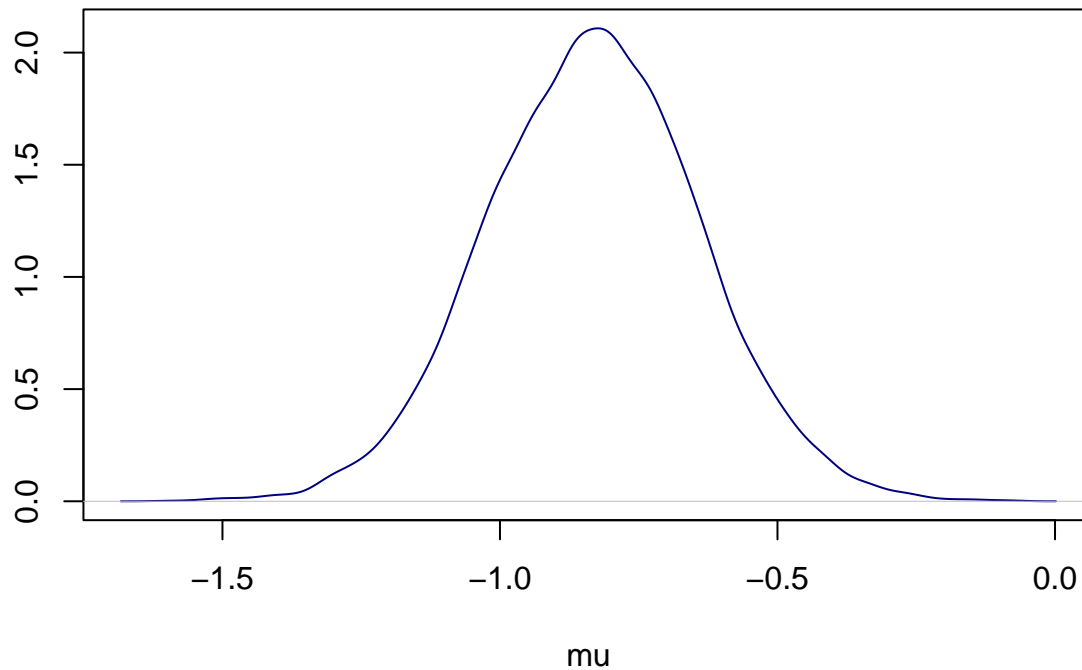
There is a substantial probability mass on both sides of 0. Thus, the effect on the selling price  $y$  from  $x_1$  is not different for apartments on the south side of the city compared to apartments which are neither in the inner city nor on the south side of the city.

### Task d

```
mu <- BetaDraws[,1] + BetaDraws[,2] * (-0.5) + BetaDraws[,3] * (-0.5) +
  BetaDraws[,6] + BetaDraws[,8] * (-0.5)

plot(density(mu), type = "l", col = "navy",
     main = "Posterior Distribution of mu",
     xlab = "mu", ylab = "")
```

## Posterior Distribution of mu



```
prob <- mean(mu>0)
```

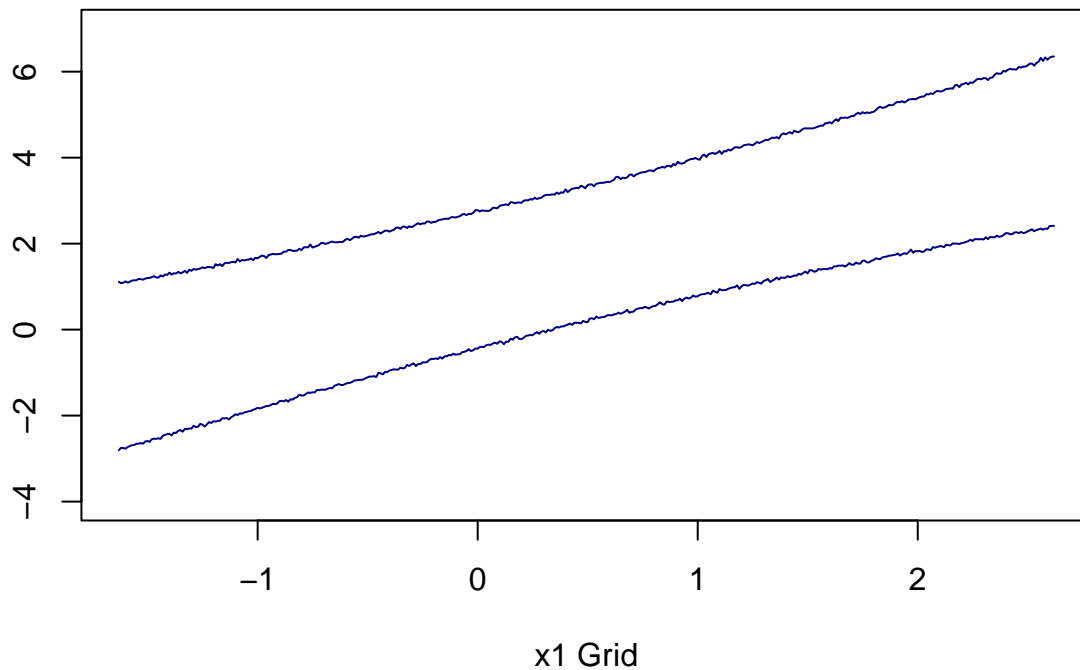
The posterior probability that  $\mu > 0$  is 0.

### Task e

```
x1Grid <- seq(min(X[,2]), max(X[,2]), 0.01)
intervals <- matrix(0,length(x1Grid),2)

for (i in 1:length(x1Grid)) {
  mu <- BetaDraws[,1] + BetaDraws[,2] * x1Grid[i] + BetaDraws[,3] +
    BetaDraws[,4] * 0.5 + BetaDraws[,5] + BetaDraws[,7] * x1Grid[i]
  intervals[i,] <- quantile(rnorm(nIter, mu, sqrt(PostDraws$sigma2Sample)),
    probs = c(0.025,0.975))
}

plot(x1Grid,intervals[,1], type = "l", col = "navy",
     main = "", xlab = "x1 Grid", ylab = "", ylim=c(-4,7))
lines(x1Grid, intervals[,2], type = "l", col = "navy")
```



## Problem 3

### Task a,b,c

Hand written solution.

### Task d

```
LogPost <- function(theta, n, sumlogx){
  res <- 2*theta*sumlogx - n*(theta^2)
  return(res)
}

thetaGrid <- seq(-1,2,0.01)
n <- 5
sumlogx <- 2

PostDens_propto <- exp(LogPost(thetaGrid,n,sumlogx))
PostDens <- PostDens_propto/0.01*sum(PostDens_propto)

plot(thetaGrid,PostDens, type = "l", col = "navy",
      main = "Posterior Distribution of theta",
      xlab = "theta", ylab = "")
```



