BayesExam

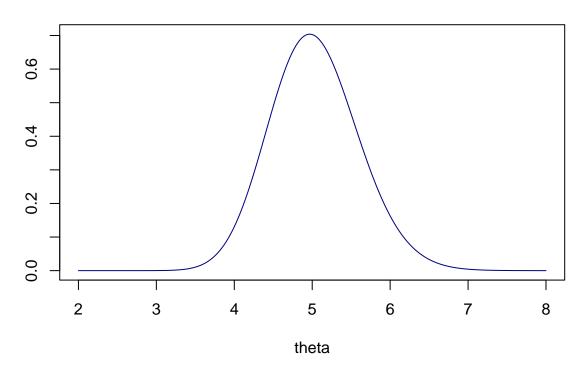
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Problem 1

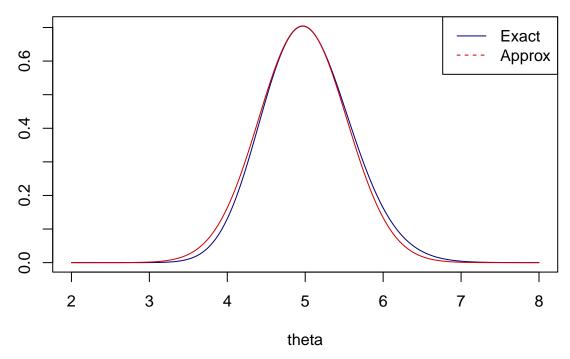
Task d

Posterior Distribution of theta



Task e

Posterior Distribution of theta



The posterior approximation very accurate. The exact posterior is very slightly skewed to the right.

Task f

```
set.seed(12345)

nSim <- 1000
T_x_rep <- matrix(0,nSim,1)

for (i in 1:nSim){
    theta <- rgamma(1, shape = 3 + sumx, rate = 0.5 + n)
    x_rep <- rpois(n,theta)
    T_x_rep[i,] <- max(x_rep)
}

prob1f <- mean(T_x_rep>14)
```

The posterior predictive p-value is 0.002, which is a very low value; thus, it is not reasonable to think that the maximum value of 14 from Gunnar originates from the Poisson distribution in this problem.

Problem 2

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

Task a

```
mu_0 <- as.vector(rep(0,3))
Sigma2_0 <- 16*diag(3)</pre>
```

```
nIter <- 10000
X <- as.matrix(X)

PostDraws <- BayesLogitReg(y, X, mu_0, Sigma2_0, nIter)

Betas <- PostDraws$betaSample

intervalB1 <- quantile(Betas[,2], probs = c(0.05, 0.95))
# table for the interval
intervalB1 <- data.frame(lower_bound = intervalB1[1], upper_bound = intervalB1[2])
colnames(intervalB1) <- c("Lower bound", "Upper bound")
rownames(intervalB1) <- c("90% Equal Tail Credible Interval")
knitr::kable(intervalB1)</pre>
```

	Lower bound	Upper bound
90% Equal Tail Credible Interval	0.2099814	1.887007

The 90% posterior probability that β_1 is on the interval.

Task b

```
prob2b <- mean(Betas[,3]>0)
```

The posterior probability that $\beta_2 > 0$ is 0.8828. The effect from variable x2 on p is positive when it changes from 0 to 1.

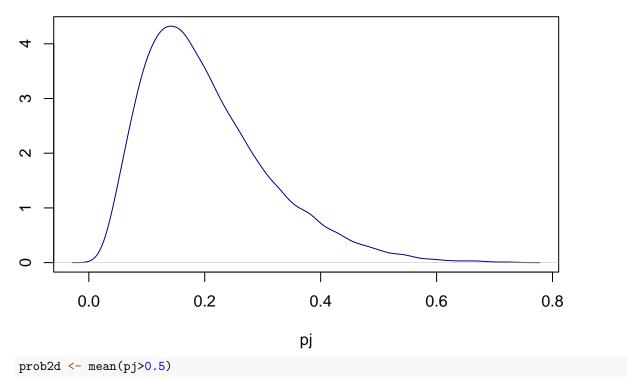
Task c

```
prob2c <- mean(Betas[,2]>0 & Betas[,3]>0)
```

The joint posterior probability that both $\beta_1 > 0$ and $\beta_2 > 0$ is 0.8711.

Task d

Posterior distribution of pj



The posterior probability that $p_j > 0.5$ for this patient is 0.0156.

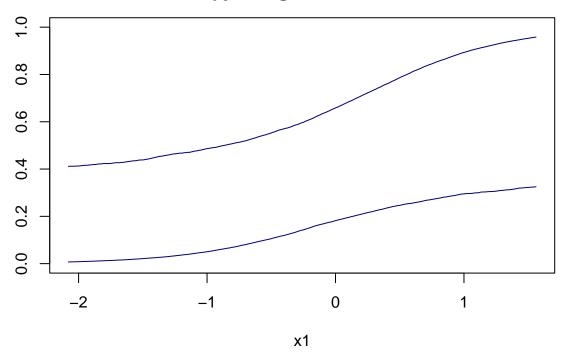
Task e

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

for (i in 1:length(x1Grid)){
   numerator <- exp(Betas[,1] + Betas[,2]*x1Grid[i] +Betas[,3])
   pj <- numerator/(1+numerator)
   intervals[i,] <- quantile(pj, probs = c(0.025,0.975))
}

plot(x1Grid, intervals[,1], type = "l", col = "navy",
        main = "95 % equal tail posterior probability intervals
        for pj on a grid of values of x1",
        xlab = "x1", ylab = "", ylim = c(0,1))
lines(x1Grid, intervals[,2], type = "l", col = "navy")</pre>
```

95 % equal tail posterior probability intervals for pj on a grid of values of x1



Problem 3

Task b

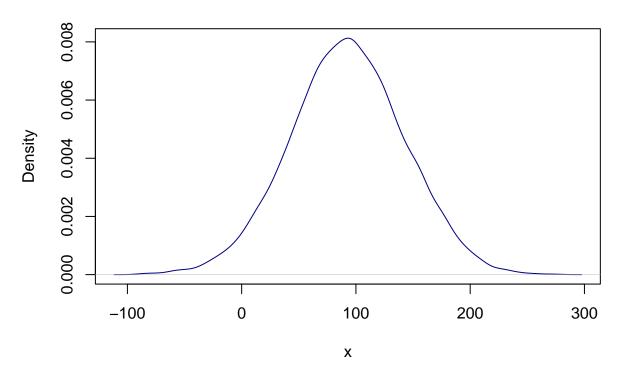
```
set.seed(12345)

nSim <- 10000
x <- matrix(0,nSim,1)

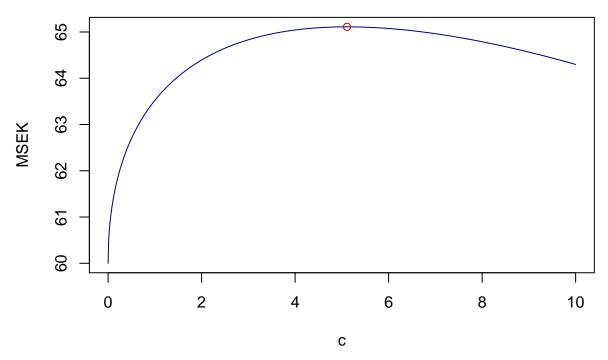
for (i in 1:nSim) {
    mu <- rnorm(1, mean = 92, sd = 2)
    x[i,] <- rnorm(1, mean = mu, sd = 50)
}

plot(density(x), col = "navy",
    main = "Posterior Predictive Distribution",
    xlab = "x")</pre>
```

Posterior Predictive Distribution



Task c



The industry spending c MSEK on advertisements is 5.11.