

# Homework 2 se

STA 360: Assignment 2, Fall 2020

Due Friday August 28, 5 PM Standard Eastern Time

```
library(tidyverse) #load in tidyverse package
```

## Lab Component

a. Task 3

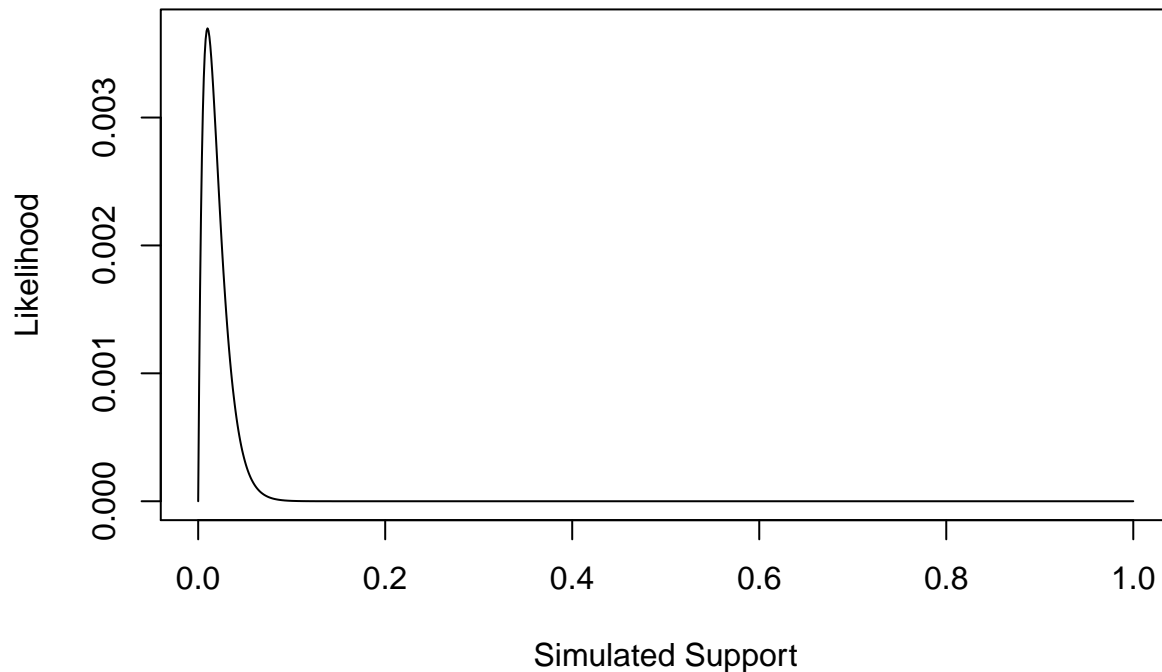
Write a function that takes as its inputs that data you simulated (or any data of the same type) and a sequence of  $\theta$  values of length 1000 and produces Likelihood values based on the Binomial Likelihood. Plot your sequence and its corresponding Likelihood function.

The likelihood function is given below. Since this is a probability and is only valid over the interval from  $[0, 1]$  we generate a sequence over that interval of length 1000.

```
set.seed(123)
### Bernoulli LH Function ###
# Input: obs.data, theta
# Output: bernoulli likelihood
obs.data <- rbinom(n = 100, size = 1, prob = 0.01)
bernLH <- function(obs.data, theta){
  N <- length(obs.data)
  x <- sum(obs.data)
  LH <- (theta ^x) * ((1-theta)^(N-x))
  return (LH)
}

### Plot LH for a grid of theta values ###
# Create the grid #
theta.sim <- seq(from = 0, to = 1, length.out = 1000)
# Store the LH values
sim.LH <- bernLH(obs.data, theta = theta.sim)
# Create the Plot
plot(theta.sim, sim.LH, type = "l", main = "Likelihood Profile",
      xlab = "Simulated Support",
      ylab = "Likelihood")
```

## Likelihood Profile



- b. Task 4 Write a function with, input: prior parameters  $a$ ,  $b$ , and the observed data. output: parameters of the Beta posterior distribution of  $\theta$  takes as its inputs prior parameters  $a$  and  $b$  for the Beta-Bernoulli model and the observed data, and produces the posterior parameters you need for the model

```
myBetaBernoulli <- function(obs.data, theta, a, b){  
  N <- length(obs.data)  
  x <- sum(obs.data)  
  BeBern <- dbeta(theta, x + a, N-x+b)  
  param1= x+a  
  param2 = N-x+b  
  print(param1)  
  print(param2)  
  return(BeBern)  
}  
  
non.informative.prior <- myBetaBernoulli(obs.data, theta.sim,1,1)
```

```
## [1] 2  
## [1] 100
```

```
informative.prior <- myBetaBernoulli(obs.data, theta.sim,3,1)
```

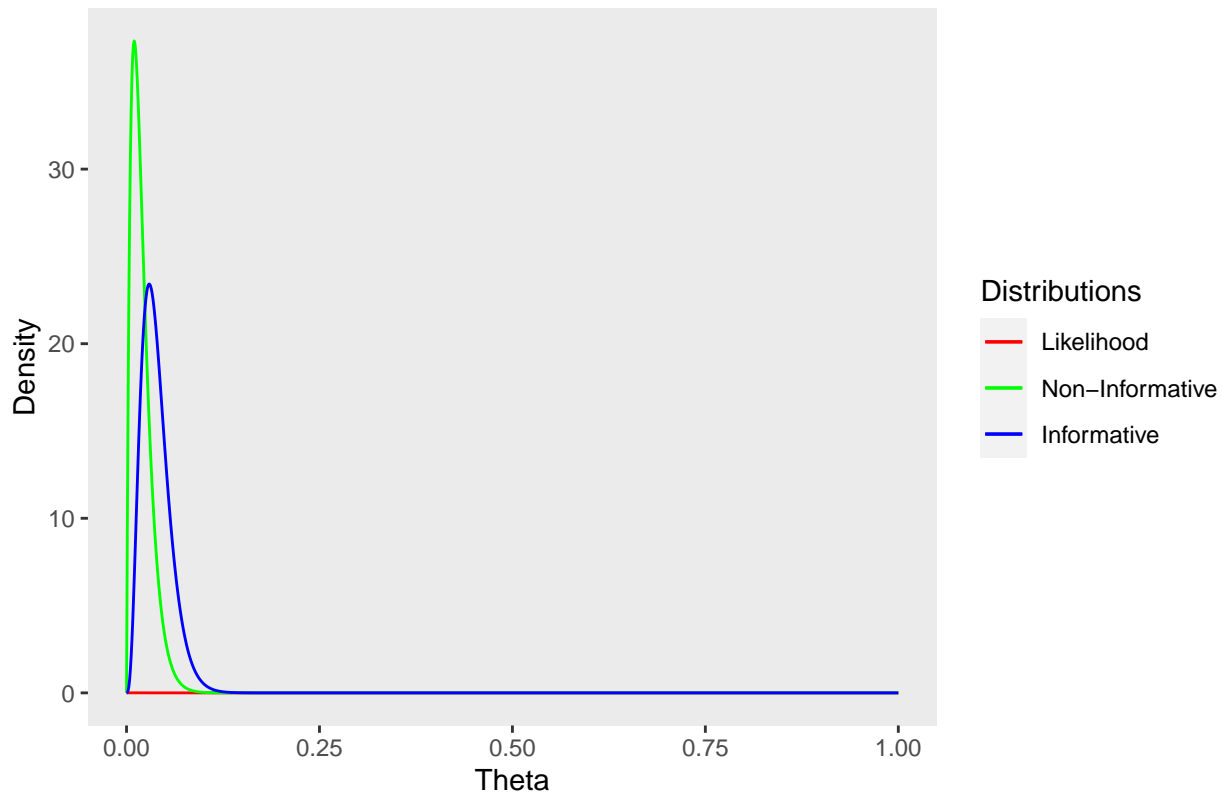
```
## [1] 4  
## [1] 100
```

The parameters for the posterior with a non-informative prior are Beta(2, 100) and for the informative, the parameters are Beta(4, 100).

- c. Task 5 Create two plots, one for the informative and one for the non-informative case to show the posterior distribution and superimpose the prior distributions on each along with the likelihood. What do you see? Remember to turn the y-axis ticks off since superimposing may make the scale non-sense.

```
df<- data.frame(theta.sim, non.informative.prior, informative.prior, sim.LH)
ggplot(df, aes(theta.sim)) +
  geom_line(aes(y=sim.LH, color="Likelihood")) +
  geom_line(aes(y=non.informative.prior,
                color= "Non-Informative")) +
  geom_line(aes(y=informative.prior, color="Informative"))+
  scale_color_manual(name = "Distributions",
                    breaks = c("Likelihood","Non-Informative",
                              "Informative"),
                    values = c("red", "green","blue"))+
  labs(title = "Likelihood, Posterior given Non-Informative/Informative Prior",
        y = "Density",
        x = "Theta" ) +
  theme(
    panel.grid.major = element_blank(),
    panel.grid.minor = element_blank(),
  )
```

Likelihood, Posterior given Non-Informative/Informative Prior



We can see that the informative prior is slightly more to the right than the non-informative prior. Having an informative prior shifted the posterior distribution so the value of theta it's centered about is greater than the non-informative prior posterior distribution. We can also see that the non-informative distribution is taller than the informative,