STAT 261, Lab 1

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Likelihood methods for the Binomial distribution

- Put your name in the author section above.
- Execute each chunk of code to ensure that your code works properly. The TA will review the code with you.
- Sometimes one of your chunks of code will not compile properly, but you must hand your document in. In that case, 'Comment' out the R code that is not working properly using # as the first character in your lines of code. Do that now for the chunk of code in Section number 4, below. Put # at the start of each line of R code.
- Complete question 6 below. Execute your question 6 code chunk to ensure that it works properly. If you cannot compile it without errors before the end of the lab, comment out the chunk.
- Save the Rmd file
- Knit the Rmd file to pdf (either directly or by knitting to Word/HTML and printing to pdf)
- You are welcome to submit this pdf into the "Lab 0 (practice submission)" activity if you have not already completed the practice submission, or if you want your instructor to look at your work.

1. First we generate 1 observation from the Binomial (100, theta=.07) distribution

[1] 6

2. We know that the MLE for the probability of success, theta, is the sample proportion.

```
thetahat <- bdat/100
thetahat
```

[1] 0.06

The MLE of theta is 0.06.

3. Compute the vector of Log-Likelihoods for a sequence of theta values given the one data point bdat.

```
#create a vector of values from 0.02 to 0.14 in steps of 0.005
theta <- seq(0.02, 0.14, by=.005)

#compute the log of binomial probabilities
bloglike <- dbinom(bdat, size=100, prob=theta, log=TRUE)

#combine the theta values and the log probabilities as column vectors
head(cbind(theta, bloglike))

## theta bloglike
## [1,] 0.020 -4.472250
## [2,] 0.025 -3.614208
## [3,] 0.030 -3.003571
## [4,] 0.035 -2.564456
## [5,] 0.040 -2.251580
## [6,] 0.045 -2.035745
```

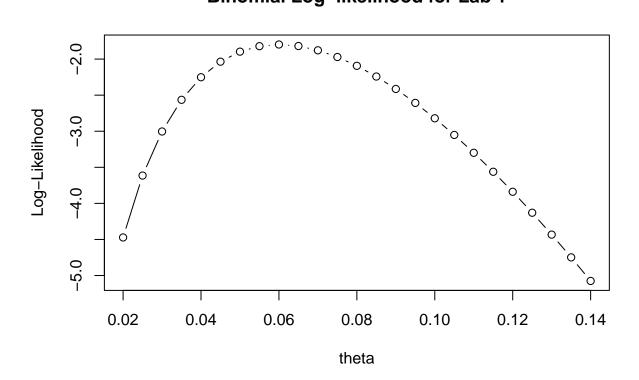
4. Plot the Log-Likelihood function versus theta.

```
There are several ways to call the plot function in R. Here are two ways to plot Y versus X. 1. plot(x, y) 2. plot(y \sim x)
```

Where y is plotted on the vertical axis and x is plotted on horizontal axis.

```
plot(bloglike ~ theta, ylab='Log-Likelihood', xlab='theta', type='b')
title('Binomial Log-likelihood for Lab 1')
```

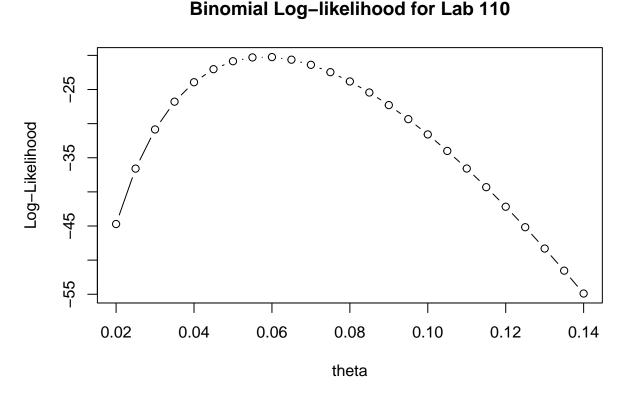
Binomial Log-likelihood for Lab 1



5. Generate n=10 observations from the same distribution and plot the loglikelihood function given the n observations.

```
set.seed(54321)
n <- 10
bdat2 <- rbinom(n, size=100, prob=.07)</pre>
                                          # this generates n binomial observations
bdat2
## [1] 6 7 5 5 5 10 3 5 6 6
theta \leftarrow seq(0.02, 0.14, by=.005)
bloglike2 <- 0
for (i in 1:n){
  bloglike2 <- bloglike2 + dbinom(bdat2[i], size=100, prob=theta, log=TRUE)
  #computes the sum of the log of binomial probabilities over the observations
}
head(cbind(theta, bloglike2))
        theta bloglike2
## [1,] 0.020 -44.71201
## [2,] 0.025 -36.58811
## [3,] 0.030 -30.85666
## [4,] 0.035 -26.78414
## [5,] 0.040 -23.93284
## [6,] 0.045 -22.02050
plot(bloglike2 ~ theta, ylab='Log-Likelihood', xlab='theta', type='b')
title(paste('Binomial Log-likelihood for Lab 1', n='', n, sep=''))
```

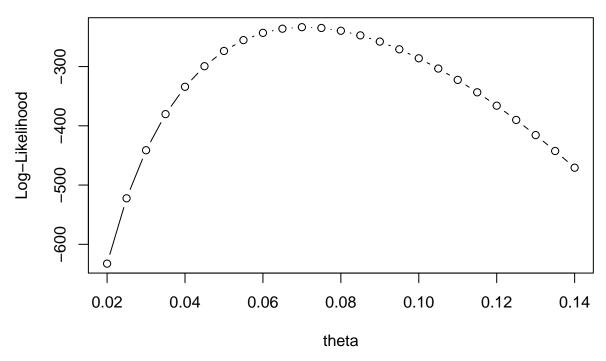
Binomial Log-likelihood for Lab 110



6. Now it is your turn. In the R code chunk below, redo question 5 for n=100 observations from the same distribution. How does the log-liklihood function compare with that in question 5? Write your comments in the section below the R chunk.

```
set.seed(54321)
n <- 100
bdat3 <- rbinom(n, size=100, prob=0.07)
##
    [1]
               5 5 5 10
                         3 5
                               6 6
                                     4 8
                                           3 8 14 8 11
                                                         7
                                                            7
                                                                  5
                                                                     3 10 5 11
        7 8 9 2 12 8 4 12 4 10
                                     6 11 9 10 7 10
                                                      5
##
   [26]
                                                         8
                                                            6 11
                                                                  7
                                                                     8 10 4 11
                                                      4 6 5 8 7
  [51] 7 9 7 4 7 7 5 6 4 6 6 10 6 7 2 7
                                                                    8 7 10 8
## [76] 3 9 4 4 4 5 11 9 7 5 12 8 8 9 2 7 8 5 10 9 7 8 7
theta \leftarrow seq(0.02, 0.14, by=.005)
bloglike3 <- 0
for (i in 1:n){
  bloglike3 <- bloglike3 + dbinom(bdat3[i], size=100, prob=theta, log=TRUE)
head(cbind(theta, bloglike3))
       theta bloglike3
## [1,] 0.020 -632.5735
## [2,] 0.025 -522.3457
## [3,] 0.030 -441.2234
## [4,] 0.035 -380.2648
## [5,] 0.040 -334.1335
## [6,] 0.045 -299.3884
plot(bloglike3 ~ theta, ylab='Log-Likelihood', xlab='theta', type='b')
title(paste('Binomial Log-likelihood for Lab 1', n='', n, sep=''))
```

Binomial Log-likelihood for Lab 1100



Comments: