Lab 2 – Beta-Binomial Distribution

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In class, you saw the Binomial-Beta model. We will now use this to solve a very real problem! Suppose I wish to determine whether the probability that a worker will fake an illness is truly 1%. Your task is to assist me! Tasks 1–3 will be completed in lab and tasks 3–5 should be completed in your weekly homework assignment. You should still upload task 3 even though this will be worked through in lab!

Task 1

Let's start by quickly deriving the Beta-Binomial distribution.

We assume that

$$X \mid \theta \sim \text{Binomial}(\theta)$$

,

$$\theta \sim \text{Beta}(a, b)$$
,

where a, b are assumed to be known parameters. What is the posterior distribution of $\theta \mid X$?

$$p(\theta \mid X) \propto p(X \mid \theta)p(\theta) \tag{1}$$

$$\propto \theta^x (1 - \theta)^{(n-x)} \times \theta^{(a-1)} (1 - \theta)^{(b-1)} \tag{2}$$

$$\propto \theta^{x+a-1} (1-\theta)^{(n-x+b-1)}. \tag{3}$$

This implies that

$$\theta \mid X \sim \text{Beta}(x+a, n-x+b).$$

Task 2

Simulate some data using the rbinom function of size n = 100 and probability equal to 1%. Remember to set.seed(123) so that you can replicate your results.

The data can be simulated as follows:

```
# set a seed
set.seed(123)
# create the observed data
obs.data <- rbinom(n = 100, size = 1, prob = 0.01)
# inspect the observed data
head(obs.data)
## [1] 0 0 0 0 0 0</pre>
```

```
## [1] 0 0 0 0 0 0 0 tail(obs.data)
```

```
## [1] 0 0 0 0 0 0
```

```
length(obs.data)
```

[1] 100

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 θ values of length 1000 and produces Likelihood values based on the Binomial Likelihood. Plot your sequence and its corresponding Likelihood function.

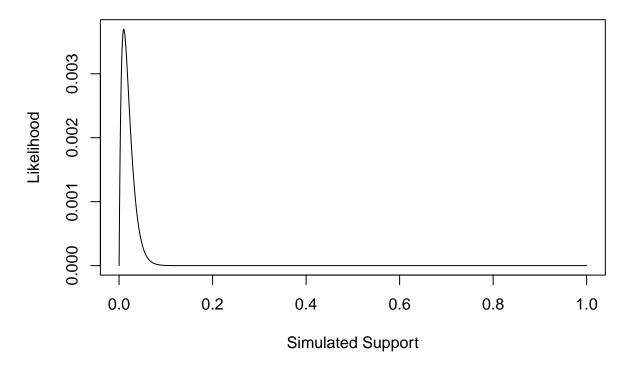
The likelihood function is given below. Since this is a probability and is only valid over the interval from [0,1] we generate a sequence over that interval of length 1000.

You have a rough sketch of what you should do for this part of the assignment. Try this out in lab on your own.

```
### Bernoulli LH Function ###
# Input: obs.data, theta
# Output: bernoulli likelihood
myBernLH <- 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 <- myBernLH(obs.data, theta.sim)
# Create the Plot
plot(theta.sim, sim.LH, type = "l", main = "Likelihood Profile", xlab = "Simulated Support", ylab = "Li</pre>
```

Likelihood Profile



Task 4 (To be completed for homework)

Write a function that 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. **Generate and print** the posterior parameters for a non-informative prior i.e. (a,b) = (1,1) and for an informative case (a,b) = (3,1).

```
postParam <- function(a, b, obs.data) {
    n <- length(obs.data)
    x <- sum(obs.data)
    post_a <- x + a
    post_b <- n - x + b
    return(c(post_a, post_b))
}

postParam(1, 1, obs.data)

## [1] 2 100

postParam(3, 1, obs.data)

## [1] 4 100</pre>
```

Task 5 (To be completed for homework)

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.