

## Exercise 3.1.5

### Contents

<b>1</b>	<b>Question</b>	<b>1</b>
<b>2</b>	<b>Comments/Solution</b>	<b>1</b>
<b>3</b>	<b>Code</b>	<b>1</b>
3.1	libraries . . . . .	1
3.2	Data . . . . .	2
3.3	Stan code . . . . .	2
3.4	code in R to run stan . . . . .	2
<b>4</b>	<b>Outputs</b>	<b>3</b>
4.1	Model summary . . . . .	3
4.2	Plots . . . . .	3

### 1 Question

Exercise 3.1.5 Alter the data to  $k = 99$  and  $n = 100$ , and comment on the shape of the posterior for the rate  $\theta$ .

### 2 Comments/Solution

Comparing at the posterior distributions (from the plots section below) of  $k=99$ ,  $n=100$  to the previous distribution of  $k=50$ ,  $n=100$ , one can observe that the  $\theta$  estimates have shifted towards the extreme right of the graph to a value very close to 1 with slight uncertainty. Since the data shows that 99 times out of 100 we get a positive, the posterior has very little doubt as to where the real value lies.

The model used to calculate the required values and the plots is scripted below. Copy/pasting the given code will generate the same result on your own machine.

### 3 Code

#### 3.1 libraries

The libraries required for the script and the plots.

```
# clears workspace
rm(list=ls())
#load libraries
library(rstan)
library(ggplot2)
library(patchwork)
```

## 3.2 Data

The data required for this particular stan model.

```
# data initialization
k <- 50
n <- 100
k_1 <- 99
n_1 <- 100
# to be passed on to Stan
stan_data <- list(k = k, n = n)
stan_data_1 <- list(k = k_1, n = n_1)
```

## 3.3 Stan code

Stan code, that can be written in R as such or in a separate new file with stan extension.

```
write("// Stan code here in this section

// Inferring theta
data {
  int<lower=1> n;
  int<lower=0> k;
}
parameters {
  real<lower=0,upper=1> theta;
}
model {
  // Prior Distribution for theta
  theta ~ beta(1, 1);

  // Observed Counts
  k ~ binomial(n, theta);
} // ",

"3_1_5.stan")
```

## 3.4 code in R to run stan

Running stan through R (with the required input parameters).

```
myinits <- list(
  list(theta=.1), # chain 1 starting value
  list(theta=.9)) # chain 2 starting value

# parameters to be monitored:
parameters <- c("theta")

# The following command calls Stan with specific options.
# For a detailed description type "?stan".
mod_fit <- stan(file="3_1_5.stan",
  data=stan_data,
  init=myinits, # If not specified, gives random inits
  pars=parameters,
  iter=2000,
  chains=2,
```

```

        thin=1,
        warmup=100, # Stands for burn-in; Default = iter/2
        seed=123 # Setting seed; Default is random seed
    )
mod_fit_1 <- stan(file="3_1_5.stan",
    data=stan_data_1,
    init=myinits, # If not specified, gives random inits
    pars=parameters,
    iter=2000,
    chains=2,
    thin=1,
    warmup=100, # Stands for burn-in; Default = iter/2
    seed=123 # Setting seed; Default is random seed
)

```

## 4 Outputs

### 4.1 Model summary

For dataset k=50 and n=100. (stan\_data)

```

## Inference for Stan model: 3_1_5.
## 2 chains, each with iter=2000; warmup=100; thin=1;
## post-warmup draws per chain=1900, total post-warmup draws=3800.
##
##          mean se_mean   sd  2.5%   25%   50%   75%  97.5% n_eff Rhat
## theta    0.50     0.00 0.05  0.40  0.46  0.50  0.53  0.59 1413   1
## lp__   -71.22     0.02 0.74 -73.22 -71.38 -70.94 -70.75 -70.70 1762   1
##
## Samples were drawn using NUTS(diag_e) at Mon Oct 19 17:00:27 2020.
## For each parameter, n_eff is a crude measure of effective sample size,
## and Rhat is the potential scale reduction factor on split chains (at
## convergence, Rhat=1).

```

For dataset k=99 and n=100. (stan\_data\_1)

```

## Inference for Stan model: 3_1_5.
## 2 chains, each with iter=2000; warmup=100; thin=1;
## post-warmup draws per chain=1900, total post-warmup draws=3800.
##
##          mean se_mean   sd  2.5%   25%   50%   75%  97.5% n_eff Rhat
## theta    0.98     0.00 0.01  0.95  0.97  0.98  0.99  1.00 1468   1
## lp__   -10.38     0.02 0.76 -12.52 -10.57 -10.10 -9.90 -9.84 1121   1
##
## Samples were drawn using NUTS(diag_e) at Mon Oct 19 17:00:27 2020.
## For each parameter, n_eff is a crude measure of effective sample size,
## and Rhat is the potential scale reduction factor on split chains (at
## convergence, Rhat=1).

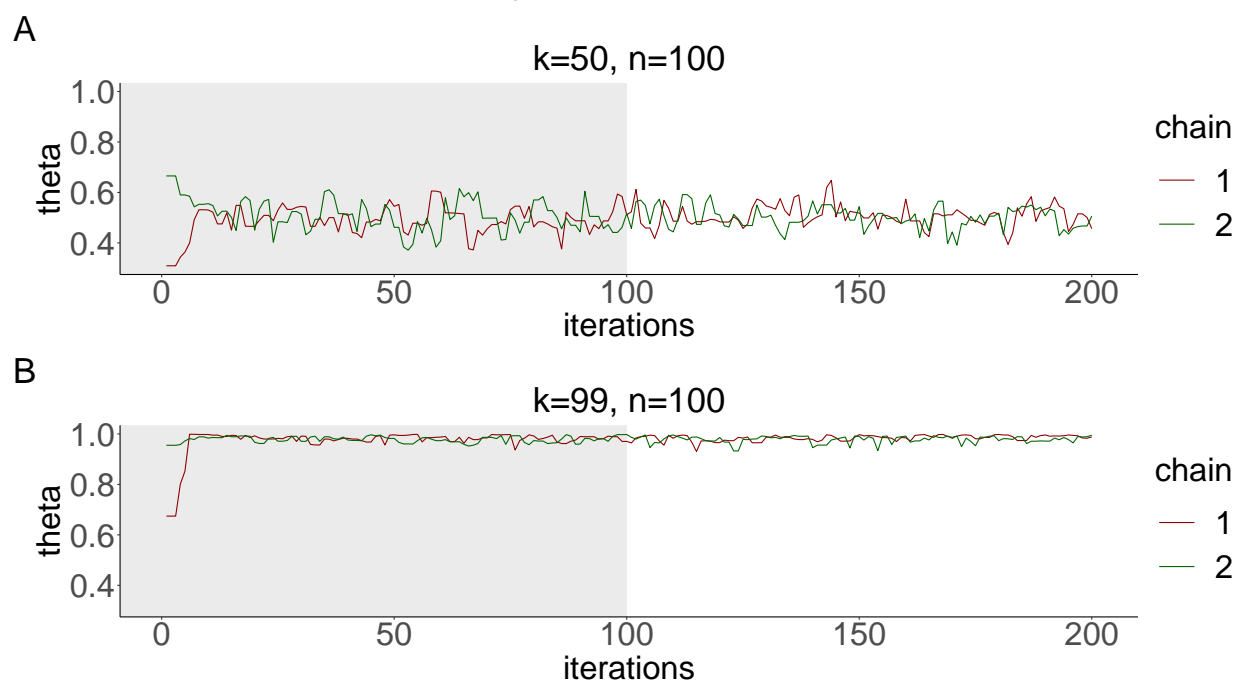
```

### 4.2 Plots

#### 4.2.1 Plot (chains)

The initial movement of the chains are shown here (including the warmup phase). The two chains begin from the initial starting points of as defined in the input parameters of the stan model.

## comparison of datasets



### 4.2.2 Plot (posterior)

The plot of the  $\theta$  values per chain superimposed on each other.

