```
library(dplyr)
```

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
## filter, lag

## The following objects are masked from 'package:base':
##
## intersect, setdiff, setequal, union

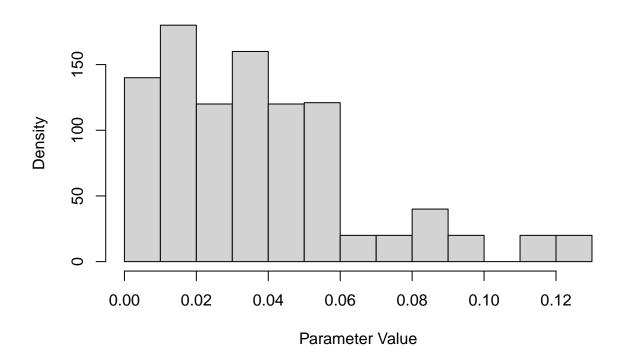
set.seed(911)
```

```
#Part 1
  #1
  network <- read.csv("network.csv")</pre>
  attributes <- read.csv("attributes.csv")</pre>
  SI_contagion_model <- function(network, initially_infected, p, time_steps) {
    for (h in 1:time_steps) {
      if (h == 1) {
        time_points <- c()</pre>
        for (i in initially_infected) {
           time_points[i] <- h
      }
      for (i in initially_infected) {
        for (j in 1:ncol(network)) {
          if (j == 1) {
             in_network <- c()</pre>
          }
          if (network[i,j] == 1) {
             in_network <- c(in_network, j)</pre>
           }
        }
        if (is.null(in_network) == FALSE) {
          random_in_network <- sample(in_network, 1)</pre>
           if (sum(random_in_network == initially_infected) == 0 & runif(1) <= p) {</pre>
             initially_infected <- c(initially_infected, random_in_network)</pre>
             time_points[random_in_network] <- h</pre>
          }
        }
      }
    }
    return(time_points)
  }
  #3
```

```
initially_infected <- subset(attributes, adoption_date==1) %>% row.names() %>% as.numeric()
 SI_contagion_model(network, initially_infected, .1, 18)
    [26] 12  1 NA NA NA NA NA NA NA NA NA 17 NA NA NA NA 9 NA 17 NA 13 NA NA NA NA
## [76] NA NA 1 NA NA NA NA NA NA NA NA NA 18 NA NA 1 NA NA NA NA NA NA NA
## [101] NA NA NA NA NA NA NA NA 4
SI_contagion_model(network, initially_infected, .5, 18)
    [1] 1 18 5 12 6 10 15 NA NA NA 2 10 9 16 NA 4 NA 10 NA 7 13 2 7 11
## [26] NA 1 15 6 NA 8 NA 18 NA 8 6 6 NA 15 11 5 6 NA 9 14 5 NA NA 5 NA
   [51] 4 9 NA 9 3 6 17 NA 16 NA NA NA NA 12 13 3 NA NA 14 NA NA NA NA NA 1
## [76] NA 8 1 NA NA 12 13 NA 17 NA 14 4 NA 13 2 NA 2 1 7 5 18 NA 6 7 7
## [101] 7 12 NA NA NA 13 NA NA 3 8 NA NA 9 9 NA NA 11
SI_contagion_model(network, initially_infected, .9, 18)
    [1] 1 9 6 2 11 10 3 NA 3 NA 3 11 7 7 NA NA 15 NA 13 6 6 8 4 NA 9
   [26] 9 1 5 7 NA 6 11 5 8 7 5 5 5 16 1 6 6 4 3 7 2 10 NA 5 NA
##
## [51] 13 5 NA 6 4 9 8 NA NA NA NA NA NA 11 8 NA NA 5 NA NA NA NA 1
## [76] NA 1 1 NA NA 3 9 NA 8 NA 7 4 NA 7 3 NA 3 1 6 13 9 NA 4 12 2
## [101] 4 7 10 NA NA 4 NA NA 11 1 12 NA 5 9 NA 11 15
#Part 2
 #4
   #98 and 99 are codes for missing data, not adoption time
   for (i in 1:nrow(attributes)) {
     if (attributes[i, "adoption_date"] == 98 | attributes[i, "adoption_date"] == 99) {
       attributes[i, "adoption_date"] <- NA
     }
   }
   ABC <- function(N, tolerance_level) {
     theta_value <- c()</pre>
     summary_statistic <- c()</pre>
     for (i in 1:N) {
      p <- runif(1)
       generated_data <- SI_contagion_model(network, initially_infected, p, 18)</pre>
       summary statistic[i] <- sum(abs(generated data - attributes$adoption date), na.rm=TRUE)
       theta_value[i] <- p</pre>
     }
     proposals <- data.frame(theta_value, summary_statistic)</pre>
     proposals <- proposals %>% arrange(summary_statistic)
     posterior_samples <- proposals$theta_value[1:N*tolerance_level]</pre>
     return(posterior samples)
   }
```

```
#5
posterior_samples <- suppressWarnings(ABC(N=1000, tolerance_level=.05))
hist(posterior_samples, ylab="Density", xlab="Parameter Value", main="Histogram of Posterior Sample</pre>
```

## **Histogram of Posterior Samples**



```
mean(posterior_samples)

## [1] 0.03687349

quantile(posterior_samples, .025)

## 2.5%

## 0.003170931

quantile(posterior_samples, .975)

## 97.5%
```

## 0.1148283