### Problem 14.1

#### Problem 14.1.2

# Priors

Assuming priors of the form:  $\pi \sim beta(\alpha_{\psi}, \beta_{\pi}), S \sim beta(\alpha_{S}, \beta_{S})$  and  $C \sim beta(\alpha_{C}, \beta_{C})$ , it is possible to code up a Gibbs sampler for this problem [8] of the form

$$Y_1|a, \pi, S, C \sim \mathcal{B}\left(a, \frac{\pi S}{\pi S + (1 - \pi)(1 - C)}\right)$$
 (14.3)

$$Y_2|b,\pi,S,C \sim \mathcal{B}\left(b, \frac{\pi(1-S)}{\pi(1-S) + (1-\pi)C}\right)$$
 (14.4)

$$\pi|a, b, Y_1, Y_2 \sim beta(Y_1 + Y_2 + \alpha_{\pi}, a + b - Y_1 - Y_2 + \beta_{\pi})$$
 (14.5)

$$S|Y_1, Y_2 \sim beta(Y_1 + \alpha_S, Y_2 + \beta_S)$$
 (14.6)

$$C|a, b, Y_1, Y_2 \sim beta(b - Y_2 + \alpha_C, a - Y_1 + \beta_C)$$
 (14.7)

Figure 1: equations

#### Priors

```
alpha_pi = 1
beta_pi = 1
alpha_S = 1
beta_S = 1
alpha_C = 1
beta_C = 1
fGibbsSampling <- function(numSamples,a, b){
  Y1[1] <- as.integer(runif(1, 1, a))
  Y2[1] <- as.integer(runif(1, a, b+a))
  lpi[1] <- runif(1, 0, 1)</pre>
  S[1] \leftarrow runif(1, 0, 1)
  C[1] \leftarrow runif(1, 0, 1)
  # cat("\nBefore Loop: ")
  # cat("\nY1:",Y1, " Y2: ", Y1, " lp: ", lpi, " S: ", S, " C: ", C)
  for(t in 2:numSamples){
    Y1[t] \leftarrow rbinom(n = 1, size = a, prob = lpi[t-1] * S[t-1] /
```

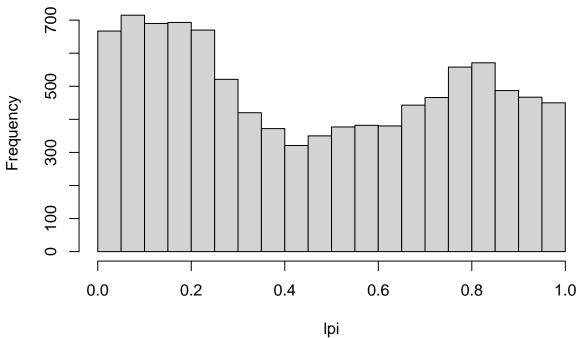
```
(lpi[t-1] * S[t-1] + (1 - lpi[t-1]) * (1 - C[t-1])))
    Y2[t] \leftarrow rbinom(n=1, size = b, prob = lpi[t-1] * (1 - S[t-1]) /
                       (lpi[t-1] * (1 - S[t-1]) + (1 - lpi[t-1]) * C[t-1]))
    lpi[t] <- rbeta(1, Y1[t] + Y2[t] + alpha_pi, a + b - Y1[t] -</pre>
                       Y2[t] + beta_pi)
    S[t] <- rbeta(1, Y1[t] + alpha_S, Y2[t] + beta_S)</pre>
    C[t] <- rbeta(1, b - Y2[t] + alpha_C, a - Y1[t] + beta_C)</pre>
    # cat("\n\nAfter t=", t)
    # cat("\nY1:",Y1, " Y2: ", Y1, " lp: ", lpi, " S: ", S, " C: ", C)
  }
  hist(lpi)
  hist(Y1)
  hist(Y2)
  hist(S)
  hist(C)
}
```

### **Problem 14.1.3**

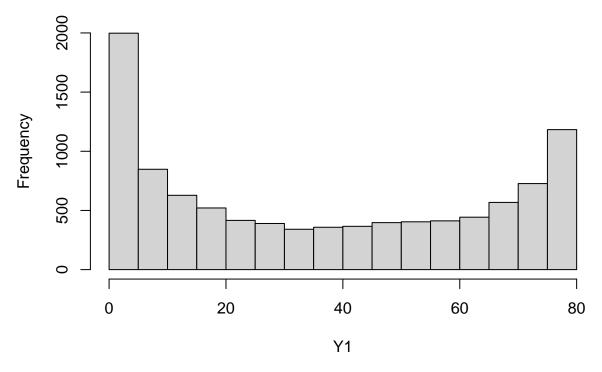
Suppose that out of a sample of 100 people, 20 of those tested negative and 80 positive. Assuming uniform priors on , S and C, use Gibbs sampling to generate posterior samples for . What do you conclude?

```
set.seed(41)
Y1 <- vector()
Y2 <- vector()
lpi <- vector()
S <- vector()
C <- vector()
s_lpi <- fGibbsSampling(10000, 80, 20)</pre>
```

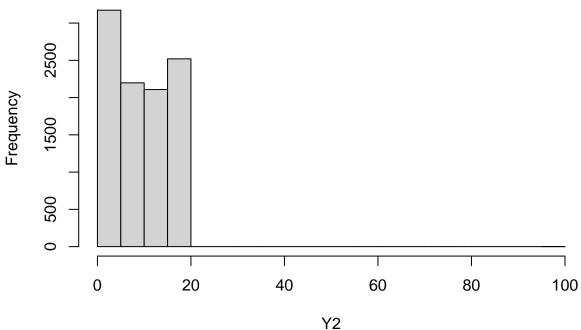
## Histogram of Ipi



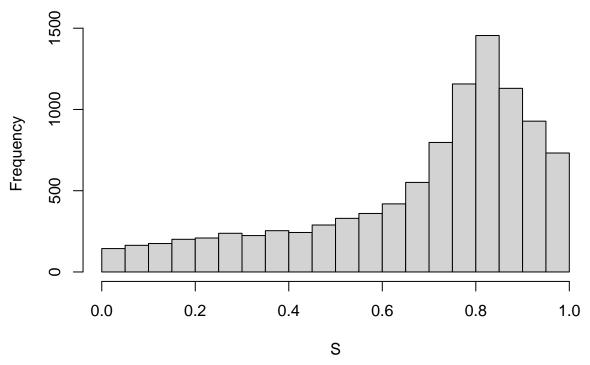




## Histogram of Y2







# Histogram of C

