

## MH AND GIBBS PRACTICAL

*#Q1) A small company improved a product and wants to infer about the proportion of potential customers who will buy the product if the new product is preferred to the old one.  
# The company is certain that this proportion will exceed 0.5, i.e. and uses the uniform prior on [0.5, 1]. Out  
# of 20 customers surveyed, 12 prefer the new product. Find the posterior for p.*

```
set.seed(2440)
theta_0 = 0
s = 10
x = c(theta_0)
m = c(theta_0)

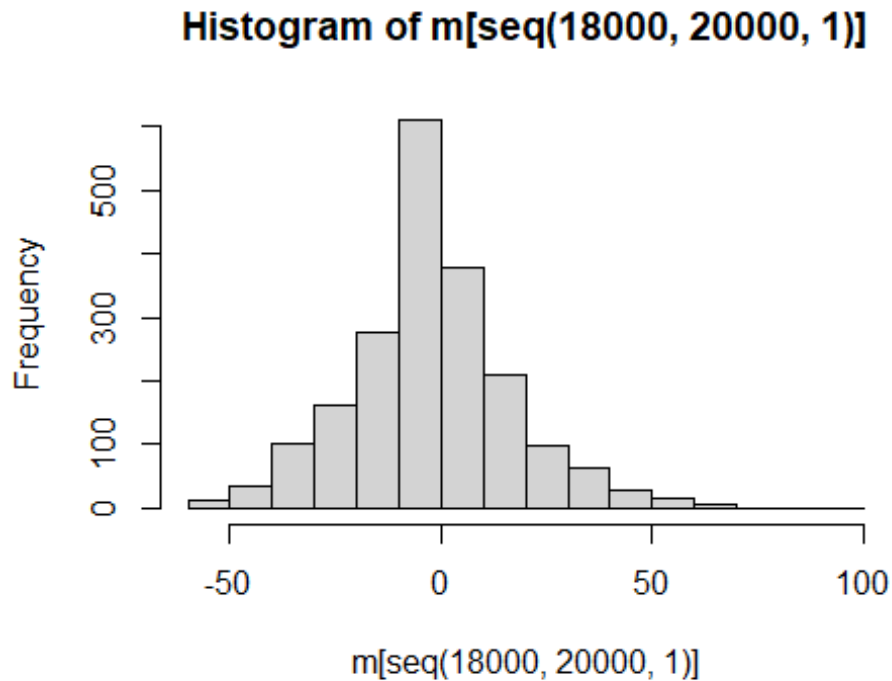
p1 = c()
p=c()

for (j in 2:20000)
{
  x[j] = rnorm(1,m[j-1],s)  #Remember: m[j-1] and not x[j-1]
  p1[j-1] = exp(x[j] - m[j-1])*((0.5+exp(x[j]))/(0.5+exp(m[j-1])))^12 * ((1+exp(m[j-1]))/(1+exp(x[j])))^22
  if (p1[j-1] > 1) {p[j-1]=1}
  else {p[j-1] = p1[j-1]}
  u1 = runif(1,0,1)
  if (u1 < p[j-1]) {m[j] = x[j]}
  else {m[j] = x[j-1]}
}
m[seq(18000,20000,1)]
hist(m[seq(18000,20000,1)])
```

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```
## Q2)
set.seed(2440)
x = matrix(rep(0,500*400), nrow = 400, ncol = 500)
x0 = runif(1,0,1)
c = 0.2
m = matrix(rep(0,500*400),nrow = 400,ncol = 500)
m[,1] = x0
for (i in 1:400)
{
  for (j in 2:500)
  {
    x[i,j] = runif(1,x0-c,x0+c)
    p1 = (x[i,j]/x0)^5
    if (p1 > 1){p=1}
    else if (x[i,j]<0){p=0}
    else if (x[i,j]>1){p=0}
    else {p = p1}
    u = runif(1,0,1)
    if (u<p)
    {m[i,j] = x[i,j]}
    else
    {m[i,j] = x[i,j-1]}
  }
}
m[,500]
```

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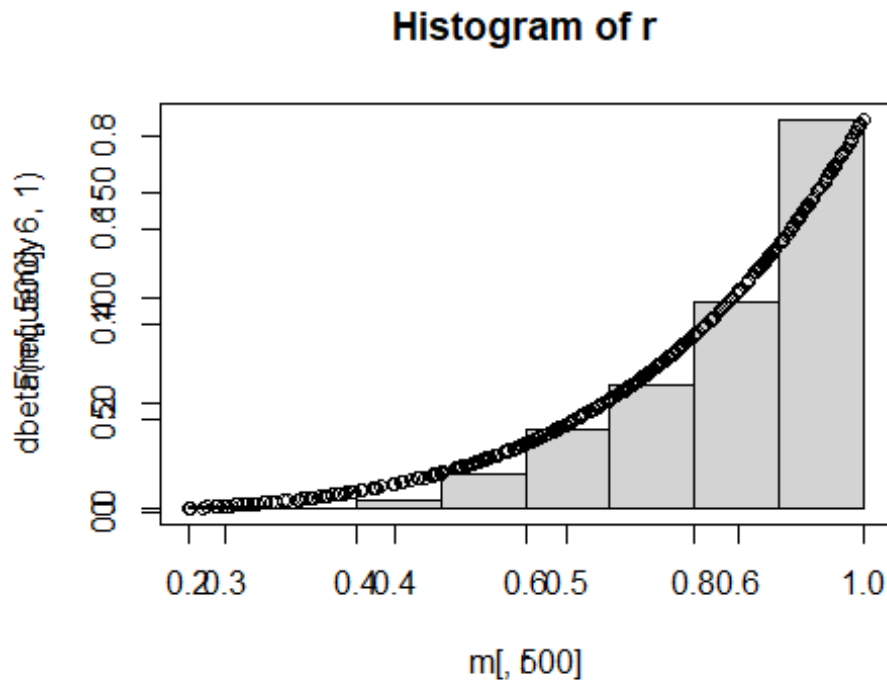
*# Visualization to check goodness of fit*

```
r = rbeta(400,6,1)
```

```
hist(r)
```

```
par(new = TRUE)
```

```
plot(m[,500],dbeta(m[,500],6,1))
```



**## GIBBS:**

**#Q1)**

```
x_0 = runif(1,0,1)
```

```
y_0 = runif(1,0,1)
```

```
x = c()
```

```
y = c()
```

```
x[1] = x_0
```

```
y[1] = y_0
```

```
for (i in 2:1000)
```

```
{
```

```
  u = runif(1,0,1)
```

```
  x[i] = (-1/y[i-1])*log(1-u+u*exp(-y[i-1]))
```

```
  y[i] = (-1/x[i])*log(1-u+u*exp(-x[i]))
```

```
}
```

```
x[1:20]
```

```
## [1] 0.754858673 0.362064850 0.743774995 0.622959544 0.556781667 0.1061421  
98
```

```
## [7] 0.575407922 0.987991759 0.160506777 0.035057668 0.681534520 0.3997657
```

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76

```
## [13] 0.651332176 0.726294141 0.219421262 0.320951016 0.978917869 0.5762040
```

76

```
## [19] 0.561246324 0.003580853
```

```
y[1:20]
```

```
## [1] 0.947210330 0.429634118 0.711490691 0.633499710 0.566271693 0.1288847  
34
```

```
## [7] 0.520049415 0.984449399 0.219403369 0.038246055 0.607084665 0.4243725  
97
```

```
## [13] 0.624640360 0.715645415 0.262712102 0.314758789 0.969899389 0.6240047  
84
```

```
## [19] 0.569000949 0.004682571
```

#Q2)

```
x_0 = 5
```

```
y_0 = 0.5
```

```
x = c()
```

```
y = c()
```

```
x[1] = x_0
```

```
y[1] = y_0
```

```
for (i in 2:1000)
```

```
{
```

```
  u = runif(1,0,1)
```

```
  x[i] = rbinom(1,16,y[i-1])
```

```
  y[i] = rbeta(1,x[i]+2,16-x[i]+4)
```

```
}
```

```
x[1:20]
```

```
## [1] 5 10 8 6 6 2 1 0 0 4 4 6 7 1 1 4 2 4 6 8
```

```
y[1:20]
```

```
## [1] 0.50000000 0.52172757 0.39042101 0.28703581 0.29789565 0.10512241
```

```
## [7] 0.06787151 0.04290157 0.22115765 0.29959999 0.28033279 0.50749852
```

```
## [13] 0.22791449 0.11293814 0.05345059 0.22547433 0.16000498 0.20706431
```

```
## [19] 0.40405134 0.29398256
```

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## MH Algorithm & Gibbs Algorithm.

Q3) Gibbs Sampler :-

Consider a conditional density

$$f(x|y=y) = ye^{-xy} \quad , \quad 0 < x < 1$$

$$f(y|x=x) = xe^{-xy} \quad , \quad 0 < y < 1$$

Proof :-

$$f(x|y=y) = ye^{-xy}$$

$$g(x|y=y) = \frac{1}{k} f(x|y=y)$$

$$k = \int_{x=0}^1 ye^{-xy} dx$$

$$= y \left[ \frac{e^{-xy}}{-y} \right]_0^1$$

$$k = 1 - e^{-y}$$

$$g(x|y=y) = \frac{ye^{-xy}}{1 - e^{-y}}$$

$$F(x|y=y) = \int_0^x g(x|y=y) dx$$

$$= \int_0^x \frac{ye^{-xy}}{1 - e^{-y}} dx$$

$$= \frac{y}{1-e^{-y}} \int_0^x e^{-xy} dx$$

$$= \frac{y}{1-e^{-y}} \left[ \frac{1-e^{-xy}}{y} \right]$$

Let,  $F[x|Y=y] = U$

$$\frac{1}{1-e^{-y}} [1-e^{-xy}] = U$$

$$1-e^{-xy} = U[1-e^{-y}]$$

$$1-e^{-xy} = U[1-e^{-y}]$$

$$e^{-xy} = 1 - U(1-e^{-y})$$

$$-xy = \ln [1 - U(1-e^{-y})]$$

$$x = -\frac{1}{y} \ln [1 - U(1-e^{-y})]$$

iii)  $y = -\frac{1}{x} \ln [1 - U(1-e^{-x})]$

Q2) For the joint dist<sup>n</sup> of  $x$  &  $y$ ,  $F(x,y) \propto \binom{n}{x} y^{x+\beta-1} (1-y)^{n-x+\beta-1}$   
 $n = 16, \alpha = 2, \beta = 4$

Ans.

$$F(x,y) \propto \binom{n}{x} y^{x+\alpha-1} (1-y)^{n-x+\beta-1}$$

$$F(x,y) = \binom{n}{x} y^{x+\alpha-1} (1-y)^{n-x+\beta-1}$$

$$f(x) = \int_0^1 f(x,y) dy$$

$$= \int_0^1 \binom{n}{x} y^{x+\alpha-1} (1-y)^{n-x+\beta-1} dy$$

$$f(x) = \binom{n}{x} \beta(x+\alpha, n-x+\beta)$$

Also,

$$F(y) = \int_0^1 f(x, y) dx$$

$$= \sum_{x=0}^n \binom{n}{x} y^{x+\alpha-1} (1-y)^{n-x+\beta-1}$$

$$= y^{\alpha-1} (1-y)^{\beta-1} \sum_{x=0}^n \binom{n}{x} y^x (1-y)^{n-x}$$

$$= y^{\alpha-1} (1-y)^{\beta-1}$$

$$F(x|Y=y) = \frac{f(x, y)}{f(y)}$$

$$= \binom{n}{x} y^x (1-y)^{n-x}$$

$$x|Y=y \sim \text{Binom}(n, y)$$

$$F(Y|x=x) = \frac{f(x, Y)}{f(x)}$$

$$= \frac{\binom{n}{x} y^{x+\alpha-1} (1-y)^{n-x+\beta-1}}{\binom{n}{x} \beta(x+\alpha, n-x+\beta)}$$

$$= \frac{1}{\beta(x+\alpha, n-x+\beta)} y^{x+\alpha-1} (1-y)^{n-x+\beta-1}$$

$$Y|x=x \rightarrow \beta_1(x+\alpha, n-x+\beta)$$