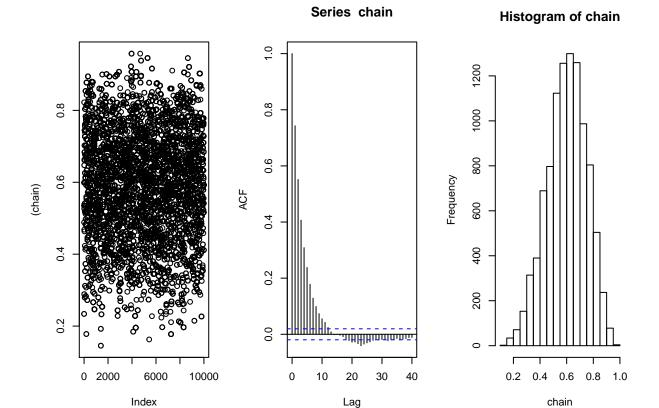
## Final 37810

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## Question 1

```
Alpha <- 6
Beta <- 4
# phi cannot be 0 or 1
phi <- function(c,oldphi){</pre>
  newphi = 0
  while (newphi == 0 \mid \mid newphi == 1){
  newphi = rbeta(1,c*oldphi,c*(1-oldphi))
  return(newphi)
}
run_metropolis_MCMC <- function(startvalue,c,iterations){</pre>
  chain = rep(0, iterations+1)
  chain[1] = startvalue
  for (i in 1:iterations){
    phi = phi(c,chain[i])
    proposal = dbeta(chain[i],c*phi,c*(1-phi))/dbeta(phi,c*chain[i],c*(1-chain[i]))
    posterior = dbeta(phi,Alpha,Beta)/dbeta(chain[i],Alpha,Beta)
    probab = proposal*posterior
    if (runif(1) < probab){</pre>
      chain[i+1] = phi
    }else{
      chain[i+1] = chain[i]
    }
  }
  return(chain)
}
startvalue = runif(1)
chain = run_metropolis_MCMC(startvalue,1, 10000)
acceptance = 1-mean(duplicated(chain))
par(mfrow=c(1,3)) #1 row, 3 columns
plot((chain)); acf(chain); hist(chain) #plot commands
```



## Gibbs Part

```
Gibbs_<-function(x0,y0,iterations,B=5,burnIn=0)</pre>
## 5 parameters are needed in this function, with B=5 and burnIn=0 by default
  if(x0>0\&&x0<B\&&y0>0\&&y0<B)
## to check whether the starting values are in the domain.
    x<-c(x0,rep(NA,iterations-1))</pre>
## Initialize the Markov chain
    y<-c(y0,rep(NA,iterations-1))
## Initialize the Markov chain
    for(i in 1:(iterations-1))
      x[i+1] < -(-log(1-runif(1)*(1-exp(-y[i]*B)))/y[i])
## use inverse transform sampling to draw sample from conditional distribution
## p(x^{i+1}|y^{i})
      y[i+1] < -(-log(1-runif(1)*(1-exp(-x[i+1]*B)))/x[i+1])
## use inverse transform sampling to draw sample from conditional distribution
## p(y^{i+1}|x^{i+1})
    if(burnIn>0)
        x<-x[-(1:burnIn)]</pre>
        y<-y[-(1:burnIn)]
        print(length(x))
## discard the first bunch of draws for the burn-in process
```

```
return(data.frame(x,y))
}
else
   stop("Initial values incorrect")
## print the information for incorrect starting values
}
```