

# Ma2

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Various packages in use:

```
set.seed(98765432)
library(igraph)
```

```
##
## Attaching package: 'igraph'
## The following objects are masked from 'package:stats':
##
##      decompose, spectrum
## The following object is masked from 'package:base':
##
##      union
```

```
library(expm)
```

```
## Loading required package: Matrix
##
## Attaching package: 'expm'
## The following object is masked from 'package:Matrix':
##
##      expm
```

w -> wide t -> tilde h -> hat m -> matrix

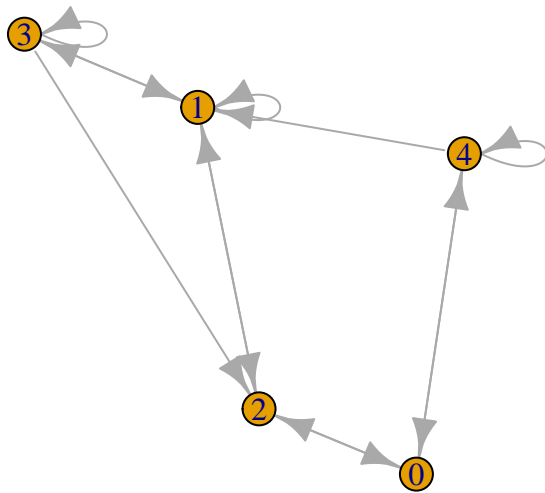
problem 2.1

a)

```
P <- matrix(nrow = 5, ncol = 5, dimnames = list(c("0","1","2","3","4"),
                                                c("0","1","2","3","4")))

P[1,] <- c(0.000, 0.000, 0.498, 0.000, 0.502)
P[2,] <- c(0.000, 0.527, 0.366, 0.107, 0.000)
P[3,] <- c(0.451, 0.549, 0.000, 0.000, 0.000)
P[4,] <- c(0.000, 0.474, 0.114, 0.412, 0.000)
P[5,] <- c(0.328, 0.249, 0.000, 0.000, 0.423)

graph_P <- graph_from_adjacency_matrix(P, weighted=TRUE, diag=TRUE)
plot(graph_P)
```



problem 2.1

b)

```
irreducibility <- function(P, M){
  P.new <- P
  for(i in 2:M){
    Pk <- P %^% i
    P.new <- P.new + Pk
  }

  P.bar <- (1/M)*P.new

  if(all(P.bar > 0)){
    print("Transition Matrix is Irreducible")
  }
  else{
    print("Transition Matrix is Reducible")
  }
}

irreducibility(P,5)

## [1] "Transition Matrix is Irreducible"
```

problem 2.1

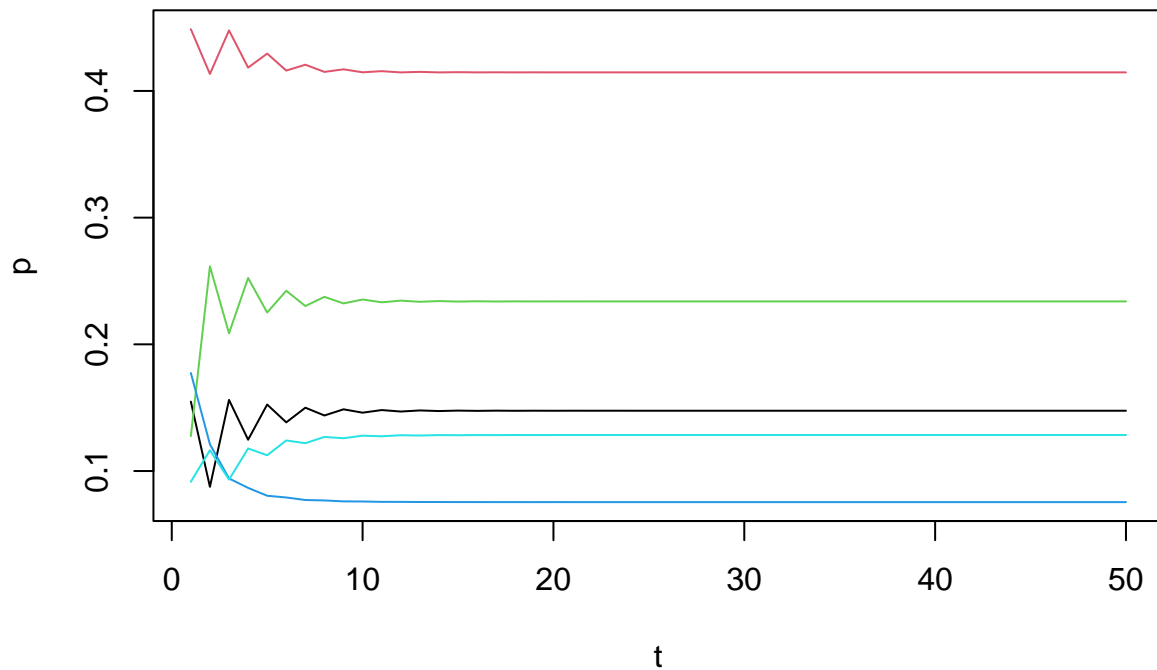
c)

```
K = 50
M = 5
p0 <- as.vector(c(0.013, 0.214, 0.197, 0.375, 0.201))

K <- 50
pk <- matrix(nrow = K, ncol = M)
for (i in 1:K) {
  pk[i, ] <- p0 %*% (P%^i)
}

plot(x = 1:K, y = pk[, 1], type = "l", col = 1, xlab = "t", ylab = "p",
ylim = c(min(pk), max(pk)))

lines(x = 1:K, y = pk[, 2], type = "l", col = 2)
lines(x = 1:K, y = pk[, 3], type = "l", col = 3)
lines(x = 1:K, y = pk[, 4], type = "l", col = 4)
lines(x = 1:K, y = pk[, 5], type = "l", col = 5)
```



## problem 2.1

d) Stationary distribution by computing eigenvalues of P

```
pi <- eigen(t(P))$vectors[,1]/sum(eigen(t(P))$vectors[,1])
```

```
pi_confirm <- pi %*% P
```

```
print(pi_confirm - pi)
```

```
##              0              1              2              3              4
## [1,] 5.551115e-17 -1.665335e-16 1.110223e-16 4.163336e-17 -5.551115e-17
```

e) Stationary distribution by iterating 50 times:  $\pi = \pi \cdot P^{50}$

```
pi_converged <- p0 %*% (P%^50)
```

As we see the difference is small

```
print(pi-pi_converged)
```

```
##              0              1              2              3              4
## [1,] 4.76654e-11 2.764483e-11 -5.232206e-11 -4.16657e-12 -1.882194e-11
```

problem 2.2

a) function for simulating large samples

```
simulMarkov<-function(p0,mP,n){  
  M<-dim(mP)[1]  
  X<-rep(0,n)  
  X[1]<-sample(M,1,prob=p0)  
  for(t in 2:n) X[t]<-sample(M,1,prob=mP[X[t-1],])  
  return(X)  
}
```

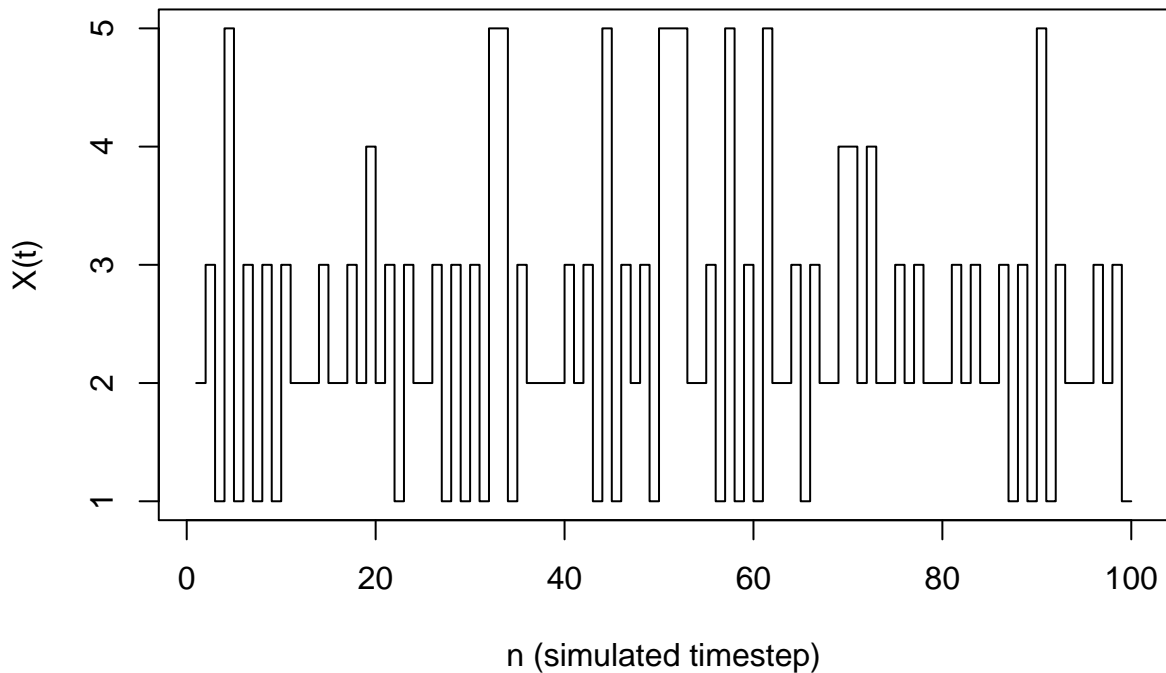
```
n<-100  
X<-rep(0,n)  
X[1]<-sample(M,1,prob=p0)  
for(t in 2:n) X[t]<-sample(M,1,prob=P[X[t-1],])
```

problem 2.2

b)

```
plot(x = 1:n, y = X, main = "Simulation of markov Chain based on (P, p0)",  
     xlab = "n (simulated timestep)", ylab = "X(t)", type = "S")
```

### Simulation of markov Chain based on (P, p0)



##problem 2.2

c)

```
iz<-4  
S_all<-which(X==iz)  
Tn<-length(S_all)
```

Tn

```
## [1] 4
```

problem 2.2

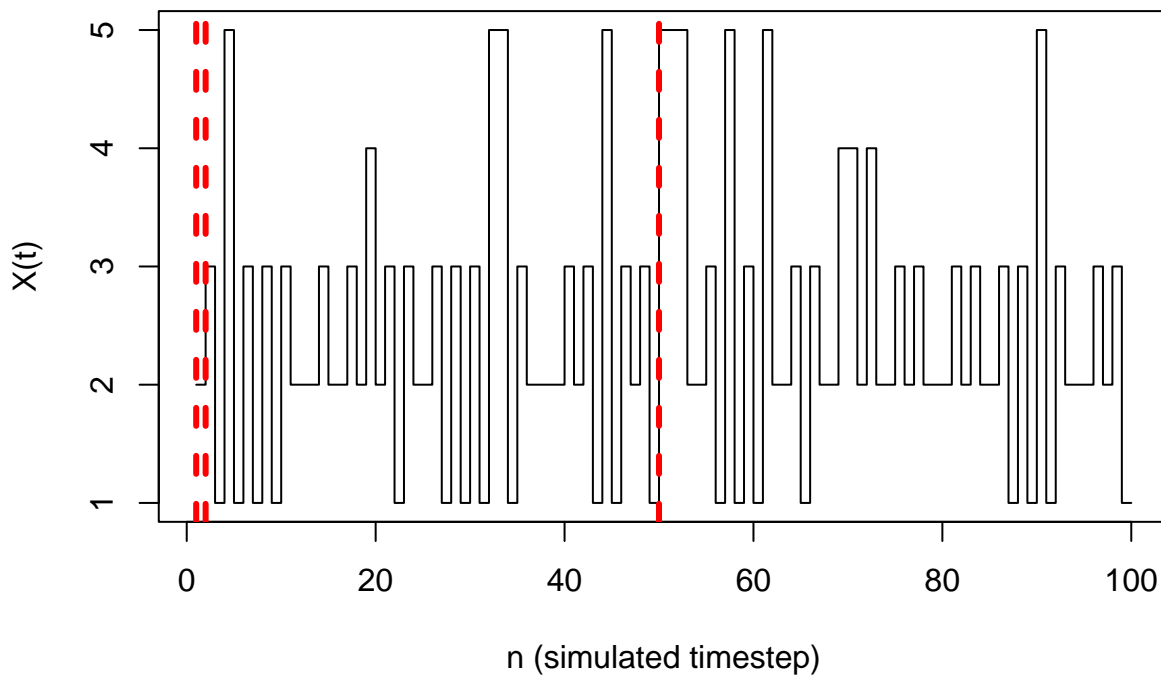
d)

```
tau<-S_all[2:Tn]-S_all[1:(Tn-1)]

plot(x = 1:n, y = X, main = "Simulation of markov Chain based on (P, p0)",
     xlab = "n (simulated timestep)", ylab = "X(t)", type = "S")

abline(v = tau, col="red", lwd=3, lty=2)
```

### Simulation of markov Chain based on (P, p0)



problem 2.3

a)

```
#set.seed(1234)
X.10000 <- simulMarkov(p0=p0, mP=P,n=10000)

iz.10000 = 4 #X.10000[1]
S_all<-which(X.10000 == iz.10000)

Tn.10000<-length(S_all)

Tn.10000
```

```
## [1] 834
```

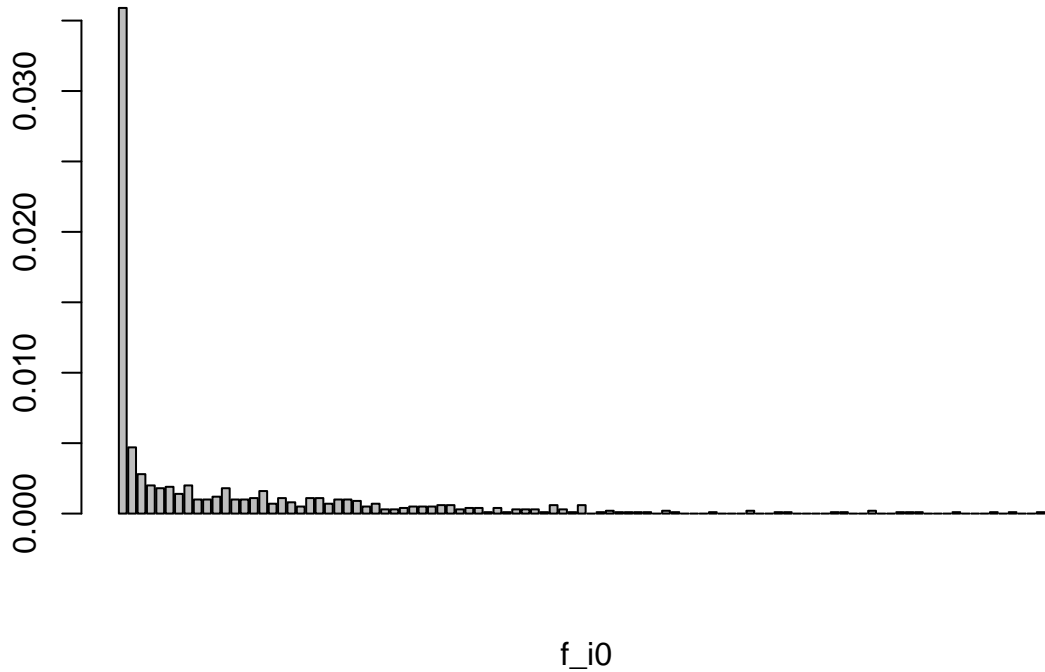
b)

```
tau.10000<-S_all[2:Tn.10000]-S_all[1:(Tn.10000-1)]

f_i0<-c(1:100)
for(j in 1:100){
f_i0[j]<-(1/10000)*length(which(tau.10000==j))
```

```
}
```

```
barplot(f_i0, xlab = "f_i0")
```



c)

```
set.seed(123)
i0 <- 4
Tn.1 <- length(which(simulMarkov(p0 = p0, mP=P, n=10000) == i0))
tau.1 <- S_all[2:Tn.1] - S_all[1:(Tn.1 - 1)]
K <- 100

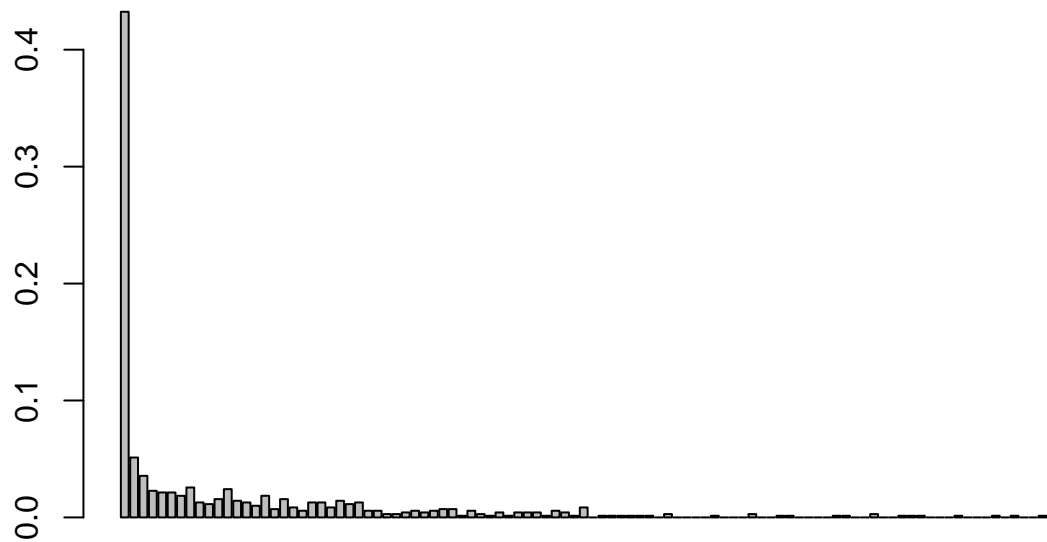
f_hat <- numeric(K)

for (s in 1:K) {
  f_hat[s] <- length(which(tau.1 == s))
}

f_hat <- f_hat / (Tn.1 - 1)

barplot(f_hat, xlab = "f_hat")
```





f\_hat

i)

```
mu_hat_i0 <- (1/(Tn.1-1))*sum(tau.1[-1]) #AKA tou_bar
mu_hat_i0
```

```
## [1] 12.21479
```

ii)

```
mu_tilde_i0 <- sum((1:K) * f_hat)
mu_tilde_i0
```

```
## [1] 11.5633
```

iii)

```
pi_i0 <- pi_converged[i0]
pi_i0
```

```
## [1] 0.07545784
```

```
rbind("1 / pi_i0" = 1 / pi_i0,
      "mu_hat_i0" = mu_hat_i0,
      "mu_tilde_i0" = mu_tilde_i0)
```

```
##           [,1]
## 1 / pi_i0 13.25243
## mu_hat_i0 12.21479
## mu_tilde_i0 11.56330
```

d)

```
n = 10000
m = n - 1
X = simulMarkov(p0 = p0, mP=P, n=n)
```

```

wtpi <- whpi <- numeric(5)
  for(i in 1:5){
    wtpi[i] <- m^(-1)*length(which(X==i))
  }

wtpi

## [1] 0.13981398 0.42854285 0.23202320 0.07760776 0.12211221

sum(wtpi)

## [1] 1.0001

```

e)

Emprirical Transition matrix:

```

P_hat <- matrix(nrow = 5, ncol = 5, 0)
wtpi <- c()
whpi <- c()
for (t in 1:(length(X) - 1)) P_hat[X[t],
  X[t + 1]] <- P_hat[X[t], X[t + 1]] + 1

  for (i in 1:5) P_hat[i, ] <- P_hat[i, ] / sum(P_hat[i, ])

```

P\_hat

```

##          [,1]      [,2]      [,3]      [,4]      [,5]
## [1,] 0.0000000 0.0000000 0.4907010 0.0000000 0.5092990
## [2,] 0.0000000 0.5376896 0.3633606 0.09894982 0.0000000
## [3,] 0.4312204 0.5687796 0.0000000 0.0000000 0.0000000
## [4,] 0.0000000 0.4484536 0.0992268 0.45231959 0.0000000
## [5,] 0.3259623 0.2571663 0.0000000 0.0000000 0.4168714

```

problem 2.4

we apply the amse matrix as in 2.3 (e)

a)

P\_hat

```

##          [,1]      [,2]      [,3]      [,4]      [,5]
## [1,] 0.0000000 0.0000000 0.4907010 0.0000000 0.5092990
## [2,] 0.0000000 0.5376896 0.3633606 0.09894982 0.0000000
## [3,] 0.4312204 0.5687796 0.0000000 0.0000000 0.0000000
## [4,] 0.0000000 0.4484536 0.0992268 0.45231959 0.0000000
## [5,] 0.3259623 0.2571663 0.0000000 0.0000000 0.4168714

```

comparing by subtracting  $P$  from  $\hat{P}$  b)

P-P\_hat

```

##          0          1          2          3          4
## 0 0.000000000 0.000000000 0.007298999 0.000000000 -0.007298999
## 1 0.000000000 -0.010689615 0.002639440 0.008050175 0.000000000
## 2 0.019779646 -0.019779646 0.000000000 0.000000000 0.000000000
## 3 0.000000000 0.025546392 0.014773196 -0.040319588 0.000000000
## 4 0.002037674 -0.008166257 0.000000000 0.000000000 0.006128583

```

as we see the difference is about two to three decimals