

Analytics for Competitive Advantage: Lab Exercise 4

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

Problem 1

In Markov100.txt, the one step transition probability matrix for a Markov chain with 100 states (State 1 to State 100) is given. Note that the data has no heading. Name of the data set Markov100. Number of rows 100. Number of columns 100

```
Markov100 <- read.table(file.path(filepath,"ex4_markov100.txt"),stringsAsFactors = F,header=F)
```

Problem 1(a)

Suppose we are at State 1 now. Find and display the probability of being in State 5 after 10 transitions.

```
# at state 1 now = initial vector  
a = c(1,rep(0,99))  
P = as.matrix(Markov100)  
# Calculate prob distribution after 10 steps given initial vector  
probDist10 <- a %*% (P %^10)  
# state 5  
state5 <- probDist10[5]  
state5
```

```
## [1] 0.045091
```

Problem 1(b)

Suppose we are at one of States 1,2, and 3 with equal probabilities. Find and display the probability of being in State 10 after 10 transitions.

```
# equal probabilities for 1,2,3  
three <- 1/3  
a2 = c(rep(three,3),rep(0,97))  
# Calculate prob distribution after 10 steps given initial vector  
probDist10.2 <- a2 %*% (P %^10)  
# state 10  
state10 <- probDist10.2[10]  
state10
```

```
## [1] 0.08268901
```

Problem 1(c)

Find the steady state probability of being in State 1.

```
# diag(n) # creates identity matrix with size n
# solve(M) # returns inverse of matrix M
n = 100
Q = t(P) - diag(n)
Q[n,] = c(rep(1,n))
rhs = c(rep(0,(n-1)),1)
# steady state probabilities
Pi = solve(Q) %%% rhs
# state 1
Pi[1]
```

```
## [1] 0.01256589
```

Problem 1(d)

Find the mean first passage time from State 1 to State 100.

```
# What is the expected number of steps to get to state 100?
B = P[1:(n-1),1:(n-1)]
Q = diag(n-1) - B
e = c(rep(1,n-1))
m = solve(Q) %*% e
# mean steps from 1 to 100
m[1]
```

```
## [1] 254.9395
```

Problem 2

You are asked to analyze the data from an website with 8 pages. Let us assume that there is a virtual page 9 that a visitor must automatically visit when the visitor leaves the website. The visitors always start their visit from Page 1. Let us formulate a Markov chain for this website.

```
webtraffic <- read.table(file.path(filepath,"ex4_webtraffic.txt"),stringsAsFactors = F,header=T)
```

Problem 2(a)

Construct 9 by 9 matrix Traffic that counts total traffic between State i to State j for all i and j. Display Traffic. Hint colSums() adds all rows for each column.

```
# Total traffic b/w state i to state j
# 1000 rows are 1000 obs. so need to sum them up
ij <- colSums(webtraffic)
# now put them in order
Traffic <- t(ij[1:9])
```

```

for(i in 2:9){
  Traffic <- rbind(Traffic,t(ij[((9*(i-1))+1):(9*i)]))
}
Traffic <- as.matrix(Traffic)
listname <- seq(1,9)
colnames(Traffic) <- paste("to",listname,sep="")
rownames(Traffic) <- paste("from",listname,sep="")
Traffic

```

```

##      to1 to2 to3 to4 to5 to6 to7 to8 to9
## from1  0 447 553  0  0  0  0  0  0
## from2  0 23 230 321  0  0  0  0 63
## from3  0 167 43 520  0  0  0  0 96
## from4  0  0  0 44 158 312 247  0 124
## from5  0  0  0  0 22 52 90 127 218
## from6  0  0  0  0 67 21  0 294 97
## from7  0  0  0  0  0 94 7 185 58
## from8  0  0  0  0 262  0  0 30 344
## from9  0  0  0  0  0  0  0  0  0

```

Problem 2(b)

Observe that Traffic has 0's in row 9 and 0's in column 1. Set $\text{Traffic}[9,1]=1000$. Construct the one step transition probability matrix P and display it.

```

Traffic[9,1] = 1000

P = Traffic
# prob matrix, x/n, n = row total
rowtot <- rowSums(P)
for(i in 1:9){
  P[i,] <- P[i,]/rowtot[i]
}
P = round(P,4)
P

```

```

##      to1    to2    to3    to4    to5    to6    to7    to8    to9
## from1  0 0.4470 0.5530 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000
## from2  0 0.0361 0.3611 0.5039 0.0000 0.0000 0.0000 0.0000 0.0989
## from3  0 0.2022 0.0521 0.6295 0.0000 0.0000 0.0000 0.0000 0.1162
## from4  0 0.0000 0.0000 0.0497 0.1785 0.3525 0.2791 0.0000 0.1401
## from5  0 0.0000 0.0000 0.0000 0.0432 0.1022 0.1768 0.2495 0.4283
## from6  0 0.0000 0.0000 0.0000 0.1399 0.0438 0.0000 0.6138 0.2025
## from7  0 0.0000 0.0000 0.0000 0.0000 0.2733 0.0203 0.5378 0.1686
## from8  0 0.0000 0.0000 0.0000 0.4119 0.0000 0.0000 0.0472 0.5409
## from9  1 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000

```

Problem 2(c)

Calculate and display the steady state probability vector Pi

```

n = 9
Q = t(P) - diag(n)
Q[n,] = c(rep(1,n))
rhs = c(rep(0,(n-1)),1)
# steady state probabilities
Pi = solve(Q) %*% rhs
rownames(Pi) = listname
Pi

```

```

##          [,1]
## 1 0.15833144
## 2 0.10086153
## 3 0.13079268
## 4 0.14012219
## 5 0.08057655
## 6 0.07583329
## 7 0.05445957
## 8 0.10069131
## 9 0.15833144

```

Problem 2(d)

The following table presents the average time that the visitors spend on each page. Page 1 2 3 4 5 6 7 8. Avg(minute) 0.1 2 3 5 5 3 3 2. Calculate and display the average time a visitor spend on the website (until she leaves).

```

# calcualte first mean passage time
# multiply that by the time
avgpg <- c(0.1, 2, 3, 5, 5, 3, 3, 2)

B = P[1:(n-1),1:(n-1)]
Q = diag(n-1) - B
# e is the average time spent on each page
e = avgpg
m = solve(Q) %*% e
# mean time spent starting from pg 1
passagevec <- t(m)
timespent <- passagevec[1]
timespent

```

```
## [1] 14.56242
```

Problem 2(e)

In the output of Problem 2(c), observe that Pages 3 and 4 are one of the most crowded pages except Pages 1 and 9. To balance the traffic, the owner of the website decided to create links from Page 2 to Pages 6,7 (hence, from State 2 to States 6,7). By adding the links, the owner anticipates that, from Page 2, 30% of the current outgoing traffic to State 3 would move to State 6, and 20% of the current outgoing traffic to State 4 would move to State 7. Calculate new steady state probability vector Π_2 to check the effect of the new links. Decide if the link helped balancing the traffic by comparing the variance of Π and Π_2 . Hint Start with matrix Traffic from Problem 2(a).

```

# .3 2,3 -> 2,6
# .2 2,4 -> 2,7
## first add the additional traffic to pg 6 and 7
Traffic[2,6] <- Traffic[2,6] + .3*Traffic[2,3]
Traffic[2,7] <- Traffic[2,7] + .2*Traffic[2,4]
## next reduce the current traffic
Traffic[2,3] <- .7*Traffic[2,3]
Traffic[2,4] <- .8*Traffic[2,4]

## recalc prob. matrix
P2 = Traffic
# prob matrix, x/n, n = row total
rowtot <- rowSums(P2)
for(i in 1:9){
  P2[i,] <- P2[i,]/rowtot[i]
}
P2 <- round(P2,4)

## recalculate steady state prob
n = 9
Q = t(P2) - diag(n)
Q[n,] = c(rep(1,n))
rhs = c(rep(0,(n-1)),1)
# steady state probabilities
Pi2 = solve(Q) %*% rhs
rownames(Pi2) <- listname
Pi2

```

```

##      [,1]
## 1 0.16163806
## 2 0.10035150
## 3 0.12105145
## 4 0.12275448
## 5 0.08163241
## 6 0.08250087
## 7 0.06002737
## 8 0.10840580
## 9 0.16163806

```

```

varPi = var(Pi)
varPi2 = var(Pi2)
varPi2 < varPi

```

```

##      [,1]
## [1,] TRUE

```

The variance decreased slightly, from 0.0014 to 0.0012.