

LabAssignment_1

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```
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.2.1 --
## v ggplot2 3.0.0    v purrr  0.2.5
## v tibble  1.4.2    v dplyr  0.7.6
## v tidyr   0.8.1    v stringr 1.3.1
## v readr   1.1.1    v forcats 0.3.0

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()

library(expm)

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

Problem 1

```
# load data
markov = read.table("~/Desktop/MSiA400/markov100.txt", head=FALSE)
P = as.matrix(markov) # transition probability matrix

names = c(1:100) # assign names to each state, from 1 to 100
colnames(P) = names
rownames(P) = names
```

1(a)

```
a <- rep(0, 100) # initial distribution
a[1] <- 1 # at State 1 with probability one
```

Since we are at State 1 now, in the initial distribution a , we are at State 1 with probability one.

```
prob5 <- a %**% (P%^%10)
prob5
```

```
##           1           2           3           4           5           6
## [1,] 0.03210252 0.03315294 0.09941539 0.06462136 0.045091 0.048171
##           7           8           9          10          11          12
## [1,] 0.02981315 0.04957693 0.0678567 0.08126983 0.1194159 0.02849685
##          13          14          15          16          17          18
## [1,] 0.03397078 0.09048926 0.07560999 0.01825296 0.02909586 0.01210839
##          19          20          21          22          23
## [1,] 0.01702716 0.0004803007 0.001104914 0.009981056 0.001970024
##          24          25          26          27          28
## [1,] 0.003182653 0.001596568 0.005224745 0.0009182241 1.76305e-06
##          29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48
## [1,] 1.761169e-06 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
##          49 50 51 52 53 54 55 56 57 58 59 60 61 62 63 64 65 66 67 68 69 70 71
## [1,] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
##          72 73 74 75 76 77 78 79 80 81 82 83 84 85 86 87 88 89 90 91 92 93 94
## [1,] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
##          95 96 97 98 99 100
## [1,] 0 0 0 0 0 0
```

```
prob5[1,5]
```

```
##           5
## 0.045091
```

Given we are at State 1 now, the probability of being in State 5 after 10 transitions is:

$$p_{1,5}(10) = 0.045091$$

1(b)

```
a2 <- rep(0, 100)
a2[1:3] <- 1/3
```

Since we are at one of States 1,2, and 3 with equal probabilities, the initial probability of being in one of these three is $\frac{1}{3}$

```
prob10 <- a2 %**% (P%^%10)
prob10
```

```
##           1           2           3           4           5           6
## [1,] 0.03365299 0.03473429 0.103148 0.06647602 0.04691916 0.04978609
##           7           8           9          10          11          12
## [1,] 0.03016254 0.04937492 0.06766467 0.08268901 0.1200149 0.02824977
##          13          14          15          16          17          18
## [1,] 0.03319286 0.08936531 0.07294184 0.01694442 0.0272714 0.01097558
##          19          20          21          22          23
## [1,] 0.01564754 0.000398366 0.0009532201 0.008850776 0.001593833
##          24          25          26          27          28
## [1,] 0.002601547 0.001301921 0.004332957 0.0007160195 2.002671e-05
##          29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48
## [1,] 2.000535e-05 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
```

```
##      49 50 51 52 53 54 55 56 57 58 59 60 61 62 63 64 65 66 67 68 69 70 71
## [1,]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##      72 73 74 75 76 77 78 79 80 81 82 83 84 85 86 87 88 89 90 91 92 93 94
## [1,]  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0  0
##      95 96 97 98 99 100
## [1,]  0  0  0  0  0  0
```

```
prob10[1,10]
```

```
##      10
## 0.08268901
```

Given we are initially at State 1,2,3 with equal probability, the probability of being in State 10 after 10 transitions is:

$$p_{i,10}(10) = 0.08268901$$

where i=1,2,3 with equal probability

1(c)

```
Q = t(P)-diag(100)
Q[100,] = rep(1,100) # replace the last row with all ones
rhs = rep(0,100)
rhs[100] = 1
Pi = solve(Q) %%% rhs
Pi # steady state probabilities
```

```
##      [,1]
## 1  0.0125658938
## 2  0.0091915303
## 3  0.0255231969
## 4  0.0187240817
## 5  0.0115751391
## 6  0.0121239209
## 7  0.0083136486
## 8  0.0152539239
## 9  0.0201849886
## 10 0.0227006031
## 11 0.0338552495
## 12 0.0083157699
## 13 0.0117828234
## 14 0.0289037115
## 15 0.0298638379
## 16 0.0095552449
## 17 0.0139102321
## 18 0.0079487586
## 19 0.0126688098
## 20 0.0047145165
## 21 0.0039498476
## 22 0.0143966186
## 23 0.0120881289
## 24 0.0173623697
## 25 0.0069709745
```

26 0.0265861556
27 0.0130348800
28 0.0037966314
29 0.0036423185
30 0.0010794500
31 0.0016783263
32 0.0074044721
33 0.0068897274
34 0.0014189398
35 0.0017930324
36 0.0109944087
37 0.0120631618
38 0.0129222379
39 0.0178424966
40 0.0082080694
41 0.0138441177
42 0.0046340255
43 0.0182176477
44 0.0023285120
45 0.0031248893
46 0.0009744419
47 0.0018880951
48 0.0037302200
49 0.0017600842
50 0.0037672559
51 0.0033658985
52 0.0036948907
53 0.0018605719
54 0.0047856565
55 0.0050767072
56 0.0023760149
57 0.0028493610
58 0.0101897104
59 0.0044646936
60 0.0108659367
61 0.0157483299
62 0.0080770433
63 0.0324202507
64 0.0214827961
65 0.0234131029
66 0.0624730859
67 0.0629301714
68 0.0220613102
69 0.0072403690
70 0.0132948300
71 0.0123211134
72 0.0163058220
73 0.0168143679
74 0.0062696308
75 0.0119441327
76 0.0048514374
77 0.0021763481
78 0.0014480905
79 0.0010413500

```
## 80 0.0026629249
## 81 0.0019927676
## 82 0.0007585553
## 83 0.0014826699
## 84 0.0013118807
## 85 0.0013298209
## 86 0.0025981346
## 87 0.0030361321
## 88 0.0062437668
## 89 0.0040033770
## 90 0.0044848282
## 91 0.0025584427
## 92 0.0008277174
## 93 0.0017875498
## 94 0.0015061251
## 95 0.0022677010
## 96 0.0032878670
## 97 0.0015725118
## 98 0.0019396680
## 99 0.0025299448
## 100 0.0039071739
```

As shown above, the steady state probability of being in State 1 is: $\pi_1 = 0.0125658938$

1(d)

```
B = P[1:99,1:99] # exclude destination State 100
Q2 = diag(99) - B
e = rep(1,99)
m = solve(Q2) %*% e
m # mean first passage time from State i to State 100, i != 100
```

```
##      [,1]
## 1 254.939463
## 2 255.756780
## 3 255.553434
## 4 252.020994
## 5 254.689848
## 6 253.741872
## 7 253.572908
## 8 251.866892
## 9 249.390794
## 10 247.986623
## 11 244.808027
## 12 249.271139
## 13 244.048246
## 14 242.112390
## 15 232.030348
## 16 233.108471
## 17 217.761197
## 18 230.098329
## 19 207.242442
## 20 207.618923
```

21 202.962858
22 195.385117
23 201.024569
24 197.401354
25 197.152996
26 190.532650
27 172.667559
28 158.833969
29 153.487433
30 152.755309
31 156.096022
32 150.594867
33 149.507037
34 147.958820
35 148.265543
36 149.636556
37 146.005518
38 147.766862
39 143.005036
40 143.644636
41 148.465041
42 146.009273
43 138.935358
44 121.384733
45 120.998033
46 121.322578
47 129.179597
48 118.386090
49 120.272528
50 115.940519
51 115.605339
52 120.678477
53 112.276507
54 112.637460
55 111.270399
56 109.257070
57 113.623850
58 109.555265
59 109.371776
60 106.390883
61 107.104019
62 106.498192
63 103.143540
64 103.791399
65 104.727505
66 102.882204
67 99.471192
68 99.637840
69 72.153204
70 94.184715
71 33.476861
72 54.763689
73 50.445604
74 47.521308

```
## 75 46.317640
## 76 44.836599
## 77 16.035969
## 78 19.133208
## 79 14.031529
## 80 12.607691
## 81 11.518157
## 82 13.244137
## 83 12.369547
## 84 13.130539
## 85 11.202843
## 86 11.427886
## 87 10.839776
## 88 9.471288
## 89 8.418722
## 90 6.158265
## 91 6.321574
## 92 5.554380
## 93 4.674446
## 94 6.190252
## 95 4.431775
## 96 3.795454
## 97 3.521875
## 98 2.931884
## 99 2.389126
```

From above, the mean first passage time from State 1 to State 100 is: $m_{1,100} = 254.939463$

Problem 2

```
# load data
webtraffic <- read.table("~/Desktop/MSiA400/webtraffic.txt", head=TRUE)
```

2(a)

```
columnSum <- unname(colSums(webtraffic))
Traffic <- t(matrix(columnSum, nrow = 9, ncol = 9))
Traffic
```

##		[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]	[,8]	[,9]
##	[1,]	0	447	553	0	0	0	0	0	0
##	[2,]	0	23	230	321	0	0	0	0	63
##	[3,]	0	167	43	520	0	0	0	0	96
##	[4,]	0	0	0	44	158	312	247	0	124
##	[5,]	0	0	0	0	22	52	90	127	218
##	[6,]	0	0	0	0	67	21	0	294	97
##	[7,]	0	0	0	0	0	94	7	185	58
##	[8,]	0	0	0	0	262	0	0	30	344
##	[9,]	0	0	0	0	0	0	0	0	0

From above, matrix Traffic that counts total traffic between State i to State j for all i = 1,2,3,4,5,6,7,8,9 and j = 1,2,3,4,5,6,7,8,9 is:

$$Traffic = \begin{bmatrix} 0 & 447 & 553 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 23 & 230 & 321 & 0 & 0 & 0 & 0 & 63 \\ 0 & 167 & 43 & 520 & 0 & 0 & 0 & 0 & 96 \\ 0 & 0 & 0 & 44 & 158 & 312 & 247 & 0 & 124 \\ 0 & 0 & 0 & 0 & 22 & 52 & 90 & 127 & 218 \\ 0 & 0 & 0 & 0 & 67 & 21 & 0 & 294 & 97 \\ 0 & 0 & 0 & 0 & 0 & 94 & 7 & 185 & 58 \\ 0 & 0 & 0 & 0 & 262 & 0 & 0 & 30 & 344 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

2(b)

```
Traffic[9,1] <- 1000

for (i in 1:nrow(Traffic)){
  rowSum = sum(Traffic[i,])
  Traffic[i,] = Traffic[i,] / rowSum
}

Traffic # P

##      [,1]      [,2]      [,3]      [,4]      [,5]      [,6]
## [1,]    0 0.44700000 0.55300000 0.00000000 0.00000000 0.00000000
## [2,]    0 0.03610675 0.36106750 0.50392465 0.00000000 0.00000000
## [3,]    0 0.20217918 0.05205811 0.62953995 0.00000000 0.00000000
## [4,]    0 0.00000000 0.00000000 0.04971751 0.1785311 0.35254237
## [5,]    0 0.00000000 0.00000000 0.00000000 0.0432220 0.10216110
## [6,]    0 0.00000000 0.00000000 0.00000000 0.1398747 0.04384134
## [7,]    0 0.00000000 0.00000000 0.00000000 0.0000000 0.27325581
## [8,]    0 0.00000000 0.00000000 0.00000000 0.4119497 0.00000000
## [9,]    1 0.00000000 0.00000000 0.00000000 0.0000000 0.00000000
##      [,7]      [,8]      [,9]
## [1,] 0.00000000 0.00000000 0.00000000
## [2,] 0.00000000 0.00000000 0.0989011
## [3,] 0.00000000 0.00000000 0.1162228
## [4,] 0.27909605 0.00000000 0.1401130
## [5,] 0.17681729 0.24950884 0.4282908
## [6,] 0.00000000 0.61377871 0.2025052
## [7,] 0.02034884 0.53779070 0.1686047
## [8,] 0.00000000 0.04716981 0.5408805
## [9,] 0.00000000 0.00000000 0.00000000
```

One step transition probability matrix is:

$$P = \begin{bmatrix} 0 & 0.44700000 & 0.55300000 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0.03610675 & 0.36106750 & 0.50392465 & 0 & 0 & 0 & 0 & 0.0989011 \\ 0 & 0.20217918 & 0.05205811 & 0.62953995 & 0 & 0 & 0 & 0 & 0.1162228 \\ 0 & 0 & 0 & 0.04971751 & 0.1785311 & 0.35254237 & 0.27909605 & 0 & 0.1401130 \\ 0 & 0 & 0 & 0 & 0.0432220 & 0.10216110 & 0.17681729 & 0.24950884 & 0.4282908 \\ 0 & 0 & 0 & 0 & 0.1398747 & 0.04384134 & 0 & 0.61377871 & 0.2025052 \\ 0 & 0 & 0 & 0 & 0 & 0.27325581 & 0.02034884 & 0.53779070 & 0.1686047 \\ 0 & 0 & 0 & 0 & 0.4119497 & 0 & 0 & 0.04716981 & 0.5408805 \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

2(c)

```
Q3 = t(Traffic)-diag(9)
Q3[9,] = rep(1,9)
rhs2 = rep(0,9)
rhs2[9] = 1
Pi2 = solve(Q3) %*% rhs2
Pi2
```

```
##           [,1]
## [1,] 0.15832806
## [2,] 0.10085497
## [3,] 0.13077897
## [4,] 0.14012033
## [5,] 0.08058898
## [6,] 0.07583914
## [7,] 0.05446485
## [8,] 0.10069664
## [9,] 0.15832806
```

From above, the steady state probability vector is:

$$\pi = \begin{bmatrix} 0.15832806 \\ 0.10085497 \\ 0.13077897 \\ 0.14012033 \\ 0.08058898 \\ 0.07583914 \\ 0.05446485 \\ 0.10069664 \\ 0.15832806 \end{bmatrix}$$

2(d)

```
avg_time <- c(0.1,2,3,5,5,3,3,2,0)
expected <- t(Pi2) %*% avg_time
expected
```

```
##           [,1]
## [1,] 2.305731
```

The average time a visitor spend on a page is 2.305731 minutes

```
B2 = Traffic[1:8,1:8] # exclude destination State 9
Q_m = diag(8) - B2
e2 = rep(1,8)
m2 = solve(Q_m) %*% e2
m2 # mean first passage time from State i to State 9, i != 9
```

```
##           [,1]
## [1,] 5.316000
## [2,] 4.401776
## [3,] 4.246666
## [4,] 3.392390
## [5,] 2.429794
## [6,] 2.749343
## [7,] 2.940475
## [8,] 2.100010
```

From above, the mean first passage time from Page 1 to Page 9 is: $m_{1,9} = 5.316$

Therefore, average time a visitor spend on the website (until she leaves) is:

$5.316 \times 2.305731 = 12.25727$ minutes

2(e)

In the output 2(c), Pages 3 and 4 have higher values than others, excluding Pages 1 and 9.

```
Traff <- t(matrix(columnSum, nrow = 9, ncol = 9))
Traff # Traffic matrix
```

```
##           [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9]
## [1,]      0  447  553      0      0      0      0      0      0
## [2,]      0   23  230  321      0      0      0      0     63
## [3,]      0  167   43  520      0      0      0      0     96
## [4,]      0    0    0   44  158  312  247      0    124
## [5,]      0    0    0    0   22   52   90  127    218
## [6,]      0    0    0    0   67   21    0  294     97
## [7,]      0    0    0    0    0   94    7  185     58
## [8,]      0    0    0    0  262    0    0   30    344
## [9,]      0    0    0    0    0    0    0    0      0
```

```
two_to_three <- Traff[2,3] #current outgoing traffic to State3 from State2
two_to_four <- Traff[2,4] #current outgoing traffic to State4 from State2
# new assignments after linking Page 2 to 6,7
Traff[2,3] <- two_to_three*0.7
Traff[2,6] <- two_to_three*0.3
Traff[2,4] <- two_to_four*0.8
Traff[2,7] <- two_to_four*0.2
Traff
```

```
##           [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9]
## [1,]      0  447  553  0.0      0      0  0.0      0      0
## [2,]      0   23  161 256.8      0   69  64.2      0     63
## [3,]      0  167   43 520.0      0    0  0.0      0     96
## [4,]      0    0    0  44.0  158  312 247.0      0    124
```

```
## [5,] 0 0 0 0.0 22 52 90.0 127 218
## [6,] 0 0 0 0.0 67 21 0.0 294 97
## [7,] 0 0 0 0.0 0 94 7.0 185 58
## [8,] 0 0 0 0.0 262 0 0.0 30 344
## [9,] 0 0 0 0.0 0 0 0.0 0 0
```

calculate new steady state probability

```
Traff[9,1] <- 1000
# calculate one step transtion probability matrix and stored in Traff
for (i in 1:nrow(Traff)){
  rowSum = sum(Traff[i,])
  Traff[i,] = Traff[i,] / rowSum
}

Q4 = t(Traff)-diag(9)
Q4[9,] = rep(1,9)
rhs3 = rep(0,9)
rhs3[9] = 1
Pi3 = solve(Q4) %*% rhs3
Pi3
```

```
##           [,1]
## [1,] 0.16162840
## [2,] 0.10034341
## [3,] 0.12104331
## [4,] 0.12275720
## [5,] 0.08164613
## [6,] 0.08250884
## [7,] 0.06003218
## [8,] 0.10841213
## [9,] 0.16162840
```

The new steady state probability vector is:

$$\pi_2 = \begin{bmatrix} 0.16162840 \\ 0.10034341 \\ 0.12104331 \\ 0.12275720 \\ 0.08164613 \\ 0.08250884 \\ 0.06003218 \\ 0.10841213 \\ 0.16162840 \end{bmatrix}$$

Comparing with original π , after creating new links from Page 2 to Page 6,7, steady state probabilities in Page 3,4 decrease and steady state probability in Page 2 slightly decreases. Steady state probabilities in other Pages all increase.

```
# variance of pi before change
var(Pi2)
```

```
##           [,1]
## [1,] 0.001410675
```

```
# variance of pi after change  
var(Pi3)
```

```
##           [,1]  
## [1,] 0.001219604
```

From above, $\text{var}(\pi) = 0.001410675 > \text{var}(\pi_2) = 0.001219604$

As the variance of steady state probability decreases after creating new links, there is less variation after introducing the new links. So, the link helped balancing the traffic.