

STAT220 Oblig 4-5

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Library in use:

```
library(tinytex)
library(matrixcalc)
library(shape)
library(diagram)
library(igraph)
```

```
##
## Attaching package: 'igraph'

## The following object is masked from 'package:matrixcalc':
##
##      %s%

## The following objects are masked from 'package:stats':
##
##      decompose, spectrum

## The following object is masked from 'package:base':
##
##      union
```

```
library(devtools)
```

```
## Loading required package: usethis
```

Problem 6.1

Vi har gitt transition matrix P:

```
P=matrix(c(0.7, 0.2, 0.1, 0.0, 0.6, 0.4, 0.5, 0.0, 0.5), nrow=3, ncol=3, byrow = TRUE)
dimnames(P)=list(c('0','1','2'), c('0','1','2'))
print(P)
```

```
##      0    1    2
## 0 0.7 0.2 0.1
## 1 0.0 0.6 0.4
## 2 0.5 0.0 0.5
```

a) Since $P_{i,j}^m = P(x_{m+n} = j | x_n = i)$ and $P^{(m)} = P^m$:

$\Rightarrow P^{(3)} = P^3$ og vi får følgende matrise

```
P_3 = matrix.power(P,3)
print(P_3)
```

```
##      0    1    2
## 0 0.478 0.264 0.258
## 1 0.360 0.256 0.384
## 2 0.570 0.180 0.250
```

```
print(P_3[1,2])
```

```
## [1] 0.264
```

slik at: $P(x_3 = 1 | x_0 = 0) = \underline{\underline{0.264}}$

samme egenskap gjelder også for $P^{(3)} = P^3$

```
P_4 = matrix.power(P,4)
print(P_4)
```

```
##      0    1    2
## 0 0.4636 0.2540 0.2824
## 1 0.4440 0.2256 0.3304
## 2 0.5240 0.2220 0.2540
```

```
print(P_4[1,2])
```

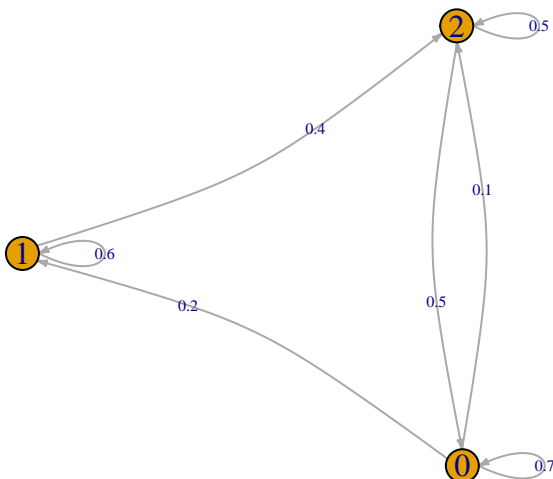
```
## [1] 0.254
```

og vi får da:

$P(x_3 = 1 | x_0 = 0) = \underline{\underline{0.254}}$

b) As we see from the graphic presentation below, it is possible to go from any state to another through one or more steps:

```
h <- graph_from_adjacency_matrix(P, weighted = "prob")
E(h)$prob <- ifelse(is.nan(E(h)$prob), NA, E(h)$prob)
plot(h, edge.label = round(E(h)$prob, 2), edge.arrow.size = .25, edge.curved=-0.2, edge.label.cex = .5)
```



from state 0: $0 \rightarrow 2 \rightarrow 1$

from state 1: $1 \rightarrow 0 \rightarrow 2$

from state 2: $2 \rightarrow 1 \rightarrow 0$ and $1 \rightarrow 0$

It is therefore irreducible.

Periodicity: Since the period of a state is the largest d that satisfy following properties: - $p_{ii}^{(n)} = 0$ whenever n is not divisible by d . - The period i is shown by $d(i)$. - If $P_{ii}^{(n)} = 0$, for all $n > 0 \rightarrow i = \infty$

and . . .

i is periodic if $d(i) > 1$ and aperiodic if $d(i) = 1$

It is aperiodic since there are several sequences of steps to go from a state and back again ($i \rightarrow i$), including the fact that all the states are self periodic the Markov chain is aperiodic.

Transient or recurrent? \rightarrow It is recurrent! It might stop and loop at a position or between state 2 and 0 which are communicating, but eventually it will occur in any of the states by a certainty of 100% as in the formal definition: $f_{ii} = P(X_n = i, \text{ for some } n \geq 1 | w_0 = i)$ for any state i is Recurrent if $f_{ii} = 1$ and transient if Recurrent if $f_{ii} < 1$.

Problem 6.?

a) $G : \text{Green } R : \text{Red}$

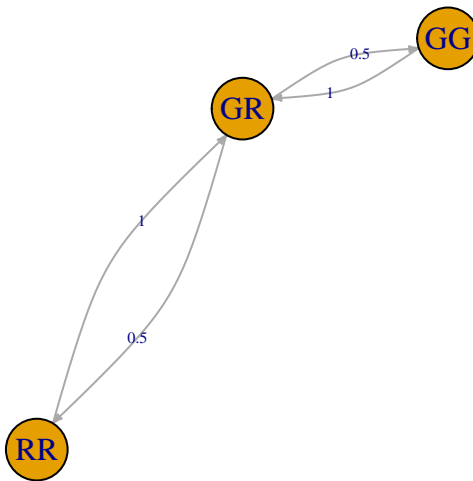
We start initially at GR and pick a random ball with $p = 1/2$. We are then sent to a state of RR or GG where the probability of being returned to State GR is 100%

```

A=matrix(c(0, 1, 0, 0.5, 0, 0.5, 0, 1, 0), nrow=3, ncol=3, byrow = TRUE)
dimnames(A)=list(c('GG','GR','RR'), c('GG','GR','RR'))

f <- graph_from_adjacency_matrix(A, weighted = "prob")
E(f)$prob <- ifelse(is.nan(E(f)$prob), NA, E(f)$prob)
plot(f, edge.label = round(E(f)$prob, 2), vertex.size=30, edge.arrow.size = .25, edge.curved=0.3, edge.l

```



Wich gives following matrix

```
print(A)
```

```

##      GG GR  RR
## GG 0.0  1 0.0
## GR 0.5  0 0.5
## RR 0.0  1 0.0

```

and following for picking red ball in the future entering the state of $GR \rightarrow P(X_n = j | X_0 = 1)$: #her må jeg få sendt ett spørsmål

```
print(A[2,])
```

```

##  GG  GR  RR
## 0.5 0.0 0.5

```

b) $d(GR) = 2$ og $d(GG) = d(RR) = 4$ som gir største felles nevner 2 og dermed er den periodisk med periode 2.

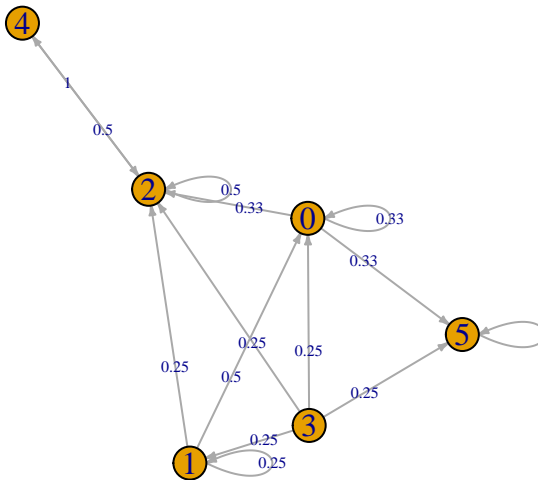
Problem 6.2

Due to previous use of P as variable, it is here substituted for B a)

```
B=round(matrix(c((1/3), 0.0, (1/3), 0.0, 0.0, (1/3),
                (1/2), (1/4), (1/4), 0.0, 0.0, 0.0,
                0.0, 0.0, (1/2), 0.0, (1/2), 0.0,
                (1/4), (1/4), (1/4), 0.0, 0.0, (1/4),
                0.0, 0.0, 1.0, 0.0, 0.0, 0.0,
                0.0, 0.0, 0.0, 0.0, 0.0, 1.0), nrow=6, ncol=6, byrow = TRUE),2)
dimnames(B)=list(c('0','1','2','3','4','5'), c('0','1','2','3','4','5'))
print(B)
```

```
##      0      1      2 3      4      5
## 0 0.33 0.00 0.33 0 0.0 0.33
## 1 0.50 0.25 0.25 0 0.0 0.00
## 2 0.00 0.00 0.50 0 0.5 0.00
## 3 0.25 0.25 0.25 0 0.0 0.25
## 4 0.00 0.00 1.00 0 0.0 0.00
## 5 0.00 0.00 0.00 0 0.0 1.00
```

```
g <- graph_from_adjacency_matrix(B, weighted = "prob")
E(g)$prob <- ifelse(is.nan(E(g)$prob), NA, E(g)$prob)
plot(g, edge.label = round(E(g)$prob, 2), edge.arrow.size = .25, edge.label.cex = .5)
```



b) As we se from the graph above:

We have three recurrent states, 2, 4 and 5, divided in two classes.
class 1 - State 5 is self-recurrent class 2 - State 2 and 4

Transient States:

State 0, 1, and 3 will at some pint possibly loop, but ultimatly it will end up in recurrent state class 1 or 2.

c) Communicating States:

(2 ↔ 4), (1 ↔ 1), (2 ↔ 2) and (5 ↔ 5)

d) Matrix reallocated to statespace = {1,3,0,2,4,5}: (Matrix denoted - C to separate matrixes in the code)

```
C=round(matrix(c((1/4), 0.0, (1/2), (1/4), 0.0, 0.0,
(1/4), 0.0, (1/4), (1/4), 0.0, (1/4),
0.0, 0.0, (1/3), (1/3), 0.0, (1/3),
0.0, 0.0, 0.0, (1/2), (1/2), 0.0,
0.0, 0.0, 0.0, 1.0, 0.0, 0.0,
0.0, 0.0, 0.0, 0.0, 0.0, 1.0), nrow=6, ncol=6, byrow = TRUE),2)
dimnames(C)=list(c('1','3','0','2','4','5'), c('1','3','0','2','4','5'))
print(C)
```

```
##      1 3    0    2    4    5
## 1 0.25 0 0.50 0.25 0.0 0.00
## 3 0.25 0 0.25 0.25 0.0 0.25
## 0 0.00 0 0.33 0.33 0.0 0.33
## 2 0.00 0 0.00 0.50 0.5 0.00
## 4 0.00 0 0.00 1.00 0.0 0.00
## 5 0.00 0 0.00 0.00 0.0 1.00
```

e)

```
R<-round(matrix(c(C[1, 4:6],
C[2, 4:6],
C[3, 4:6]), nrow=3, ncol=3, byrow = TRUE), 2)
dimnames(R)=list(c('1','3','0'), c('2','4','5'))
print(R)
```

```
##      2 4    5
## 1 0.25 0 0.00
## 3 0.25 0 0.25
## 0 0.33 0 0.33
```

This gives $P_0(X_T = 5) = P(X_T = 5 | X_{T-1} = 0) = 0.3333 = \frac{1}{3}$

and the expectation:

$$\begin{aligned} E(X_T = 5 | X_{T-1}) &= 1 + \sum_{j \in S} E(T_P = j | X_{T-1} = 0) \cdot P(X_T = j | X_{T-1} = 0) \\ &= 1 + \sum_{j \in S} m_{0j} \cdot P(X_T = j | X_{T-1} = 0) \end{aligned}$$

We are proceeding for state 5 such that $m_{j,0} = 1$, $m_{2,0} = 0$.

$$\begin{aligned} E(X_T = 5 | X_{T-1}) &= 1 + m_{0,0} \cdot P(X_T = 0 | X_{T-1} = 0) \\ &+ m_{2,0} \cdot P(X_T = 2 | X_{T-1} = 0) + m_{5,0} \cdot P(X_T = 5 | X_{T-1} = 0) \\ &= 1 + 1 \cdot \frac{1}{3} + 1 \cdot \frac{1}{3} + 0 \cdot \frac{1}{3} \\ &= \underline{\underline{\frac{5}{3} \approx 1,667}} \end{aligned}$$

f) for p^∞ :

```
P_infinity = round(matrix.power(B,1000000),2)
print(P_infinity)
```

```
##   0 1   2 3   4   5
## 0 0 0 0.33 0 0.16 0.49
## 1 0 0 0.44 0 0.22 0.33
## 2 0 0 0.67 0 0.33 0.00
## 3 0 0 0.36 0 0.18 0.46
## 4 0 0 0.67 0 0.33 0.00
## 5 0 0 0.00 0 0.00 1.00
```

We see the matrix converges in the recurrent states in class 1 and 2.

problem 6.3