



When first throw is 4,

chance of repeating = (number of throws not resulting in 4 or 7)/36  
=  $27/36 = 3/4$

Similarly, chance of loosing =  $1/6$

chance of winning =  $1/12$

We have calculated this for 10,5,9,6,8 as well.

## 2. Expected number of throws per game :

7,11,2,3,12 decides outcome in first throw. We can obtain 7,11,2,3,12 in 12 out of 36 ways.

The other 24 throws will go to point.

Number of outcomes for 4 =  $\{(1,3), (3,1), (2,2)\} = 3$

Number of outcomes for 5 =  $\{(1,4), (4,1), (2,3), (3,2)\} = 4$

Number of outcomes for 6 =  $\{(1,5), (5,1), (2,4), (4,2), (3,3)\} = 5$

Number of outcomes for 8 =  $\{(2,6), (6,2), (3,5), (5,3), (4,4)\} = 5$

Number of outcomes for 9 =  $\{(3,6), (6,3), (4,5), (5,4)\} = 4$

Number of outcomes for 10 =  $\{(4,6), (6,4), (5,5)\} = 3$

When the point is 4 or 10,

Number of wins = 3

Number of loss = 6

Therefore 9 out of 36 outcome decides.

Hence Expected Throws =  $36/9 = 4$

When the point is 5 or 9,

10 out of 36 outcome decides. Expected throws = 3.6

When point is 6 or 8,

11 out of 36 outcomes decides. Expected throws = 3.27

Therefore adding these together,  
Expected number of throws after point is established,

$$4 * (6/24) + 3.6 * (8/24) + 3.27 * (10/24) = 3.57$$

But  $24/36 = 2/3$  throws go up to a point. Therefore expected number of point throws  
=  $3.57 * (2/3) = 2.38$

Now there is a first expected throw. The sum of these two gives total expected  
number of throws =  $1 + 2.38 = 3.38$

### 3. Probability of winning :

Probability of win with an outcome of 7 or 11 =  $2/9$

Establish point 4 or 10 an win =  $(1/12) * (1/3) = (1/36)$

Establish point 5 or 9 an win =  $(1/9) * (2/5) = (2/45)$

Establish point 6 or 8 an win =  $(5/36) * (5/11) = (25/396)$

Therefore probability of winning =  $(2/9) + 2(1/36 + 2/45 + 25/396)$   
=  $0.222 + 2(0.028 + 0.044 + 0.063)$   
= **0.492**

Question 2. 1. Let received = p.

We have  $AFp \equiv \neg EG(\neg p)$

Now,  $EG(\neg p) = EG(\{s_0, s_1\})$

Now we have, one transient class and one recurrent class. For recurrent class  $s_2$  does not satisfy  $\{s_0, s_1\}$ .

For the transient states that satisfies  $\{s_0, s_1\}$  is  $s_0$  and  $s_1$ . But none of them can reach a recurrent class that satisfy  $\{s_0, s_1\}$ .

Therefore no state satisfies  $EG(\{s_0, s_1\})$ . And all states satisfy  $\neg EG(\neg p)$ , thus  $AFp$ .

Therefore the model satisfies  $AFp$ .

2. We are given the initial probability distribution  $p_0 = [1 \ 0 \ 0]$

Therefore  $p_o[Z] = [1 \ 0]$ .

We have transition probability matrix,  $P = \begin{bmatrix} 0 & 1 & 0 \\ \alpha & 0 & 1-\alpha \\ 0 & 0 & 1 \end{bmatrix}$

We have  $P[Z, Z] = \begin{bmatrix} 0 & 1 \\ \alpha & 0 \end{bmatrix}$

From the section 14.4 of the lecture we know that,  $n(P[Z, Z] - I) = -p_0[Z]$ .

Let the vector  $n = [a, b]$ .

Now,  $(P[Z, Z] - I) = \begin{bmatrix} -1 & 1 \\ \alpha & -1 \end{bmatrix}$

Therefore,  $[a \ b] \begin{bmatrix} -1 & 1 \\ \alpha & -1 \end{bmatrix} = [1 \ 0]$

By solving the equation we get  $a = 1/(1-\alpha)$  and  $b = a$ .  
Therefore average time =  $2/(1-\alpha)$ .

3. Initial probability distribution is given as  $p_0 = [1 \ 0 \ 0]$

Now we need to calculate the limiting probability  $p_\infty$  which is given by the formula  $p_\infty = [0, n.P[Z, A] + p_0[A]]$

From the previous problem we have calculated  $n$  to be  $[1/(1-\alpha), 1/(1-\alpha)]$

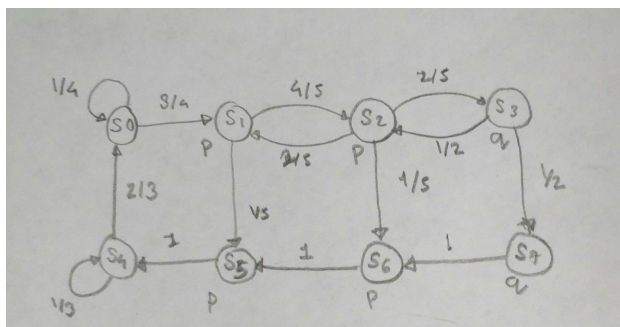
For this DTMC,

$P[Z, A] = \begin{bmatrix} 0 \\ (1-\alpha) \end{bmatrix}$

so we obtain  $n.P[Z, A] = [1/(1-\alpha)^2]$

Therefore the probability that message will be received at time  $t$  is given by  $1/(1-\alpha)^2$ .

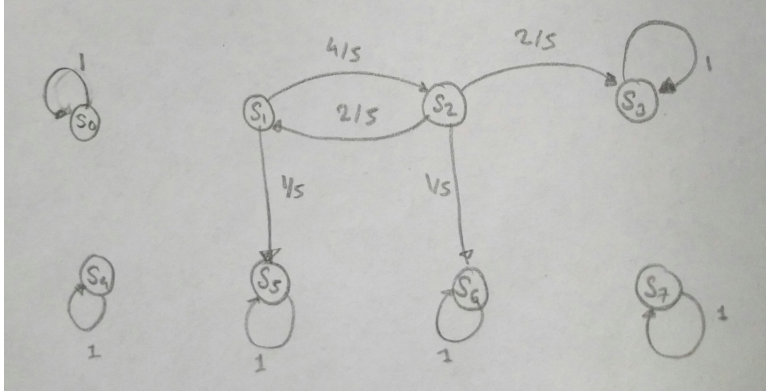
Question 3. We have,  $P_{>1/2} p U q$ .



States that satisfy  $q = \{s_3, s_7\}$

States that does not satisfy  $EpUq = \{s_0, s_4, s_5, s_6\}$

We make this states absorbing states. The modified DTMC is,



Transition Probability matrix is,

Transient States =  $\{s_1, s_2\}$

Absorbing States =  $\{s_0, s_3, s_4, s_5, s_6, s_7\}$

$$P = \begin{bmatrix} 0 & 4/5 & 0 & 0 & 0 & 1/5 & 0 & 0 \\ 2/5 & 0 & 0 & 2/5 & 0 & 0 & 1/5 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

Now we want to measure  $X_\infty[i] = \lim_{t \rightarrow \infty} P^t \cdot f$

The absorbing part is  $X_\infty[A] = f(a)$

$$f(a) = [0 \quad 1 \quad 0 \quad 0 \quad 0 \quad 1]$$

The transient part is  $X_\infty[Z] = N \cdot P[Z, A] \cdot f[A]$

$$\Rightarrow N^{-1} \cdot X_\infty[Z] = P[Z, A] \cdot f[A]$$

$$\Rightarrow (P[Z, Z] - I) \cdot X_\infty[Z] = -P[Z, A] \cdot f[A]$$

$$\text{Now, } P[Z, Z] - I = \begin{bmatrix} 0 & 4/5 \\ 2/5 & 0 \end{bmatrix} - \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Now,  $-P[Z, A].f[A]$

$$= \begin{bmatrix} 0 & 0 & 0 & 1/5 & 0 & 0 \\ 0 & 2/5 & 0 & 0 & 1/5 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 2/5 \end{bmatrix}$$

$$\text{Now, } (P[Z, Z] - I)^{-1} = (-25/17) \begin{bmatrix} 1 & 4/5 \\ 2/5 & 1 \end{bmatrix}$$

$$\text{Therefore } X_{\infty}[Z] = \begin{bmatrix} 8/17 \\ 10/17 \end{bmatrix}$$

$$X_{\infty} = \begin{bmatrix} 0 & 8/17 & 10/17 & 1 & 0 & 0 & 0 & 1 \end{bmatrix}$$

Therefore the following set of states satisfy the property,

$$h = \begin{bmatrix} 0 & 0 & 1 & 1 & 0 & 0 & 0 & 1 \end{bmatrix}$$

Therefore states  $s_2, s_3, s_7$  satisfy the property.

- Question 4. 1. From the definition of absorbing DTMC, every state is either transient or absorbing. Or in other words it is a DTMC in which every state can reach an absorbing state.  
An Absorbing state is such state that once entered, can not be left.

If X is an absorbing DTMC then m is an absorbing state. Which means  $\alpha = 1$ . Because we know that absorbing state loops into itself with probability 1.

If  $\alpha = 1$  then in DTMC  $X'$  the edge from m to m+1 will have probability 1.

And the state m+1 will loop into itself with probability  $\alpha = 1$ .

The outward edge from m+1 to m will have probability  $(1 - \alpha) = 0$ .

Which means once in state m, the system will always enter state m+1 which is an absorbing state and it can not left (m+1) as the probability at the outward edge is 0.

This is in line with the definition of absorbing chain. Therefore  $X'$  is an absorbing chain given X is an absorbing chain.

2.  $p_0$  is the initial probability distribution of absorbing DTMC X.  
 $p_0'$  is the initial probability distribution of absorbing DTMC  $X'$  where  $p_0' = [p_0, 0]$

n is the vector of expected number of visits to each state in X.

n' is the vector of expected number of visits to each state in  $X'$ .

In X m is an absorbing state. Let (m-1) be a transient state just before m. Let the number of visits of state (m-1) i.e  $n[m-1] = a$ . On reaching m from (m-1) it will be

absorbed and can not return.

In  $X'$ , number of visits of  $(m-1)$  until it reaches state  $m$  is also  $n[m-1]=a$ . Now if it returns from  $m$  to  $m-1$  then  $n[m-1]$  is not anymore.

But in  $X'$ ,  $(m+1)$  is an absorbing state, hence  $\alpha = 1$ . Now there is an edge from  $m$  to  $m+1$  with probability  $\alpha$ . If  $\alpha = 1$ , then  $m$  can not have any more outward edges. That means there will be no more outward edges from  $m$ . Therefore once it reaches  $m$  it can not return back to  $m-1$  and number of visits  $n[m-1]$  remains unchanged.

Therefore  $n[m-1]=n'[m-1]$  and in general,  $n[i]=n'[i]$  for all  $i \leq m$ .

If  $i$  is an absorbing state in  $X$  and  $X'$ , then  $n[i]=n'[i]$  is trivially true.

Therefore,  $n[i]=n'[i]$  for all  $i \leq m$ .

3.  $n_m$  is the vector of expected number of visits to each state in  $x_m$ .

Now,  $n[m+1] = \alpha / (1 - \alpha) \cdot n[m]$  from part 2.

Here our starting state  $m$  is 2. i.e expected number of visits to state 2 is  $n[2]$ .

Therefore  $n[3] = \alpha / (1 - \alpha) n[2]$

Similarly we have  $n[4] = \alpha / (1 - \alpha)^2 n[2]$  and

$n[m] = \alpha / (1 - \alpha)^{m-2} n[2]$

Therefore vector  $n_m$  is given by,

$[n[2], \alpha / (1 - \alpha) n[2], \alpha / (1 - \alpha)^2 n[2], \dots, \alpha / (1 - \alpha)^{m-2} n[2]]$

4.  $t_m$  is the mean time to absorption for  $x_m$ .

We have,  $t_m = \sum_{m=2}^m n[m]$

$= n[2] + n[3] + \dots + n[m]$

$= n[2] [1 + \alpha / (1 - \alpha) + \dots + \alpha / (1 - \alpha)^{m-2}]$

When  $n[m]$  = Expected number of visits to a state  $m$ .

From part 2 we know that  $n[m+1] = \alpha / (1 - \alpha) n[m]$ . Using this formula we get this expression of  $t_m$ .