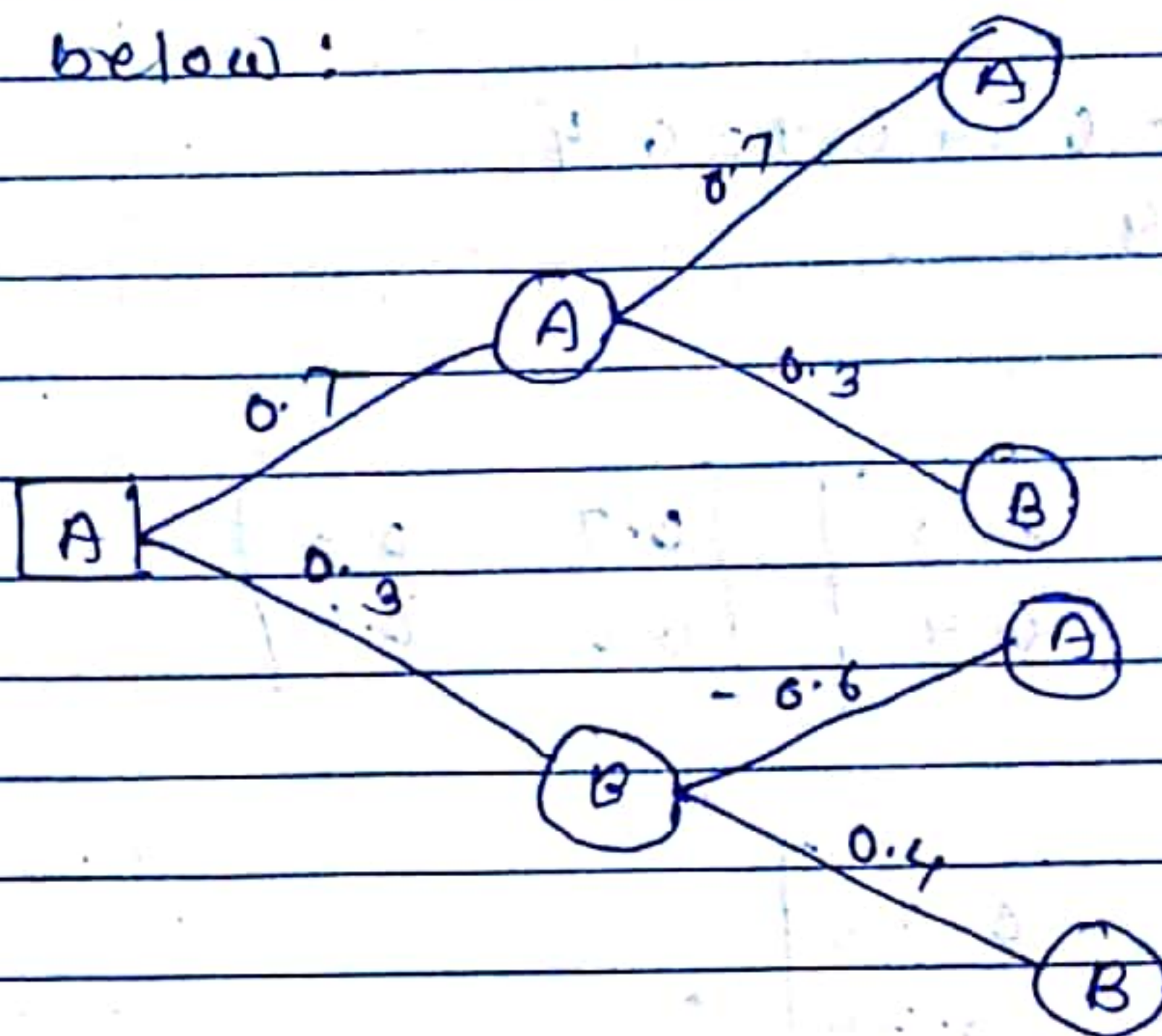


## Assignment - 2

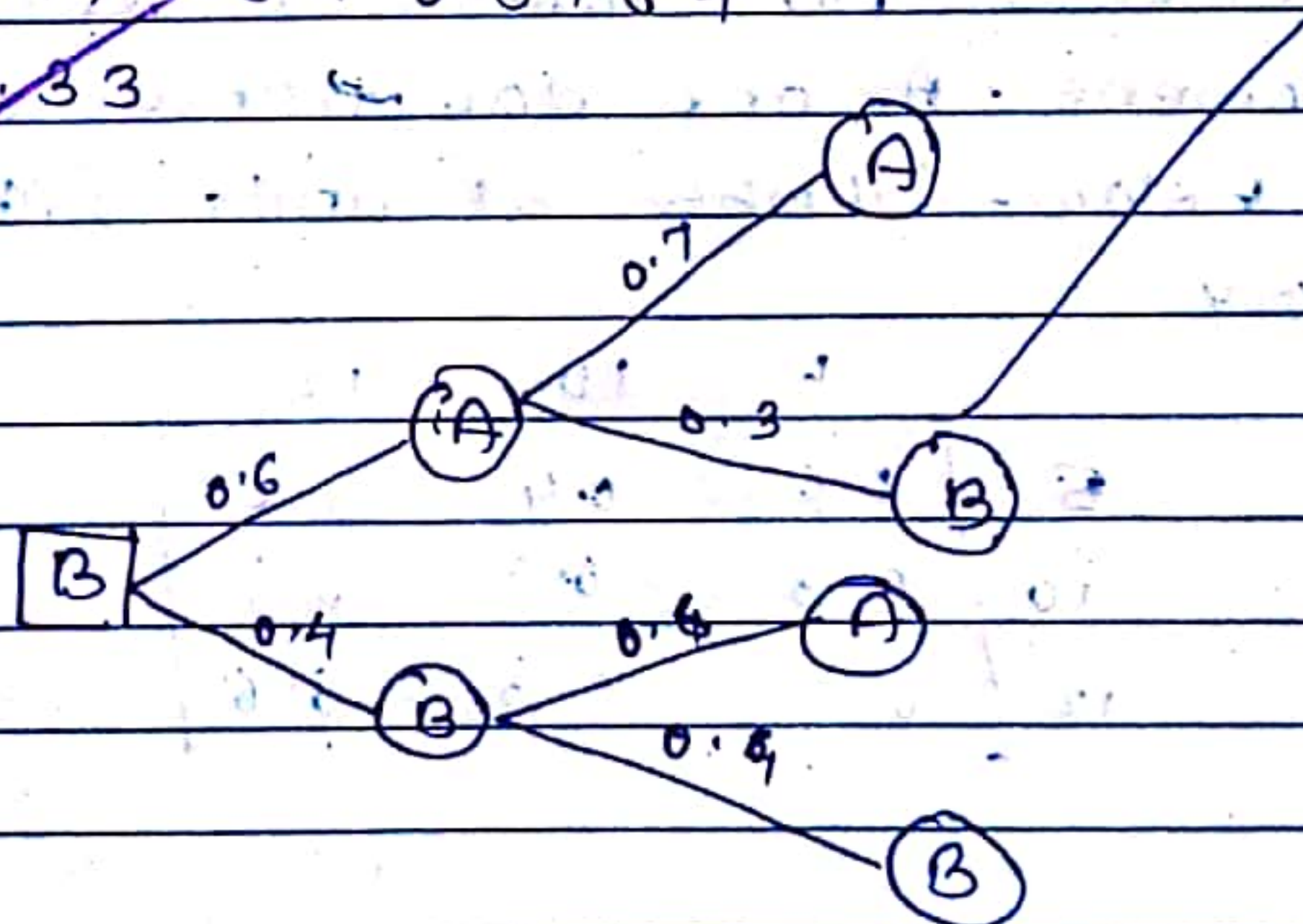
1) The transition of state can also be depicted in the form of probability tree which is shown below:



work out the state probabilities

$$P_{11}^{(2)} = 0.7 \times 0.7 + 0.3 \times 0.6$$
$$= 0.67$$

$$P_{12}^{(2)} = 0.7 \times 0.3 + 0.3 \times 0.4$$
$$= 0.33$$





$$P_{21}^{(2)} = 0.6 \times 0.7 + 0.4 \times 0.6 = 0.66$$

$$P_{22}^{(2)} = 0.6 \times 0.3 + 0.4 \times 0.4 = 0.34$$

$$P = \begin{bmatrix} 0.7 & 0.3 \\ 0.6 & 0.4 \end{bmatrix}$$

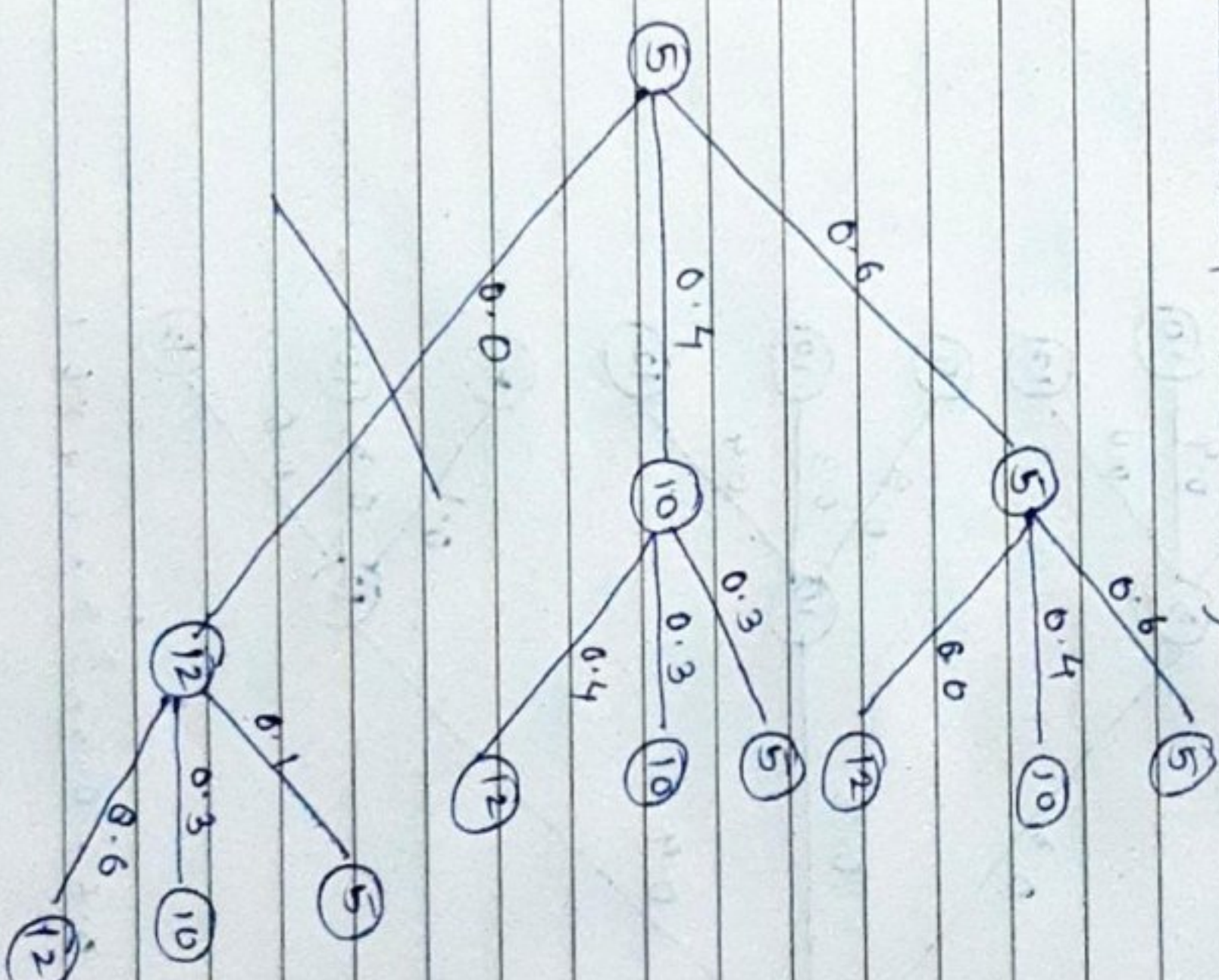
$$P = \begin{bmatrix} 0.67 & 0.33 \\ 0.66 & 0.34 \end{bmatrix}$$

2) The number of units of an item that are withdrawn from inventory on day-to-day basis follow a Markov chain process in which requirements for tomorrow depend on today's requirement. A one day transition matrix is given below - Number of units withdrawn from the inventory

	5	10	12
5	0.6	0.4	0.0
10	0.3	0.3	0.4
12	0.1	0.3	0.6

- Construct a tree diagram showing inventory requirements on two consecutive days
- Develop a two-day transition matrix
- Comments on how two-day matrix might be helpful to manager who is responsible for inventory management.

(a) probability tree diagram

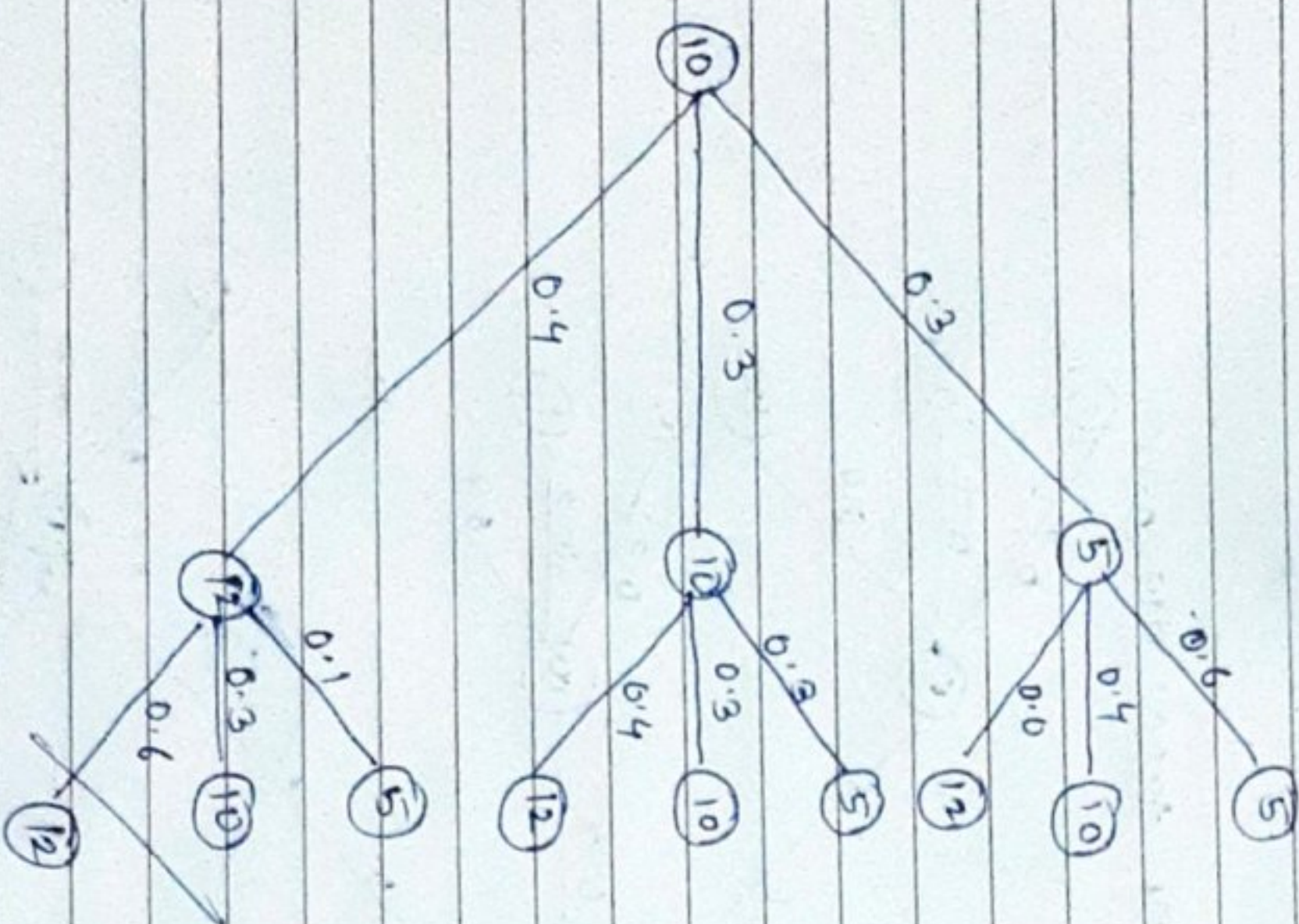


$$P_{11}^{(2)} = 0.6 \times 0.6 + 0.4 \times 0.3 + 0.0 \times 0.1 = 0.48$$



$$P_{12}^{(0)} = 0.6 \times 0.4 + 0.4 \times 0.3 + 0.0 \times 0.3 = 0.36$$

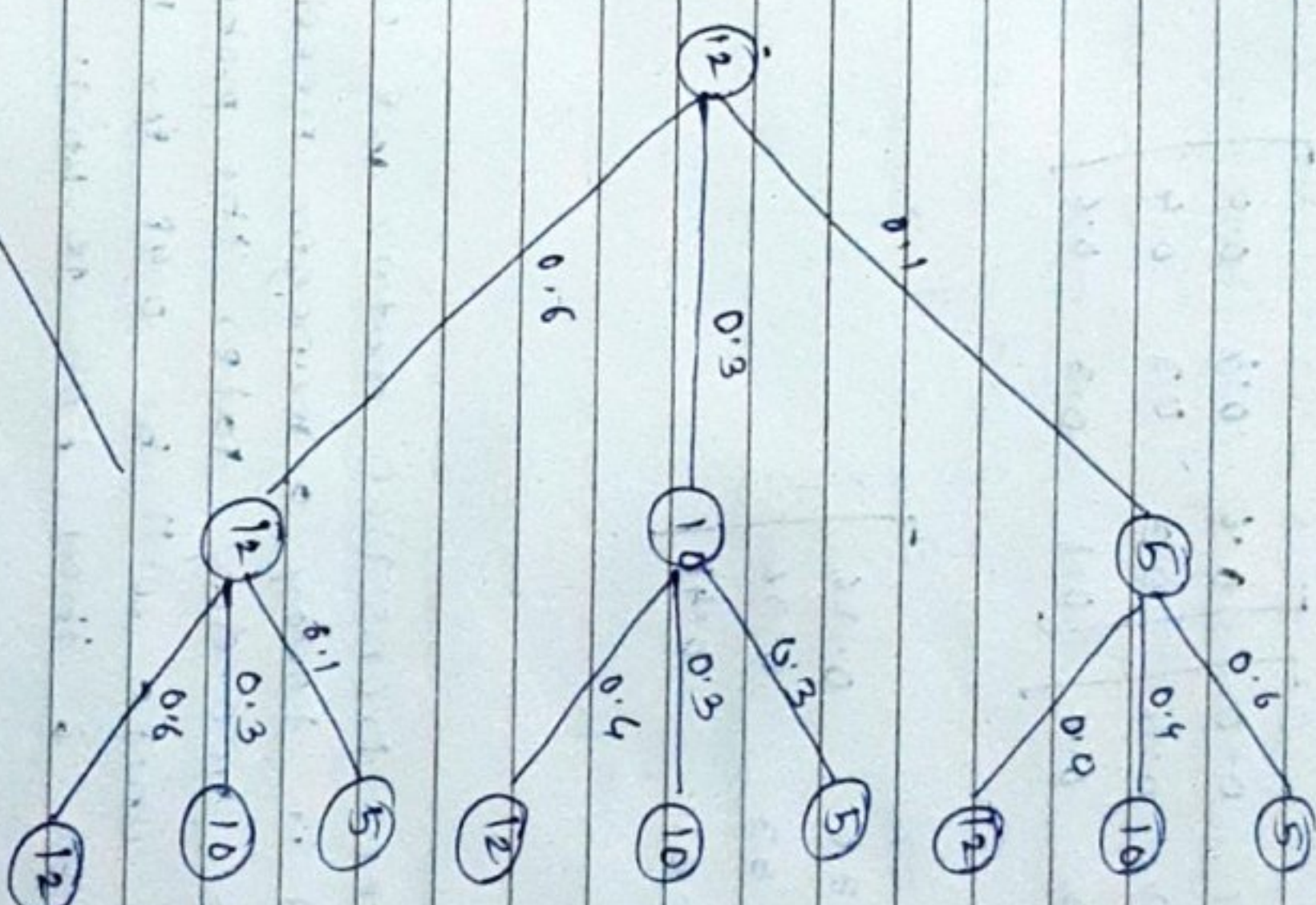
$$P_{13}^{(0)} = 0.6 \times 0.0 + 0.4 \times 0.4 + 0.0 \times 0.6 = 0.16$$



$$P_{21}^{(0)} = 0.3 \times 0.6 + 0.3 \times 0.3 + 0.4 \times 0.1 = 0.31$$

$$P_{22}^{(0)} = 0.3 \times 0.4 + 0.3 \times 0.3 + 0.4 \times 0.3 = 0.33$$

$$P_{12}^{(1)} = 0.3 \times 0.6 + 0.3 \times 0.4 + 0.4 \times 0.6 = 0.36$$



$$P_{13}^{(1)} = 0.1 \times 0.6 + 0.3 \times 0.3 + 0.6 \times 0.1 = 0.21$$

$$P_{21}^{(1)} = 0.1 \times 0.4 + 0.3 \times 0.3 + 0.6 \times 0.3 = 0.31$$

$$P_{22}^{(1)} = 0.1 \times 0.0 + 0.3 \times 0.4 + 0.6 \times 0.6 = 0.48$$



b) Develop a two day transition matrix

$$P^{(2)} = \begin{bmatrix} 0.6 & 0.4 & 0.0 \\ 0.3 & 0.5 & 0.4 \\ 0.1 & 0.3 & 0.6 \end{bmatrix} \begin{bmatrix} 0.6 & 0.4 & 0.0 \\ 0.3 & 0.3 & 0.4 \\ 0.1 & 0.3 & 0.6 \end{bmatrix}$$

$$P^{(2)} = \begin{bmatrix} 0.48 & -0.36 & 0.16 \\ 0.31 & 0.33 & 0.36 \\ 0.11 & 0.31 & 0.48 \end{bmatrix}$$

c)

From the given transition matrix we can conclude that today a manager needs 5 units then two day later the probability of needing 5 units will be 0.48 for 10 units it is 0.36 & then for 14 units it is 0.16

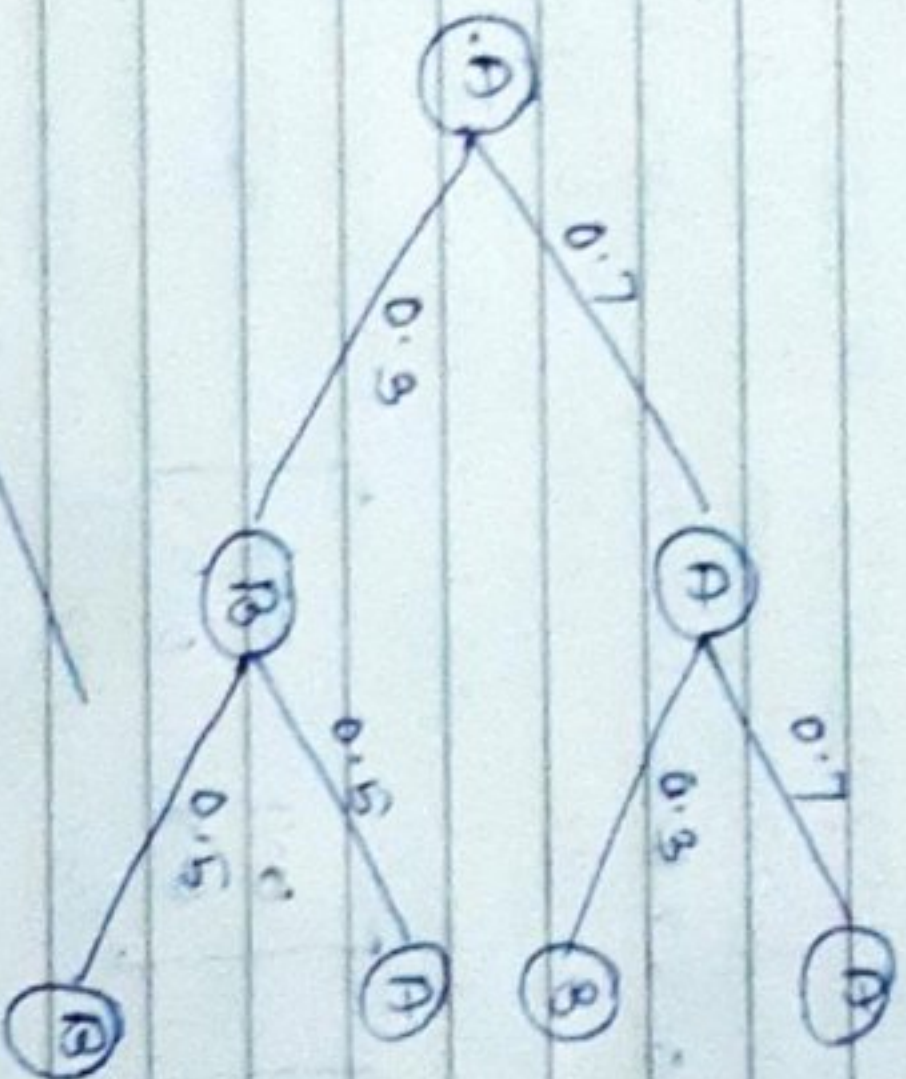
a) Two traders are in for tough competition and trader has been lucky to retain his customer 70% of the time and also is attracting a customer earlier using the product of his competitor trader to switch over two his product 30% of the time

Construct and interpret a transition matrix in terms of

- a) Retention & loss
- b) Retention & gain

The transition matrix can be arranged in following manner

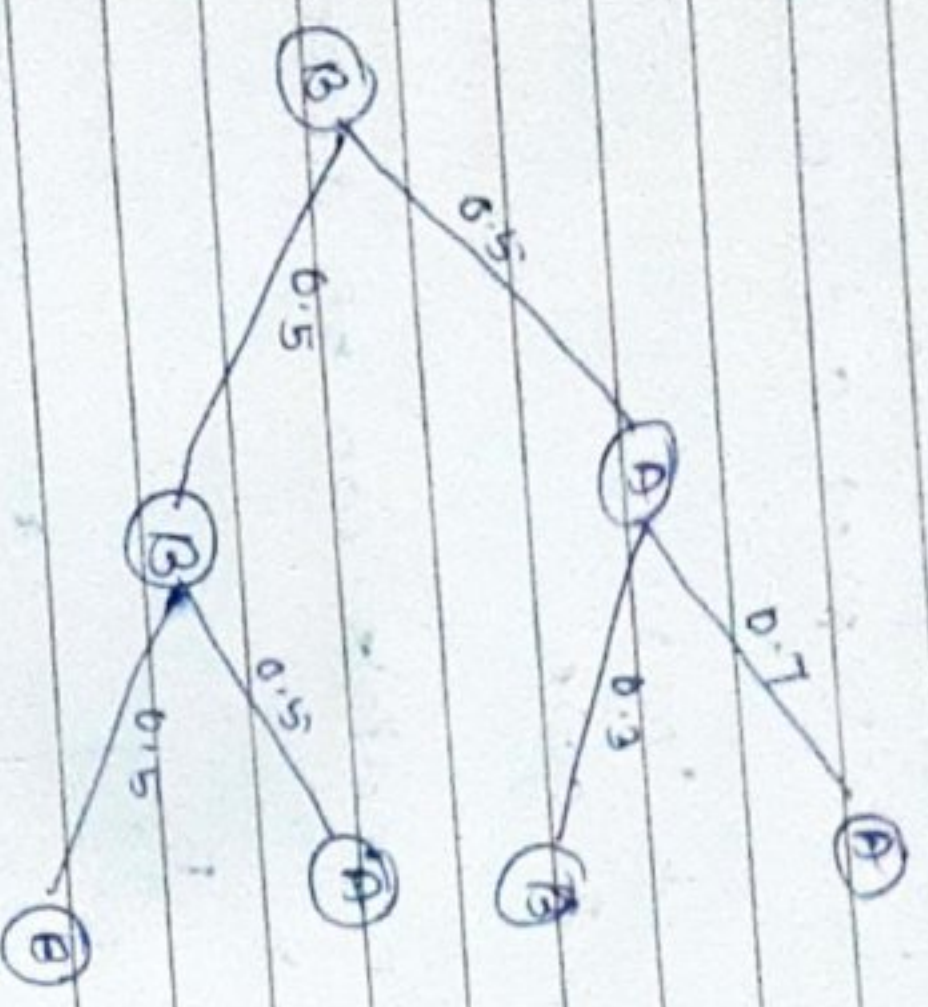
$$P = \begin{bmatrix} A & B \\ B & A \end{bmatrix} = \begin{bmatrix} 0.7 & 0.3 \\ 0.5 & 0.5 \end{bmatrix}$$



$$P_{11}^{(2)} = 0.7 \times 0.7 + 0.3 \times 0.5 = 0.64$$

$$P_{12}^{(2)} = 0.7 \times 0.3 + 0.3 \times 0.5 = 0.36$$





$$P_{2,1}^{(2)} = 0.5 \times 0.7 + 0.5 \times 0.5$$

$$= 0.6$$

$$P_{2,2}^{(2)} = 0.5 \times 0.3 + 0.5 \times 0.5$$

$$= 0.4$$

$$P = \begin{bmatrix} 0.7 & 0.3 \\ 0.5 & 0.5 \end{bmatrix} \quad \begin{bmatrix} 0.7 & 0.3 \\ 0.5 & 0.5 \end{bmatrix}$$

$$P = \begin{bmatrix} 0.64 & 0.36 \\ 0.6 & 0.4 \end{bmatrix}$$

return  
gain

Ap Markov chains named after the Russian mathematician Andrey Markov, are stochastic models used to describe a sequence of events where the probability of each event depends only on the state attained in the previous event. In other words, it's a process where future states depend only on the current state & not on the events that preceded it.

Markov chains can be applied successfully in various areas of management, including:

- 1) Operation Management: Markov chains can be used to model and analyze the flow of materials, products, or information through different stages of a production or service process. This can help identify bottlenecks, optimize resource allocation, and improve process efficiency.

- 2) Inventory Management: Markov chains can be used to model inventory systems where the state represents the inventory level. By analyzing transitions (probabilities) between states, businesses can better predict future inventory levels, optimize reorder points, and reduce holding costs.

- 3) Project Management: Markov chains can be used to model project workflows, where the

4) Thus, given understanding by Markov chain? In the areas of management can it applied successfully



states represent different stages or stages of completion. This can help project manager forecast project timelines, identify critical paths, and allocate resources effectively.

4) Queue management: Queue chains can be used to model and analyse various types of queues, such as financial market fluctuations, media default probabilities, or insurance claim patterns. By understanding queue probabilities, different nodes states, businesses can develop robust risk management strategies and minimize potential losses.

5) Let us take a case service station where customers arrive randomly. The arrival and service pattern have been simulated, along with the simulated pattern for arriving to help design the service facility.

Service time (min)	Frequency
0.5	15
1.0	31
1.5	26
2.0	19
2.5	7
3.0	5

Arrival pattern for customers

Time bet <sup>n</sup> arrival (min)	Frequency
0.5	2
1.0	6
1.5	10
2.0	25
2.5	20
3.0	14
3.5	10
4.0	7
4.5	4
5.0	2

Step 1: Table for arrival pattern

Arrival time (min)	Frequency	Cumulative frequency	Cumulative Probability	Random Number
0.5	2	2	0.02	00-01
1.0	6	8	0.08	02-07
1.5	10	18	0.18	08-17
2.0	25	43	0.43	18-42
2.5	20	63	0.63	43-62
3.0	14	77	0.77	63-76
3.5	10	87	0.87	77-86
4.0	7	94	0.94	87-93
4.5	4	98	0.98	94-97
5.0	2	100	1.00	98-99



Step 2-1 Table for simulation system service

Arrival Number	Pattern Number	Arrival Time	Time for Arrival bet
1	78	8.5	8.5
2	78	9.5	9.0
3	06	1.0	8.0
4	04	1.0	9.0
5	97	4.5	13.5
6	71	3.0	16.5
7	78	3.5	20.0
8	59	3.5	22.5
9	05	1.0	23.5
10	95	4.5	28.0

Service time	Frequency	Cumulative frequency	Cumulative Probability	Range
0.5	12	12	0.12	00-11
1.0	21	33	0.33	12-32
1.5	36	69	0.69	33-68
2.0	19	88	0.88	69-87
2.5	7	95	0.95	88-94
3.0	5	100	1.00	95-99

Service Number	Random Number	Service time
1	94	1.5
2	26	1.0
3	51	1.5
4	45	1.5
5	46	1.5
6	84	2.0
7	58	1.5
8	58	1.5
9	60	1.5
10	24	1.0
		14.5

= 14.5 min

Service time	Service start	Service finish
1.5	8.5	5.0
1.0	7.0	8.0
1.5	8.0	9.5
1.5	9.5	11.0
1.5	13.5	15.0
2.0	16.5	18.0
1.5	20.0	21.0
1.5	22.5	24.0
1.5	24.0	25.0
1.0	28.0	29.0



Total elapsed time = 29 min

Including arrival time = 10 min

Percentage of customer time =  $\frac{100}{29} = 3.44$

$\approx 3\%$

*[Signature]*