

Introduction to Probability Theory and Stochastic Processes
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Lecture - 86

There are many applications of queueing system we are going to discuss the abstract queueing system in the further lecture.

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Applications of Queueing Systems

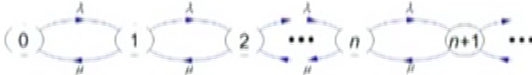
- Analyzing Network delays.
- Telephone conversations.
- Aircraft landing problems.
- Barber's shop




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M/M/1 Queueing Model

- Arrival process: Poisson Process with rate λ .
- Service times: Exponential with parameter μ
- Service times and inter-arrival times are independent
- Single server
- Infinite capacity in a system
- $N(t)$: Number of customers in a system at time t (state)



State transition diagram



The easiest or the simplest queueing model that is, a Markovian queueing model that is M/M/1 queueing model. Later, we are going to relate with the birth death process also. In the M/M/1 queueing model the inter arrival time is exponentially distributed. As I discuss the Poisson process in the previous lecture, whenever you have a arrival follows a Poisson process. Then the inter arrival time follows a exponential distribution and or independent also.

So, here the 1st information is a arrival process follows a Poisson process with the intensity or rate λ ; that means, a the inter arrival times are independent and each one is exponentially distributed with the parameter λ .

The 2nd information that is service time service times are exponentially distributed with the parameter μ . And the service times are independent for each customers and that is also independent with the arrival process; that means, there is no dependency over the arrival pattern with the service pattern, service times and the inter arrival times are independent. Then the 3rd information only 1 server in the system that is a queueing system in which only 1 server and the 4th information is missing; that means, it is a default it is infinite capacity model infinite capacity model.

Now, our interest is to find out the behavior of a queueing system or the behavior of number of customers in the system at any time t . Therefore, you can define a random variable n of t that is nothing but the number of customers in the system at time t . Therefore, this is going to follow form a stochastic process over the t .

Since, the inter arrival time is a exponentially distributed and the service times are exponentially distributed, the memory less property is going to be satisfied throughout all the time. Therefore, this Stochastic process there is a Discrete state continuous time stochastic process satisfying the Markov property. Therefore, this is a Markov process since, a inter arrival time is exponentially distributed and the service time is exponentially distributed and both are independent.

And the service time is also independent for each customers, therefore, this stochastic process satisfies the memory less property at all time points. Therefore, this discrete state because the possible values of n of t since, it is a number of customers the possible values are 0, 1, 2 and so on countably infinite. Therefore, it is a discrete state and you are observing the system over the time. Therefore, it is a continuous time therefore; this

stochastic process is a discrete state, continuous time, stochastic process satisfying the Markov property based on these assumptions.

Therefore, n of t is a Markov process since the state space is a discrete. Therefore, this is a Markov chain therefore, this is a continuous time Markov chain therefore, n of t is a CTMC. So, one can write the state transition diagram for this CTMC; that means, the possible states are 0, 1, 2 and so on. So, this will form a notes and you try to find out, what is the rate in which the system is moving from one state to other state.

Since it is a MM1queue model queueing model therefore, whenever the system is in the whenever the system is in the state 0 by the inter arrival time which is exponentially distributed the number of customers in the system will be incremented by 1. Therefore, that rate will be λ or the system moving from the state 0 to 1 it spends exponentially distributed amount of time here before moving into the state 1.

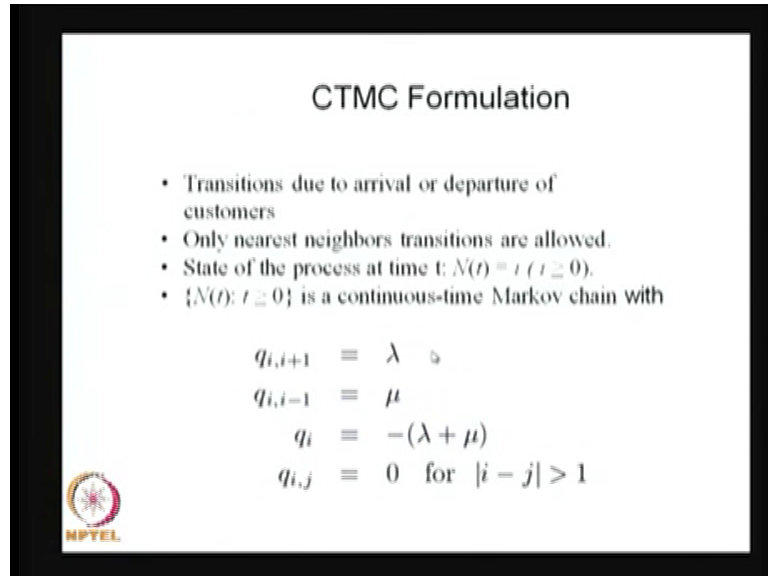
Once the system come to the state 1 either one more arrival is possible or the customer who is under service then service could have been finished. Therefore, the service time is a exponentially distributed with the parameter μ . Therefore, the system goes from the state 1 to 0 with the parameter μ . Similarly, from 1 to 2 because of the inter arrival time is exponentially distributed with the parameter λ therefore, this is λ .

Since, the arrival follows a Poisson process in a very small interval of time; only one customer is possible with the probability $\lambda \Delta t$ and so on. Therefore, there is no way the system goes from one state to jump into more than 1 state that is not possible forward. So, only one step forward is possible because of the arrival process follows a Poisson process and the since we have only one servers in the system the system also decremented by only one level below

Therefore, this is going to form a birth death process. The reason for this CTMC going to be a birth death process because of that arrival process follows a Poisson process. So, whatever the assumptions we have it for the Poisson process that is going to be satisfied. And since, we have only one server in the system and he does the service for only one customer at a time. After finishing that server after finishing the customer service, then, it move into the next service immediately and so on if the customers are available in the queue.

Therefore, the system goes to the 1 state below by only one move only it would not move from 2 to 0 or 3 to 1 and so on therefore, this CTMC is a birth death process.

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The slide is titled "CTMC Formulation" and contains the following content:

- Transitions due to arrival or departure of customers
- Only nearest neighbors transitions are allowed.
- State of the process at time t : $N(t) = i$ ($i \geq 0$).
- $\{N(t); t \geq 0\}$ is a continuous-time Markov chain with

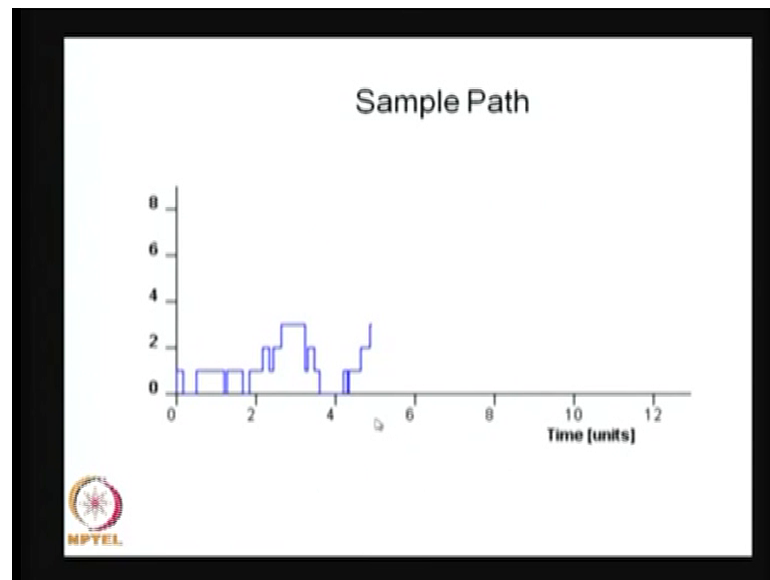
$$q_{i,i+1} = \lambda$$
$$q_{i,i-1} = \mu$$
$$q_i = -(\lambda + \mu)$$
$$q_{i,j} = 0 \quad \text{for } |i - j| > 1$$

In the bottom left corner of the slide, there is a logo for NPTEL (National Programme on Technology Enhanced Learning).

Therefore, I am connecting the CTMC with the MM1queue in particular the CTMC the birth death process. Because, of the transitions due to arrival or a departure of a customer and only nearest neighbor's transitions are allowed.

Because, of the assumptions which you have made therefore, this is going to a continuous time Markov chain with the rate in which the system moves from the state i to i plus 1 that, rate is λ . And the system moves from the state i to i minus 1 that rate is μ and all other rates are going to be 0 other than the diagonal element. And these rates also constraint not the state dependent rates. Therefore, this is a birth death process with the birth rates λ under death rates μ .

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So, this is a sample path suppose, at time 0 1 customer in the system then, services over then the 2nd customer enter into the system. Now, the number of customers in the system is 1 and so on so; that means, this duration is the service time for the first customer and from this point to this point, that is a inter arrival time of the 2nd customer entering to the system and from this time point to this time point, that is the service time for the second customer which is independent of the service time for the first customer.

And this is the time point 2nd customer enter and this is a time point in which the 3rd customer enters. Therefore, the inter arrival time is from this point to this point and so on. So, this is the dynamics of a number of customers in the system over the time. Therefore, this stochastic process is a Discrete State Continuous Time stochastic process at showing the Markov property. Therefore, this is a continuous time Markov chain.

So, later I am going to simulate the MM1 queueing model using some simulation technique.

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Stationary Distribution


$$\bar{\pi} = (\bar{\pi}_0, \bar{\pi}_1, \dots) ; \bar{\pi}_i \geq 0 ; \sum_i \bar{\pi}_i = 1$$

$$\bar{\pi} Q = 0 \quad \bar{\pi}_i = P[N=i]$$

$$0 = -\lambda \bar{\pi}_0 + \mu \bar{\pi}_1 \quad N = \sum_{k=0}^{\infty} N(k)$$

$$0 = \lambda \bar{\pi}_{i-1} - (\lambda + \mu) \bar{\pi}_i + \mu \bar{\pi}_{i+1}, \quad i \geq 1$$

$$\bar{\pi}_1 = \frac{\lambda}{\mu} \bar{\pi}_0$$

$$\bar{\pi}_{i-1} = \frac{\lambda}{\mu} \bar{\pi}_i = \frac{\lambda^{i-1}}{\mu^{i-1}} \bar{\pi}_0, \quad i = 1, 2, \dots$$


So, the conclusion is the under the underlying stochastic process for the MM1 queueing model is a birth death process the n of t is a stochastic process. So, this stochastic process is a birth death process therefore, now we are going to discuss the stationary distribution time dependent probabilities and so on.

So, how to find the stationary distribution? Solve πQ is equal to 0 π is the vector consists of a π_i is where π_i is are nothing but what is the probability that, n customers in the system. What is the probability that i customers in the system in a long run? So, that long run is defined in this way the n of t is a stochastic process as a t tends to infinity. The number of customers in the system in a long run that is going to be the n and π_i is nothing but a probability that n, i customers in the system in a longer run.

So now we are going to solve a πQ is equal to 0 with the normalized equation summation of π_i is equal to 1. So, once you frame the equation you will get a π_1 in terms of π_0 and the π_{i-1} in terms of first π_i then substitute recursively you will get in terms of π_0 .

So, since it is a homogeneous equation you will get all π_i is in terms of π_0 . So, use the normalizing equation summation of π_i is equal to 1.

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Take $\rho = \frac{\lambda}{\mu}$

Then,

$$\pi_0 = 1 - \rho$$
$$\pi_n = (1 - \rho) \rho^n \quad ; \quad \rho < 1 \text{ (stable system)}$$

$n = 1, 2, \dots$

ρ : offered load (traffic intensity)
 ρ : server utilization



You will get π_0 so the π_0 is equal to $1 - \rho$ where ρ is λ by μ . And since, I am relating this stochastic process with a birth death process with the infinite capacity. If you recall the stationary distribution exists as long as the denominator of a π_0 that series converges. So, that will converge only if λ by μ is less than 1 if λ by μ is greater than or equal to 1. Then, that denominator diverges accordingly you would not get the stationary distribution.

So, to have a stationary distribution you need a ρ has to be less than 1 that also you can intuitively say whenever the system is stable that is corresponding to ρ is less than 1. In that, you will have a stationary distribution; that means, in a longer run this is a proportion of the time the system will be empty and the π_n is nothing but the n customers in the system in a longer run that is $1 - \rho$ times ρ^n where ρ is less than 1.

This ρ can be visualized as the offered load also because the ρ is nothing but the mean arrival rate and the μ is the mean service rate and this ratio will give the offered load. And $1 - \pi_0$ that is the probability that the system is nonempty and that is nothing but the server utilization server utilization is nothing but what is the probability that the server is busy? The server will be busy as long as the system is not empty. So, the ρ is the server utilization that can be obtained in the from this formula and in a longer run the server utilization is ρ .