

Average Number in the System

$$\begin{aligned} E(N) &= \text{Average number of customers} \\ &\text{in the system in steady-state} \\ &= \sum_n n P^n (1-P) = P(1-P) \sum_{n=0}^{\infty} n P^{n-1} \\ &= P(1-P) \frac{1}{(1-P)^2} \end{aligned}$$

$$E(N) = \frac{P}{1-P}$$

$$\text{Also, } \text{var}(N) = \frac{P}{(1-P)^2}$$



Other than stationary distribution one can find out the average measures also in the system. So suppose you make a E of N that is nothing but the average number of customers in the system in steady-state since you know the probability distribution substitute P in's here therefore n times P in summation over n that is going to be the average number of customers in the system. If you do little simplification you will get ρ divided by $1 - \rho$ where ρ is less than 1. So this is the average number of customers in the system and also one can get variance of the number of customers in the system also for that you have to find out the E of N square and then using that formula you can get the variance of N also. So here we are getting a mean and variance of number of customers in the system in steady state.

Average Number in the Queue

$E(Q)$: Average number of customers in the queue in steady-state

$$= \sum_{n=1}^{\infty} n \pi_{n+1} = \sum_{n=1}^{\infty} n \rho^n (1-\rho)$$


$$= \rho^2 (1-\rho) \cdot \frac{1}{(1-\rho)^2}$$

$$= \frac{\rho^2}{(1-\rho)}$$



Also one can find average number in the queue. So the letter Q is a random variable and here we are finding the expectation of Q that is average number of customers in the queue. That means that before getting the service or how many customers in the system we have only one server in the system and whenever the service is going on and all other arriving customers will be queued. That means when n plus 1 customers in the system n people are in the queue. Therefore summation n times pi y n plus 1. Do the simplification you will get average number of customers in the queue also substitute pi n plus 1 from the one I have discussed in the stationary distribution.

Little's Law




John Little (1961)

In long run,

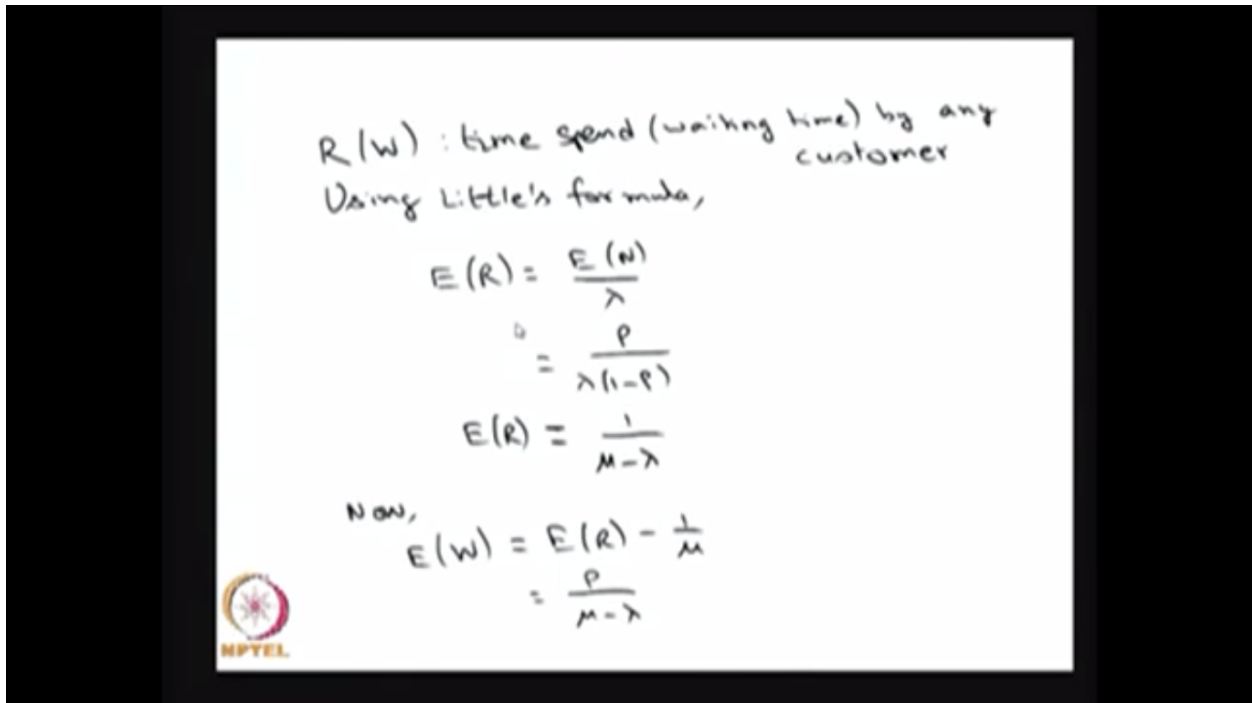
$$\lambda E(R) = E(N)$$

Here,
 λ : Average arrival rate
 $E(R)$: Average time spends in the system
 $E(N)$: Average number in the system

$$\lim_{t \rightarrow \infty} E(N(t)) = \lim_{t \rightarrow \infty} \lambda(t) \cdot \lim_{t \rightarrow \infty} E(R(t))$$



Here I'm going to relate the average measures using the Little's law. This is proven by John Little 1961. This is valid for any system in which arrival comes into the pattern with the arrival rate lambda and R is a time spent in the system and leave the system after the service or whatever the things are over then in a longer run one can say the arrival rate multiplied by the average time spent in the system that is same as average number in the system. So this relation is a valid for whatever be the underlying distribution of the service, underlying distribution of the arrival. What it says if you have a system in which a the arrival rate is a mean arrival rate is lambda and the mean time spent in the system is a expectation of R then that product will give average number in the system. Since directly it says whenever the system has a long run in a stable system the expectation of a average number of customers during the interval 0 to t as t tends to infinity that is going to be have a limit expectation of N and the arrival rate lambda of t that also has the mean arrival rate as t tends to infinity that is also going to be a sum having a limit constant lambda and similarly the average spent by the customers in the system at any time t and if you make a t tends to infinity that expectation quantity also has a limit therefore you will have a lambda times expectation of R is same as expectation of N. Now using this Little's law I am going to find out the the measures for the MM1 queue model.



So suppose R denotes the time spent in the system by the customer and W denotes the waiting time by any customer in the system. Then I can use the Little's formula, the Little's law in the previous one so if I know the mean arrival rate and if I know mean number of customers in the system in a longer run using these I can find out average time spent in the system. If I know to using Little's law if I know the average number of customers in the system longer run and if I know the arrival rate then I can find out the average time spent in the system in a long run.

Similarly I can, once I know the average time spend in the system, if I subtract the average time of my own service then that is going to be the average waiting in the queue. So this is average time waiting in the queue that is same as an average at time spent in the system minus my own average service time. The MM1 queue model the service time is exponentially distributed with the parameter μ therefore the average is $1/\mu$ so the difference will give the average time waiting in the queue by any customer. Not only the average measures for the MM1 queue one can find out the actual distribution for R as well as W also because this is a very simplest Markovian queuing model whereas for all other models it is little complicated but still one can get it. So this is the easy model in which one can find out the distribution of the time spent in the system as well as their time or as well as the waiting time by a customer in the queue. First let us go for finding out the distribution of waiting time.

Distribution of Waiting Time

$$W = \begin{cases} 0, & n=0 \\ S_1 + S_2 + \dots + S_n, & n=1, 2, 3, \dots \end{cases}$$

$$P[W \leq t] = \begin{cases} 0, & t < 0 \\ 1-p, & t = 0 \\ ?, & 0 < t < \infty \end{cases}$$

$$W/N=n \sim \text{Gamma}(n, \mu)$$

$$\text{For } t > 0 \quad P[W \leq t] = \sum_{n=1}^{\infty} \int_0^t \frac{\mu^n x^{n-1} e^{-\mu x}}{(n-1)!} dx (1-p)^n$$



Waiting time means if no one in the system when you arrive then your waiting time is zero. You are immediately going to get the service. So the service time is your time spent in the system. Usually the time spent in the system is the time of your service plus the time of the waiting time. So here I am finding the only the distribution of waiting time first. So whenever the system is zero your waiting time is zero. Whenever no customer in the system the waiting time is zero. Whenever more than or equal to one customer in the system then the waiting time is same as the remaining service time for the customer who is under service plus the customers in the queue before you join in the queue. So those people service time addition plus the residual or the remaining service time of the customer who the first customer who is under service. So this total time is the waiting time whenever the system is non-empty. Whenever the system is empty then the waiting time is 0.

Therefore the W's are random variable. Either it takes a value 0 or it takes a value greater than 0 based on the times time of a service of a previous N people in the ahead of you. Therefore W is a mixed random variable which has the probability mass function at 0 as well as a density function between the interval 0 to infinity.

So let me try for finding out the CDF of this random variable. So the CDF is going to be 0 till 0. At 0 it has the CDF of 1 minus Rho because and the waiting time is 0 that is equivalent of no one in the system. So in the long run no one in the system that is a pie naught and the pie naught probabilities 1 minus Rho. The system is the empty in a long run that is 1 minus Rho therefore the CDF at 0 that is same as 1 minus Rho of that is pie naught. Between the interval 0 to infinity one you have to find out the distribution of W.

Whenever N customers before you join in the system that conditional distribution, the distribution of W given the number of customers in the system is the N that distribution is

nothing but the service time of the N customers the first customer has the remaining service time the service time of the first customer is exponential distribution. The residual or remaining a service time of the first customer that is also exponential distribution because of memoryless property. So this is exponential distribution. This is second customer service times that is exponential distribution and similarly for the N th customer also the service time is exponential distributed and the way we made assumption all the service times are independent and each one is exponentially distributed with the parameter μ therefore this is a sum of N independent exponentially distributed random variables. Therefore the sum of N exponentials that is going to be a Gamma distribution with the parameters N and μ .

There are many ways of finding out the distribution but here I am just explaining through the distribution concept. This is sum of N independent exponential distribution therefore you can conclude it is Gamma distribution with the parameters N and μ . Once you know the conditional distribution our interest is to find out the unconditional one. That means for t is greater than 0 and CDF at the point a t that is nothing but what is the conditional density, probability density and what is the probability of N customers in the system that multiplication with the possible N will give the CDF between the interval 0 to t . So I have a density function of a Gamma distribution probability density function with the parameters N and μ and this is a probability density function multiplied and integration between 0 to t that will give the CDF and the unconditional multiplied by probability of N customers in the system that would be the summation that will give the unconditional. Therefore the CDF is going to be summation n is equal to 1 to infinity integration 0 to t of the probability density function of Gamma distribution multiplied by N customer in the system. If you do the simplification you will get a $1 - e^{-\rho t}$. Therefore you can substitute here. Here I made a mistake so here it is multiplied by ρ . So $1 - e^{-\rho t}$. So $1 - e^{-\rho t}$ times $e^{-\mu t}$ that is going to be the – so once you are getting the CDF you can conclude this is a mixed random variable with the probability mass at the 0 is $1 - \rho$ and the density function between the interval 0 to infinity that is a $\rho \mu e^{-\mu t}$; that is a probability density function for a distribution of waiting time.

Distribution of Response Time

$$R = S + S_1 + S_2 + \dots + S_n$$

$$P[R \leq t] = \begin{cases} 0 & , t \leq 0 \\ ? & , 0 < t < \infty \end{cases}$$

$$R/n \sim \text{Gamma}(n+1, \mu)$$

$$\begin{aligned} \text{For } t > 0 \\ P[R \leq t] &= \sum_{n=0}^{\infty} \int_0^t \frac{\mu^n x^{n-1} e^{-\mu x}}{n!} dx (1-\rho)^n \\ &= 1 - e^{-\mu(1-\rho)t} \end{aligned}$$



Similarly one can get the distribution of a response time also or the total time spent in the system. The total time spent in the system that is nothing but that's a random variable and the residual service time of the first customer who is in the system plus all the remaining N customers in the system in the queue plus your own service time. Therefore, here this is not a mixed random variable this is a continuous random variable because your service time is a continuous random variable which is exponentially distributed. Therefore the R is going to be sum of your own service plus the remaining service of the first person in the system if and so on till the n th customer who is in the queue. Therefore this is the CDF of the random variable R . Here also one can argue when N customer in the system before him who entered into the system that is a sum of exponential independent random variable and so on therefore this is going to be a Gamma distribution with the parameters n plus 1 μ and for t greater than 0 find out the CDF using the first conditional then unconditional multiplied by 1 minus ρ times ρ power n summation over n is equal to 0 to infinity because there is a possibility no one in the system or one customer, two customer so on therefore the running index is 0 to infinity. Do the simplification you will get to 1 minus e power minus ρ times this. Therefore, you can substitute here and if you see the CDF is the same as the CDF of exponential distribution with the parameter that is μ times 1 minus ρ . Therefore, you can conclude the total time spent in the system is exponentially distributed with the parameters μ times 1 minus ρ . If you find out the average time that is going to be 1 divided by the parameter that is this. The same thing you got it in the average response time and from the Little's formula using once you know the value of λ and expected number in the system. Using Little's law you got expectation of time spent in the system that is same result.

So here we are getting first finding the distribution of time spent in the system are response time. Then we are finding the average time.