Simulation & Analysis of M|G|1 Queuing Systems

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Project Report
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1 Introduction

In this project we simulate an M|M|1, $M|E_k|1$, and M|D|1 queuing systems and study the performance of each. We compare the mean queue length for all systems using the Pollaczek Khinchin formula.

2 Generating Random Variables

2.1 Uniform Random Variable

A uniform random variable distribution denoted by U[a,b] entails that for n values within the interval [a,b], each has a probability of $\frac{1}{n}$. The rand command in MATLAB generates random numbers between [0,1] according to a uniform random variable. In Figure 1 we plot P(X>x) for $x\in(0.5,1)$, where X is a uniformly distributed random variable.

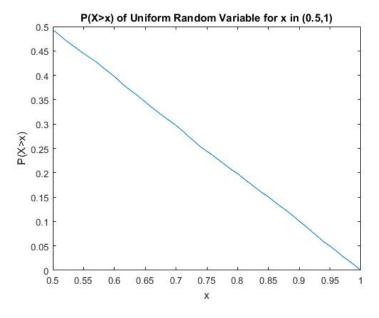


Figure 1: 1-CDF of Uniform Random Variable.

2.2 Exponential Random Variable

We can generate the exponential random variable X from its cumulative distribution function as follows:

$$F_X(x) = 1 - e^{-\lambda x} = P(U \ge e^{-\lambda x}) = P(\frac{-\ln(U)}{\lambda} \le x) = P(X \le x)$$

where $U \in [0,1]$ is a uniformly distributed random variable and $\lambda > 0$ is the rate. We can see that $X = \frac{-\ln(U)}{\lambda}$. The following MATLAB function generates exponential random variables:

Matlab Code

```
function e = exp_rv(n,avg)
% n : a vector of uniformly distributed random variables
% avg : the average of the desired exponential = 1/rate
% e : exponentially distributed random variables
e=-log(n)*avg;
end
```

In Figure 2 we plot P(X > x), where X is a exponentially distributed random variable with mean value of 2.

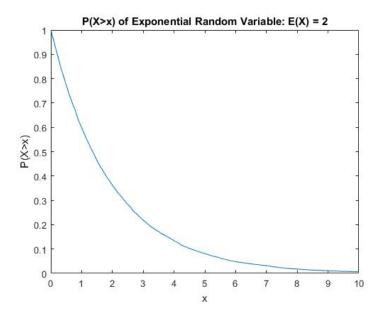


Figure 2: 1-CDF of Exponential Random Variable.

2.3 Poisson Random Variable

Similar to the steps followed when We generating an exponential random variable, we can generate a poisson random variable x according to the following equation:

$$U = e^{-\lambda} \sum_{k=0}^{\lfloor x \rfloor} \frac{\lambda^k}{k!}$$

where $U \in [0,1]$ is a uniformly distributed random variable and $\lambda > 0$ is the rate. The following MATLAB function generates an exponential random variable x:

Matlab Code

```
function p= pois_rv(n,avg)
% n : a vector of uniformly distributed random variables
% avg : the average of the desired poisson = rate
% p : poisson distributed random variables

p = zeros(length(n),1);
for i =1:length(n)
```

```
k=0;
9
       m=n(i);
10
        u=(avg.^k)/factorial(k);
11
        while m > \exp(-avg) * u
12
            m=m*n(i);
13
            k=k+1;
14
             u=u+(avg.^k)/factorial(k);
15
        end
16
        p(i) = k;
17
   end
18
   end
19
```

In Figure 3 we plot P(X > x), where X is a poisson distributed random variable with mean value of 2.

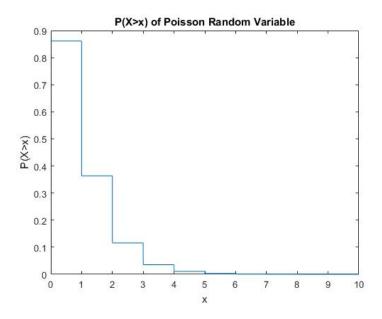


Figure 3: 1-CDF of Poisson Random Variable.

3 Simulation of an M|M|1 Queue

An M|M|1 has an infinite capacity queue and one server. The inter arrival time of packets to the queue follows a Poisson process with rate λ , and the service time for each packet follows an exponential distribution with rate μ . The following MATLAB code simulates an M|M|1 queue.

Matlab Code

```
% M|M|1 Queue
c = 120000;
                            %customers
 step size = 0.001;
                            %minutes
st = 80*50/step\_size;
                            % hour simulation time
lambda = 5;
                            % arrivals per minute
                            %departure per minute
u = 6;
rho = lambda/u;
% Generate arrival times, service and departure times according to M|M
% queue
a_{time} = zeros(c,1);
                         %real time of arrival for each packet
                         %service time for each packet
s time = zeros(c,1);
t \text{ time} = zeros(c,1);
                         %total time for each packet
```

```
14
   %Interarrival time is generated by poisson
15
   inter a time = \exp \operatorname{rv}(\operatorname{rand}(c,1),1/\operatorname{lambda});
16
   a \operatorname{time}(1) = \operatorname{inter} a \operatorname{time}(1);
   for i = 2:c
18
        a \text{ time}(i) = inter \ a \ time(i) + a \ time(i-1);
19
   end
20
21
   %Service time is generated by exponential
22
   s_{time} = \exp_{rv}(rand(c,1),1/u);
23
24
   %leaving time for each packet in system
   t \quad time(1)=a \quad time(1)+s \quad time(1);
26
   for i=2:c
27
        L = a\_time(i)+s\_time(i);
28
        K = t \quad time(i-1)+s \quad time(i);
29
        if L>K
30
             t time(i) = L;
31
        else
             t_{time}(i) = K;
33
        end
34
   end
35
36
   %simulate arrival and departure
37
   a \log = \operatorname{zeros}(\operatorname{st}, 1);
                                % 1 if arrival happens, 0 o.w , for debugging
38
   d \log = zeros(st,1);
                                % 1 if departure happens, 0 o.w , for debugging
39
   n \log = zeros(st,1);
                                % #packets in system at each simulation step
                                % current packets in system
   cps = 0;
41
   tpa = 0;
                                % total packets arrived
42
   tpl = 0;
                                % total packets that left
43
   for i = 1:st
        n \log(i) = cps;
45
46
        % Customer arrival
47
        if \ tpa < c
             if step size*i >= a time(tpa+1)
49
                  a \log(i) = 1;
50
                  cps = cps+1;
51
                  tpa = tpa+1;
52
             end
53
        end
54
        %Customer departure
56
        if tpl < c
57
        if step size*i >= t time(tpl+1) && cps>0
58
             d \log(i) = 1;
             cps = cps - 1;
60
             tpl = tpl+1;
61
        end
62
        end
   end
64
65
   v = 250;
66
   pn = zeros(v,1); %probability of number of packets
   pn(i) = size(find(n log=i))/size(n log);
69
   end
70
```

```
72
73 % average number
74 En=0; %average number of customers
75 for i=1:v
76 En=En+pn(i)*i;
77 end
```

When Simulating this program, the average number of packets in the queue E[n] = 4.939 at any time, and the average delay E[d] = 0.9878 minutes, for arrival rate $\lambda = 5$ packets per minute and departure rate $\mu = 6$ packets per minute.

In Figure 4 we plot the probability distribution of the number of packets in the queue P_n and the number of packets n.

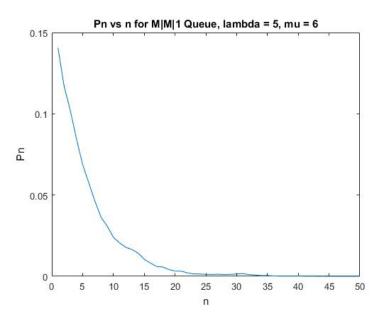


Figure 4: Pn vs n

4 Simulation of an $M|E_k|1$ Queue

The $M|E_k|1$ is different from the M|M|1 queue in that the service time for each packet follows an Erlang distribution, where

$$Erlang := Y_k = \sum_{i=1}^k X_i$$
$$X_i \sim exp(k\lambda)$$

The inter arrival time of packets to the queue also follows a Poisson process with rate λ . The MATLAB code to simulate an $M|E_k|1$ queue is the same as the previous code for the M|M|1 case except for the service time distribution. The following edit is made for k=4:

Matlab Code

```
^{1} %Service time is generated by an erlang random variable ^{2} k=4; ^{3} for i=1:k ^{4} s_time =s_time + exp_rv(rand(c,1),1/(k*u)); ^{5} end
```

When Simulating this program, the average number of packets in the queue E[n] = 2.97 at any time, and the average delay E[d] = 0.59 minutes, for arrival rate $\lambda = 5$ packets per minute,

departure rate $\mu = 6$ packets per minute, and k=4.

In Figure 5 we plot the probability distribution of the number of packets in the queue P_n and the number of packets n.

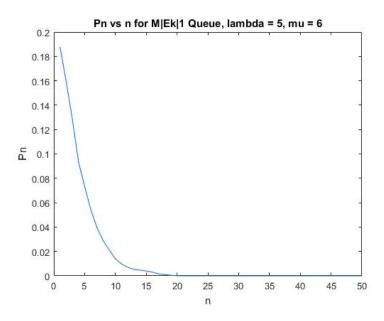


Figure 5: Pn vs n

When comparing the average number of packets E[n] in the queues, it's clear that this value is greater for the M|M|1 queue. To justify this we use the Pollaczek Khinchin formula for the mean queue length which states:

$$E[n] = \frac{\rho}{1 - \rho} [1 - \frac{\rho}{2} (1 - \mu^2 \sigma^2)]$$

where $\sigma^2=\frac{1}{\mu^2}$ for the M|M|1 queue and $\sigma^2=\frac{1}{k\mu^2}$ for the M|E_k|1 queue. Therefore the expected queue length for both queues becomes:

$$M|M|1:E[n] = \frac{\rho}{1-\rho}$$

$$M|E_k|1:E[n] = \frac{\rho}{1-\rho}[1-\frac{\rho}{2}(1-\frac{1}{k})]$$

Since $\rho \leq 1$ and k > 1, from the above equations, we can see that the expected queue length for the $M|E_k|1$ queue is less than that of the M|M|1 queue.

4.1 Comparison of an $M|E_k|1$ and an M|D|1 queue

In theory we learnt that an M|D|1 system provides a lower bound on congestion for a system with Poisson arrivals (lecture notes, Shroff). When calculating the expected queue length for an M| E_k |1 system it was noticed that as k increases the expected queue length decreases until it reaches a lower bound that is similar to that of an M|D|1 system.

To see this, we can compare the mean queue length of both systems again using the Pollaczek Khinchin formula. For a deterministic service time the variance $\sigma^2 = 0$. The mean queue length for both is:

$$M|E_k|1:E[n] = \frac{\rho}{1-\rho} [1 - \frac{\rho}{2} (1 - \frac{1}{k})]$$

 $M|D|1:E[n] = \frac{\rho}{1-\rho} [1 - \frac{\rho}{2}]$

We notice that as $k \to \infty$, the mean queue length of both becomes equal.

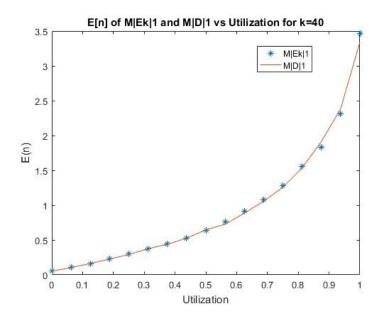


Figure 6: Expected number for $\mathcal{M}|E_k|1$ and $\mathcal{M}|\mathcal{D}|1$ system with k=40

Above in Figure 6 we simulate the expected queue length for different values for the utilization ρ for an M|E_k|1 and an M|D|1 system, with k=40 and show that the previous argument holds.