

THE COST FOR SINGLE CHANNLE WAITING LINE MODELS

ABSTRACT

Queuing theory is basically a mathematical approach applied to the analysis of waiting lines. The queuing models are very powerful tool for determining that how to manage a queuing system in the most effective manner. The aim of this study is to compare between three single channel waiting line models. The cost for the three single channel waiting line models is calculated with different distributions. The study results that first the cost which calculated from system which depend on the cost of queue is less for (M/D/1) model than the other two models when the same data are used. Second the results showed that the system cost and queue cost which calculated from (M/G/1) model when the service rate (μ) is followed weibull distribution is less the system cost and queue cost which calculated from exponential and gamma distributions which used.

Keywords: Waiting line cost; Poisson distribution; gamma distribution; weibull distribution; exponential distribution; Single Channel models;

1- Introduction

Queuing theory had its beginning in the research work of a Danish engineer named Anger Krarup Erlang In 1909; Erlang's experimented with fluctuating demand in telephone traffic. At the end of World War II, Erlang's early work was extended to more general problems and to business applications of waiting lines [1]. Queuing theory is basically a mathematical approach applied to the analysis of waiting lines. The queuing model are very powerful tool for determining that how to manage a queuing system in the most effective manner [2]. Queues or waiting lines are very common in everyday life whereby certain business situations require customers to wait in line for a service [3].

Uses models to represent the various types of queuing systems. Formula for each model indicates how the related queuing system should perform, under a variety of conditions. The queuing theory is also known as the random system theory, which studies the content of: the behavior problems, the optimization problem and the statistical inference of queuing system [4].

Applications of the queuing theory such as traffic flow (vehicles, aircraft, people, communications, transportation networks), scheduling (patients in hospitals, jobs on machines, programs on computer), facility design (banks, post offices, supermarkets, manufacturing) [5]. Most banks used queuing models. It is very useful to avoid standing in a queue for a long time to give tickets to all customers. Queuing is used to generate a sequence of customers' arrival time and to choose randomly between three different services: open an account, transaction, and balance, with different period of time for each service. [6]

Mehri *et al* [7] introduced the basic concepts of queuing models and showed how linear programming can be used to estimate the performance measures of a system. They studied Tunisian transport, and found widespread use in the analysis of service facilities, production and many other situations where congestion or competition for scarce resources may occur.

Edith *et al* [8] Regression analysis was employed to model the banks' queue system. They found that The Coefficient of determination, R^2 value was close to unity for multiple linear regression and unity for non-linear regression. Also, the Degree of Correlation obtained was found to be 92% and 100% for the multiple linear regression and non-linear regression.

Dhar & Rahman [9] used queuing model to derive the arrival rate, service rate, utilization rate, waiting time in the queue and the average number of customers in the queue. Queuing can help bank ATM to increase its quality of service, by anticipating, if there are many customers in the queue. In ATM, bank customers arrive randomly and the service time.

Muruganantha and Usha [10] calculated average queue length, average number of customer in the system. Average customer waiting time and average number of customer time spent in the queue in kanyakumari district at various places are introduced.

Santhi and Saravanan [11] discussed about several queuing model for cloud computing. These models are used to reduce waiting time of customer (calls) and increase performance of the system. Furthermore, they presented comparison of several queuing models results which are used for cloud computing environment.

The organization of the study is as follows: In Section 2 the study described Identifying Models Using Kendall Notation. Section 3 described the Waiting Line costs. Simulation study discusses in Section 4. Finally, discussion concluding remarks are provided in Section 5.

2-Identifying Models Using Kendall Notation:

David. G. Kendall (1953) developed a notation that has been widely accepted for specifying the pattern of arrivals, the service time distribution, and the number of channels in a queuing model. There are three symbols Kendall notation as follows

Arrival distribution / Service time distribution / Number of service channels open.

The following letters are commonly used in Kendall notation:

G = general distribution with mean and variance known,

D = constant (deterministic) rate, and

M = Poisson distribution for number of occurrences (or exponential times), [4].

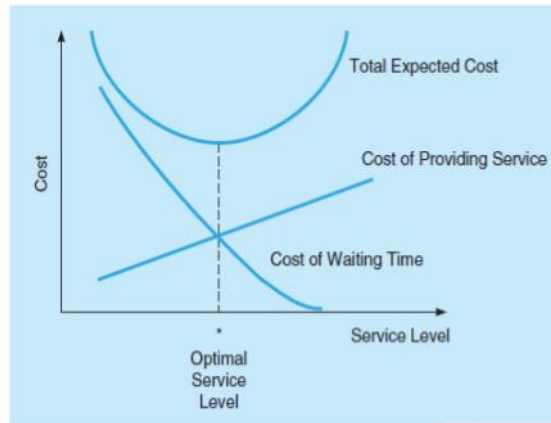
Single-Channel($M/M/1$), Constant – Service Time Model($M/D/1$), General Service queuing model($M/G/1$).the previous models are the three models which will used in this study.

3- Waiting Line costs

One of the goals of queuing analysis is finding the best level of service for an organization. Its objective is usually to find the medium between two extremes. On the other hand, a firm can retain a

large staff and provide many service facilities. This, can become expensive .The other extreme is to have the minimum possible number of checkout lines, such as gas pumps, or teller windows open. This keeps the service cost down but may result in customer dissatisfaction. As the average length of the queue increases and poor service results, customers and goodwill may be lost. Managers must deal with the trade-off between the cost of providing good service and the cost of customer waiting time. One means of evaluating a service facility is thus to look at a total expected cost; this the sum of expected service costs plus expected waiting costs. As service improves in speed, however, the cost of time spent waiting in lines decreases. This waiting cost may reflect lost productivity of workers while their tools or machines are awaiting repairs or may simply be an estimate of the costs of customers lost because the poor service and long queues. The objective is to minimize total expected costs. by minimize the sum of service costs and waiting costs. [7].

Figure (1) queuing cost and service levels



$$\text{Total expected service cost} = (\text{Number of channels}) (\text{Cost per channel}) = m C_s \quad (1)$$

Where

m = number of channels

C_s = service cost (labor cost) of each channel

The waiting cost when the waiting time cost is based on time in the system is

$$\begin{aligned} \text{Total expected waiting cost} &= (\text{Total time spent waiting by all arrivals}) (\text{Cost of waiting}) \\ &= (\text{Number of arrivals}) (\text{Average wait per arrival}) C_w \end{aligned}$$

So,

$$\text{Total expected waiting cost} = (\lambda W) C_w \quad (2)$$

If the waiting time cost is based on time in the queue, this becomes

$$\text{Total expected waiting cost} = (\lambda W_q) C_w \quad (3)$$

These costs are based on whatever time units (often hours) are used in determining λ . Adding the total service cost to the total waiting cost, have the total cost of the queuing system.

When the waiting cost is based on the time in the system, this is

Total expected cost = Total expected service cost + Total expected waiting cost

$$\text{Total expected cost} = m C_s + \lambda W C_w \quad (4)$$

When the waiting cost is based on time in the queue, the total cost is

$$\text{Total expected cost} = m C_s + \lambda W_q C_w . \quad (5)$$

4- Simulation Study

This section discusses the numerical simulation which used to evaluate the performance of the three waiting lines models; Single - Channel ($M/M/1$), Constant – Service Time model ($M/D/1$), and general - service queuing model ($M/G/1$). The study evaluate the performance for three waiting line single channel models when the cost for each model is calculated when different distributions are used.

*The first model Single - Channel ($M/M/1$) the arrival rate λ followed Poisson distribution and the service rate μ is followed exponential distribution.

*The second model ($M/D/1$) the arrival rate λ is followed the Poisson distribution and the service rate μ is followed exponential distribution and constant service rate model.

*The third model ($M/G/1$) the arrival rate λ is followed the Poisson distribution and the service rate μ is follow exponential, Gamma and Weibull distributions. The study suggested three distributions to evaluate the performance for three waiting lines models.

The numerical simulation study takes the following steps:

- 1- The study depends on the data which generated by Arinze *et al* [12] from NNPC mega petroleum station Owerri and NNPC mega petroleum station Enugu.
- 2- The study solved the three models for $n=52$ and showed the results in the following paragraph.
- 3- The first model Single - Channel ($M/M/1$) applied when the arrival rate λ followed Poisson distribution and the service rate μ is followed exponential distribution with different parameters.
- 4- The second model ($M/D/1$) applied when the arrival rate λ is followed the Poisson distribution and when the service rate μ is followed two constant service rate model. This model applied first when μ service rate used as in Arinze *et al* [12]. Second the model ($M/D/1$) applied when μ was a following constant (deterministic) value.
- 5- The third model ($M/G/1$) applied when the arrival rate λ is followed Poisson distribution and the service rate μ is followed any distribution. The study suggested three distributions to calculate the cost with different distribution.
- 6- The study generated $N=350$ is followed Gamma distribution and weibull distribution by Minitab program to choose the sample size $n=52$ which selected. Goodness of fit is used by easy fit program to be sure that the data which selected follow Gamma distribution and weibull distribution. The package program “QM for windows V5” is used to solve the three models under consideration.
- 7- To study the effect of distribution for each model the study suggests that the server cost = 4 and waiting cost = 2 as a constant for all cases:

Table (1) The three models with $\lambda \sim \text{Poisson}$, $\mu \sim \text{exponential}$.

Day	λ	μ	M/M/1		M/D/1		M/G/1	
			Waiting cost	system cost	Waiting cost	system cost	Waiting cost	system cost
1	29	30	60.07	62.00	32.03	33.97	54.74	56.67
2	30	31	62.06	64.00	33.03	34.97	58.14	60.08
3	31	32	64.06	66.00	34.03	35.97	61.71	63.65
4	32	33	66.06	68.00	35.03	36.97	65.44	67.38
5	33	34	68.06	70.00	36.03	37.97	69.35	71.29
6	34	35	70.06	72.00	37.03	38.97	73.44	75.39
7	35	36	72.06	74.00	38.03	39.97	77.72	79.66
8	36	37	74.05	76.00	39.03	40.97	82.18	84.13
9	37	38	76.05	78.00	40.03	41.97	75.19	77.14
10	38	39	78.05	80.00	41.03	42.97	91.71	93.66
11	39	40	80.05	82.00	42.03	43.98	96.78	98.73
12	40	41	82.05	84.00	43.02	44.98	69.26	71.22
13	29	31	31.13	33.00	17.56	19.44	29.3	31.17
14	30	32	32.13	34.00	18.06	19.94	31.02	32.9
15	31	33	33.12	35.00	18.56	20.44	32.83	34.71
16	33	35	35.11	37.00	19.56	21.44	36.71	38.59
17	36	38	38.11	40.00	21.05	22.95	43.21	45.11
18	37	39	39.10	41.00	21.55	23.45	39.6	41.49
19	38	40	40.10	42.00	22.05	23.95	48.04	49.94
20	39	41	41.10	43.00	22.55	24.45	35.02	36.92
21	40	42	42.10	44.00	23.05	24.95	44.05	45.95
22	28	31	20.86	22.67	12.43	14.24	19.72	21.53
23	29	32	21.52	23.33	12.76	14.57	20.83	22.65
24	30	33	22.18	24.00	13.09	14.91	22	23.82
25	32	35	23.50	25.33	13.75	15.58	24.5	26.33
26	36	39	26.15	28.00	15.08	16.92	30.24	32.09
27	38	41	27.48	29.33	15.74	17.59	23.63	25.49
28	40	43	28.81	30.67	16.4	18.26	25.58	27.44

29	41	44	29.47	31.33	16.73	18.6	26.6	28.46
30	29	33	16.74	18.50	10.37	12.13	16.62	18.37
31	37	41	20.70	22.50	12.35	14.15	17.96	19.77
32	38	42	21.19	23.00	12.6	14.4	26.24	28.05
33	39	43	21.69	23.50	12.84	14.66	19.38	21.2
34	40	44	22.18	24.00	13.09	14.91	20.13	21.95
35	41	45	22.68	24.50	13.34	15.16	20.9	22.73
36	29	34	13.89	15.60	8.95	10.65	14.09	15.8
37	35	40	16.25	18.00	10.13	11.88	18.95	20.7
38	36	41	16.64	18.40	10.32	12.08	14.57	16.33
39	37	42	17.04	18.80	10.52	12.28	17.71	19.47
40	38	43	17.43	19.20	10.72	12.48	15.68	17.45
41	39	44	17.83	19.60	10.91	12.69	16.27	18.04
42	35	41	13.96	15.67	8.98	10.69	12.33	14.04
43	36	42	14.29	16.00	9.14	10.86	14.81	16.53
44	37	43	14.61	16.33	9.31	11.03	13.23	14.95
45	40	46	15.59	17.33	9.8	11.54	14.7	16.44
46	29	36	10.67	12.29	7.34	8.95	11.23	12.84
47	39	46	13.45	15.14	8.72	10.42	12.72	14.42
48	33	42	9.76	11.33	6.88	8.45	11.45	13.03
49	36	45	10.40	12.00	7.2	8.8	9.79	11.39
50	38	47	10.83	12.44	7.41	9.03	10.43	12.05
51	35	45	9.44	11.00	6.72	8.28	8.93	10.48
52	37	47	9.83	11.40	6.91	8.49	9.49	11.06

The table (1) showed the results for the three Single- Channel models ($M/M/1$) , ($M/D/1$) and ($M/G/1$) when arrival rate (λ) is followed Poisson distribution and service rate (μ) is followed exponential distribution. For the sample size $n=52$ which are used with different values for, service rate (μ), arrival rate (λ) the arrival rate (λ) increases when the service rate (μ) increase and Average number of customers in the system (L) depend on the Average number of customers in the queue (L_q). Cost for system depends on cost for queue. Cost for system and cost for queue was decreased when different between arrival rate (λ) and service rate (μ) increased. The cost which calculated from

system which depend on the cost of queue is less for (M/D/1) model than the other two models when the same data are used.

Table (2) the (M/D/1) model with $\lambda \sim \text{Poisson}$, $\mu \sim \text{exponential}$ and constant.

Day	M/D/1				M/D/1			
	λ	μ	Waiting cost	system cost	λ	μ	Waiting cost	system cost
1	29	30	32.03	33.97	29	30	32.03	33.97
2	30	31	33.03	34.97	30	31	33.03	34.97
3	31	32	34.03	35.97	31	32	34.03	35.97
4	32	33	35.03	36.97	33	34	36.03	37.97
5	33	34	36.03	37.97	35	36	38.03	39.97
6	34	35	37.03	38.97	39	40	42.03	43.98
7	35	36	38.03	39.97	44	45	47.02	48.98
8	36	37	39.03	40.97	28	30	17.07	18.93
9	37	38	40.03	41.97	29	31	17.56	19.44
10	38	39	41.03	42.97	30	32	18.06	19.94
11	39	40	42.03	43.98	32	34	19.06	20.94
12	40	41	43.02	44.98	34	36	20.06	21.94
13	29	31	17.56	19.44	38	40	22.05	23.95
14	30	32	18.06	19.94	43	45	24.54	26.46
15	31	33	18.56	20.44	28	31	12.43	14.24
16	33	35	19.56	21.44	29	32	12.76	14.57
17	36	38	21.05	22.95	31	34	13.42	15.25
18	37	39	21.55	23.45	33	36	14.08	15.92
19	38	40	22.05	23.95	37	40	15.41	17.26
20	39	41	22.55	24.45	42	45	17.07	18.93
21	40	42	23.05	24.95	28	32	10.13	11.88
22	28	31	12.43	14.24	30	34	10.62	12.38
23	29	32	12.76	14.57	32	36	11.11	12.89
24	30	33	13.09	14.91	36	40	12.1	13.9
25	32	35	13.75	15.58	41	45	13.34	15.16
26	36	39	15.08	16.92	29	34	8.95	10.65
27	38	41	15.74	17.59	31	36	9.34	11.06
28	40	43	16.4	18.26	35	40	10.13	11.88
29	41	44	16.73	18.6	40	45	11.11	12.89
30	29	33	10.37	12.13	28	34	7.84	9.49
31	37	41	12.35	14.15	30	36	8.17	9.83
32	38	42	12.6	14.4	34	40	8.82	10.52
33	39	43	12.84	14.66	39	45	9.63	11.37

34	40	44	13.09	14.91	29	36	7.34	8.95
35	41	45	13.34	15.16	33	40	7.89	9.54
36	29	34	8.95	10.65	38	45	8.58	10.27
37	35	40	10.13	11.88	28	36	6.72	8.28
38	36	41	10.32	12.08	32	40	7.2	8.8
39	37	42	10.52	12.28	37	45	7.8	9.45
40	38	43	10.72	12.48	31	40	6.67	8.22
41	39	44	10.91	12.69	36	45	7.2	8.8
42	35	41	8.98	10.69	30	40	6.25	7.75
43	36	42	9.14	10.86	35	45	6.72	8.28
44	37	43	9.31	11.03	29	40	5.91	7.36
45	40	46	9.8	11.54	34	45	6.34	7.85
46	29	36	7.34	8.95	28	40	5.63	7.03
47	39	46	8.72	10.42	33	45	6.02	7.48
48	33	42	6.88	8.45	32	45	5.75	7.17
49	36	45	7.2	8.8	31	45	5.53	6.9
50	38	47	7.41	9.03	30	45	5.33	6.67
51	35	45	6.72	8.28	29	45	5.17	6.46
52	37	47	6.91	8.49	28	45	5.02	6.27

The table (2) showed the results for the Constant – Service Time model (M/D/1) first when arrival rate (λ) is followed Poisson distribution and service rate (μ) is followed exponential distribution. For the sample size $n=52$ which are used with different values for, service rate (μ), arrival rate (λ) the arrival rate (λ) increases when the service rate (μ) increase and Average number of customers in the system (L) depend on the Average number of customers in the queue (L_q). Cost for system depends on cost for queue. Cost for system and cost for queue was decreased when different between arrival rate (λ) and service rate (μ) increased. Second when Constant – Service Time model (M/D/1) applied with arrival rate (λ) is followed Poisson distribution and service rate (μ) is followed constant values chosen arbitrarily. The cost for the system and the cost for queue for the (M/D/1) is smallest when constant values chosen arbitrarily.

Table (3) the M/G/1 model with $\lambda \sim$ Poisson, $\mu \sim$ exponential, Gamma and Weibull.

Day	M/G/1				M/G/1				M/G/1			
	λ	μ	Waiting cost	system cost	λ	μ	Waiting cost	system cost	λ	μ	Waiting cost	system cost
1	29	30	54.74	56.67	31	32	61.71	63.65	40	41	69.26	71.22
2	30	31	58.14	60.08	39	40	96.78	98.73	41	42	72.26	74.22
3	31	32	61.71	63.65	40	41	69.26	71.22	36	37	82.18	84.13

4	32	33	65.44	67.38	41	42	72.26	74.22	39	40	96.78	98.73
5	33	34	69.35	71.29	42	43	75.36	77.32	39	41	35.02	36.92
6	34	35	73.44	75.39	43	44	78.57	80.52	36	38	43.21	45.11
7	35	36	77.72	79.66	29	31	29.3	31.17	37	39	45.58	47.47
8	36	37	82.18	84.13	33	35	36.71	38.59	37	39	45.58	47.47
9	37	38	75.19	77.14	34	36	38.78	40.67	35	37	40.95	42.84
10	38	39	91.71	93.66	35	37	40.95	42.84	40	42	36.49	38.39
11	39	40	96.78	98.73	37	39	45.58	47.47	33	36	25.84	27.68
12	40	41	69.26	71.22	38	40	48.04	49.94	31	34	23.22	25.05
13	29	31	29.3	31.17	39	41	35.02	36.92	41	44	26.6	28.46
14	30	32	31.02	32.9	40	42	36.49	38.39	29	32	20.83	22.65
15	31	33	32.83	34.71	42	44	39.57	41.48	35	38	28.71	30.55
16	33	35	36.71	38.59	43	45	41.19	43.1	38	41	23.63	25.49
17	36	38	43.21	45.11	30	33	22	23.82	39	42	24.59	26.45
18	37	39	39.6	41.49	35	38	28.71	30.55	38	42	18.66	20.47
19	38	40	48.04	49.94	36	39	30.24	32.09	40	44	20.13	21.95
20	39	41	35.02	36.92	42	45	27.65	29.52	37	41	17.96	19.77
21	40	42	44.05	45.95	31	35	18.43	20.2	33	37	20.42	22.21
22	28	31	19.72	21.53	39	43	19.38	21.2	29	33	16.62	18.37
23	29	32	20.83	22.65	40	44	20.13	21.95	40	45	16.87	18.65
24	30	33	22	23.82	43	47	22.53	24.36	37	42	15.12	16.88
25	32	35	24.5	26.33	44	48	23.38	25.21	38	43	15.68	17.45
26	36	39	30.24	32.09	45	49	24.25	26.09	36	41	14.57	16.33
27	38	41	23.63	25.49	29	34	14.09	15.8	28	33	13.41	15.11
28	40	43	25.58	27.44	30	35	14.81	16.53	39	45	14.2	15.93
29	41	44	26.6	28.46	37	42	15.12	16.88	35	41	12.33	14.04
30	29	33	16.62	18.37	40	45	16.87	18.65	37	43	13.23	14.95
31	37	41	17.96	19.77	35	41	12.33	14.04	30	36	13.03	14.69
32	38	42	26.24	28.05	36	42	12.77	14.49	38	44	13.71	15.43
33	39	43	19.38	21.2	37	43	13.23	14.95	36	42	12.77	14.49
34	40	44	20.13	21.95	37	43	13.23	14.95	29	35	12.42	14.08
35	41	45	20.9	22.73	40	46	14.7	16.44	38	45	12.3	13.99
36	29	34	14.09	15.8	29	36	11.23	12.84	33	40	13.49	15.14
37	35	40	18.95	20.7	29	36	11.23	12.84	36	43	11.49	13.16
38	36	41	14.57	16.33	35	42	11.11	12.77	40	47	13.16	14.86
39	37	42	17.71	19.47	36	43	11.49	13.16	29	36	11.23	12.84
40	38	43	15.68	17.45	40	47	13.16	14.86	35	42	11.11	12.77

41	39	44	16.27	18.04	29	37	10.34	11.91	39	46	12.72	14.42
42	35	41	12.33	14.04	38	46	11.25	12.9	34	41	10.74	12.39
43	36	42	14.81	16.53	40	48	12.01	13.67	37	45	10.88	12.53
44	37	43	13.23	14.95	36	45	9.79	11.39	32	40	11.81	13.41
45	40	46	14.7	16.44	38	47	10.43	12.05	29	37	10.34	11.91
46	29	36	11.23	12.84	40	49	11.11	12.75	38	46	11.25	12.9
47	39	46	12.72	14.42	41	50	11.47	13.11	30	39	10.07	11.61
48	33	42	11.45	13.03	32	42	8.16	9.68	31	40	10.51	12.06
49	36	45	9.79	11.39	36	46	9.2	10.77	29	38	9.65	11.18
50	38	47	10.43	12.05	37	47	9.49	11.06	36	46	9.2	10.77
51	35	45	8.93	10.48	38	48	9.78	11.36	30	40	9.49	10.99
52	37	47	9.49	11.06	44	54	11.77	13.4	32	42	8.16	9.68
average			33.77	35.59			26.78	28.55			23.07	24.81

The table (3) showed the results for general - service queuing model ($M/G/1$) when first arrival rate (λ) is followed Poisson distribution and service rate (μ) is followed exponential distribution, Second with arrival rate (λ) is followed Poisson distribution and service rate (μ) is followed Gamma distribution, third with arrival rate (λ) is followed Poisson distribution and service rate (μ) is followed Weibull distribution. For the sample size $n=52$ which are used with different values for, service rate (μ), arrival rate (λ) the arrival rate (λ) increases when the service rate (μ) increase and Average number of customers in the system (L) depend on the Average number of customers in the queue (L_q). Cost for the system depends on cost for queue. Cost for system and cost for queue was decreased when different between arrival rate (λ) and service rate (μ) increased. The average for the cost for the system and cost for the queue is decreased with (μ) is followed Weibull distribution than the same model with different distributions which used. From the previous results the ($M/G/1$) model is the better model than the two models and the cost is change - when service rate distribution change or with different distributions.

5- Conclusion

This section concerned with the results related with simulation study for the three single channel waiting lines models; Single - Channel ($M/M/1$) , Constant – Service Time model ($M/D/1$), and general - service queuing model ($M/G/1$) when different values for arrival rate (λ) and service rate (μ) are used.

The study comparison between three single channel waiting line models. First when Arinze *et al* [12] data are used for the three models.

The cost which calculated from system which depend on the cost of queue is less for ($M/D/1$) model than the other two models when the same data are used.

Second when (M/D/1) used Arinze *et al* [12] data and data chosen arbitrarily. the results shown that the cost which calculated from system which depend on the cost of queue is less for (M/D/1) model when the data chosen arbitrarily.

Third when (M/G/1) used Arinze *et al* [12] data and generate two distributions used as service rate (μ) the results showed that the system cost and queue cost which calculated from (M/G/1) model when the service rate (μ) is followed weibull distribution is less the system cost and queue cost which calculated from exponential and gamma distributions which used.

The study results that the type of distribution effect on the system cost and queue cost.

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