

Analysis of Waiting lines and queuing system in Banks in India during demonetization: A Case Study

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Abstract

Queuing theory has been fairly a successful tool in the performance analysis of waiting lines. Waiting lines and service efficiency are the important elements for any system. This paper deals with the analysis of waiting lines and queuing system in banks in India during demonetization. Long queues and snaking queues were common sight of banks in India during demonetization. The data for the study was collected by observation in which number of customers arriving at the service window was recorded. Data was collected for a period of six working days from 10.00A.M to 4.00P.M of State Bank of India, NarotJaimal Singh. Data were fitted into the model and the results were computed. The analysis of the results shows that the numbers of existing service windows are not adequate for the customer's service. In order to serve the customer better and reduce the waiting lines in the system, the number of service windows should be increased.

KEYWORDS: Arrival Rate, Service Rate, Service Unit, Servers, Performance measures.

1. Introduction

Queues or waiting lines are very common in everyday life whereby certain business situations require customers to wait for a service, namely: - telephone exchange, at a bank, in public transportation or in a traffic jam, in a supermarket, at a petrol station, at computer systems, waiting to use an ATM machine, and paying for groceries at the supermarket [1]. Studying how these lines form and how to manage them is called Queuing theory. More generally, queuing theory is concerned with the mathematical modeling and analysis of systems that provide service to random demands which deal with one of the most unpleasant experiences of life, waiting [2]. A queuing problem arises when the current service rate or facility falls short of the current service rate of customers.

Delays and queuing problems are most common features of our daily-life situations. Queuing theory was born in the early 1900s with the work of A. K. Erlang of the Copenhagen Telephone Company, who derived several important formulas for teletraffic engineering that today bear his name [2]. Erlang was the first who treated congestion problems

caused by telephone calls where the company requested him to work on the holding times in a telephone switch. He identified that the number of telephone conversations and telephone holding time fit into Poisson distribution and exponentially distributed [3]. This was the beginning of the study of queuing theory.

On November 8, 2016, government of India announced demonetization of all Rs.500 and Rs.1000 bank notes. The government took this step to curtail the shadow of economy and crack down the use of illicit and counterfeit cash to curb black money. As a result, long queues formed inside as well as outside the bank premises. During this period, withdrawal from the banks was limited. People waited for hours to get cash needed to meet their daily expenses. The objective of this paper is to analyze the existing number of service windows to meet customer's need for cash using multiserver queuing models.

2. Queuing process

The process in queuing system is the customers arriving for service, waiting for service if it is not immediate, and leaving the system once they are served [4] [5]. Typical measures of system performance are server utilization, length of waiting lines, and delays of customers, for relatively simple systems, compute mathematically, for realistic models of complex systems, simulation is usually required [6].

Key elements of queuing systems are

2.1.1 Customer: refers to anything that arrives at a facility and requires service.

2.1.2 Server: refers to any resource that provides the requested service,

2.1.3 System Capacity: a limit on the number of customers that may be in the waiting line or system.

- a) **Limited capacity**, e.g., an automatic car wash only has room for 10 cars to wait in line to enter the mechanism.
- b) **Unlimited capacity**, e.g., concert ticket sales with no limit on the number of people allowed waiting to purchase tickets.

2.1.4 Calling population: the population of potential customers may be assumed to be finite or infinite.

- a) **Finite population model:** if arrival rate depends on the number of customers being served and waiting.
- b) **Infinite population model:** if arrival rate is not affected by the number of customers being served and waiting.

2.1.5 Random arrivals: inter-arrival times usually characterized by a probability distribution. Most important model: Poisson arrival process (with rate λ), where A_n represents the inter-arrival time between customer (n-1) and customer n, and is exponentially distributed (with mean $1/\lambda$)

2.1.6 Scheduled arrivals: inter-arrival times can be constant or constant plus or minus a small random amount to represent early or late arrivals.

2.1.7 Queue behavior: the actions of customers while in a queue waiting for service to begin, for example:

- **Balk:** leave when they see that the line is too long,
- **Reneg:** leave after being in the line when it's moving too slowly,
- **Jockey:** move from one line to a shorter line.

2.1.8 Queue discipline: The logical ordering of customers in a queue that determines which customer is chosen for service when a server becomes free [7]. Some common service disciplines are

First-in first-out (FIFO)

Last-in first-out (LIFO)

Service in random order (SIRO)

Shortest processing time first (SPT)

Service according to priority (PR)

2.1.9 Service times and service mechanism: Service times of successive arrivals are denoted by S_1, S_2, S_3, \dots . These service times may be constant or random. The sequence $\langle S_1, S_2, S_3, \dots \rangle$ is usually characterized as a sequence of independent and identically distributed random variables [8].

3. Queuing Notation

A notation system for parallel server queues: A/B/c/N/K, (due to Kendall) [8], where

A represents the inter-arrival-time distribution,

B represents the service-time distribution,

c represents the number of parallel servers,

N represents the system capacity,

K represents the size of the calling population.

Primary performance measures of queuing systems are

P_n : Steady-state probability of having n customers in system,

$P_n(t)$: Probability of n customers in system at time t,

λ : arrival rate,

λ_e : Effective arrival rate,

μ : service rate of one server,

ρ : Server utilization,

A_n : Inter-arrival time between customer n-1 and n,

S_n : Service time of the nth arriving customer,

W_n : Total time spent in system by the nth arriving customer,

W_n^Q : Total time spent in the waiting line by customer n,

$L(t)$: the number of customers in system at time t,

$L_Q(t)$: The number of customers in queue at time t,

L: long-run time-average number of customers in system,

L_Q : Long-run time-average number of customers in queue,

w: long-run average time spent in system per customer,

w_Q : Long-run average time spent in queue per customer.

4. Queuing Models

Queuing models provide the analyst with a powerful tool for designing and evaluating the performance of queuing systems. Model as an idealized representation of the real life situation; in order to keep the model as simple as possible however, some assumptions need to be made [9].

Assumptions

- Single channel queue.
- There is an infinite population from which customers originate.
- Poisson arrival (Random arrivals).
- Exponential distribution of service time.
- Arrival in group at the same time (i.e. bulk arrival) is treated as single arrival.
- The queue discipline is First Come First Served (FCFS).

Although several queuing models abound; designed to serve different purposes [5], highlighted the following:

4.1 The M/M/c/∞; there are c servers to serve from a single line customer, if the arrival is less than or equals to c server every customer is being attended to; if z arrival is greater than the c servers, then z-c customers are waiting in the line.

The service utilization for c servers is given by $\rho = \frac{\lambda}{c\mu}$ (1)

The average number in the line is $L_q = \frac{P_0 \left(\frac{\lambda}{\mu}\right)^c \rho}{c!(1-\rho)^2}$ (2)

where $P_0 = \left[\sum_{m=0}^{c-1} \frac{(c\rho)^m}{m!} + \frac{(c\rho)^c}{c!(1-\rho)} \right]^{-1}$ (3)

P_0 denote the probability that there are 0 customers in the system.

The expected number of customers in the system is

$$L_s = L_q + \frac{\lambda}{\mu} \quad \dots (4)$$

Expected number of customers waiting to be served at any 't' is

$$L_w = \frac{c\mu}{c\mu - \lambda} \quad \dots (5)$$

The average waiting time of an arrival is

$$W_q = \frac{L_q}{\lambda} \quad \dots (6)$$

Average time an arrival spends in the system is

$$W_s = \frac{L_s}{\lambda} \quad \dots (7)$$

4.2 M/M/1 Systems In M/M/1 (∞ /FCFS) queuing system, the arrival and service time both has an exponential distribution, with parameters λ and μ respectively, with one server, queue discipline FCFS and the population size is infinite. The expected inter-arrival time and the expected time to serve one customer are $(1/\lambda)$ and $(1/\mu)$ respectively. An M/M/1 system is a Poisson birth-death process.

5. Research Method Used

The quantitative research method was used in this study. Data was collected for a period of six working days. Data were fitted into the model and the results were computed. This model developed was used to predict the required number of servers.

Table I: Queuing system analysis of the servers for six working days of the week

		Server 1	Server 2	Server 3	Server 4
DAY 1	Arrival Rate	159	99	126	143
Monday	Service Rate	50	32	39	47
DAY 2	Arrival Rate	112	136	150	107
Tuesday	Service Rate	33	44	49	35
DAY 3	Arrival Rate	98	97	103	99
Wednesday	Service Rate	32	29	35	40
DAY 4	Arrival Rate	109	149	87	133
Friday	Service Rate	35	47	27	50
DAY 5	Arrival Rate	116	123	114	129
Saturday	Service Rate	39	42	36	40
DAY 6	Arrival Rate	89	143	135	94
Sunday	Service Rate	24	45	43	31
Total	Arrival Rate	683	747	715	705
	Service Rate	213	239	229	243



Average System Utilization	$\frac{683}{213} = 3.2065$	$\frac{747}{239} = 3.1255$	$\frac{715}{229} = 3.1222$	$\frac{705}{243} = 2.9012$
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Table II: Queuing system analysis of the servers for six working days of the week

		Server 1	Server 2	Server 3	Server 4
DAY 1 Monday	Average Arrival Rate	26.5	16.5	21	23.83
	Average Service Rate	8.33	5.33	6.5	7.83
DAY 2 Tuesday	Average Arrival Rate	18.66	22.66	25	17.83
	Average Service Rate	5.5	7.33	8.16	5.83
DAY 3 Wednesday	Average Arrival Rate	16.33	16.16	17.16	16.5
	Average Service Rate	5.33	4.83	5.83	6.66
DAY 4 Friday	Average Arrival Rate	18.16	24.83	14.5	22.16
	Average Service Rate	5.83	7.83	4.5	8.33
DAY 5 Saturday	Average Arrival Rate	19.33	20.5	19	21.5
	Average Service Rate	6.5	7	6	6.66
DAY 6 Sunday	Average Arrival Rate	14.83	23.83	22.5	15.66
	Average Service Rate	4	7.5	7.16	5.16

Table III

	Server 1	Server 2	Server 3	Server 4
DAY 1 (Monday)	3.1812	3.0956	3.2307	3.0434
DAY 2 (Tuesday)	3.3927	3.0914	3.0637	3.0583
DAY 3 (Wednesday)	3.0637	3.3457	3.0463	2.4774
DAY 4 (Friday)	3.1149	3.1711	3.2222	2.6602
DAY 5 (Saturday)	2.9738	2.9285	3.1666	3.2282
DAY 6 (Sunday)	3.7075	3.1773	3.1424	3.0348



Table IV

	Server 1	Server 2	Server 3	Server 4	
Customer Arrival Rate (λ_i)	18.9683	20.7467	19.86	19.58	Average Arrival Rate $\lambda = 19.7888$
Customer Service Rate (μ_i)	5.915	6.6366	6.3583	6.745	Average Service Rate $\mu = 6.4137$
Average No. of Customers served $\left(\frac{\lambda_i}{\mu_i}\right)$	3.2068	3.1261	3.1234	2.9028	$\frac{\lambda}{\mu} = 3.0853$

Table V: Results of Performance measures of three server analysis

c	1	2	3	4	5	6	7	8	9
L_Q	-4.5649	-5.3213	-38.898	1.8414	0.4155	0.1170	0.0339	0.0095	0.0025
P_o	-2.0854	-0.2134	-0.0061	0.0331	0.0424	0.0448	0.0455	0.0457	0.0457
L_S	-1.4795	-2.2359	-35.044	4.9268	3.5009	3.2024	3.1192	3.0949	3.0879
W_Q	-0.2307	-0.2689	-1.9248	0.0931	0.0210	0.0059	0.0017	0.0005	0.0001
W_S	-0.0748	-0.1130	-1.7689	0.2490	0.1769	0.1618	0.1576	0.1564	0.1560

6. DISCUSSION OF RESULTS

From the above table, it has been observed that existing number of service windows required to serve the customers is not sufficient. The suitable number of servers that can serve the customers as and at when necessary without waiting for long before customers are served at the actual time should be more than four. This increase in number of service windows reduces the waiting time.

7. CONCLUSIONS

The result analysis of the above queuing system shows the need to increase the number of the service windows. The increase in the number of service windows will reduce the time that customers have to wait in line before being served.

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