

A Fuzzy-Based Server Incremental Technique for N-Policy M/G/n Queue Network

ABSTRACT

The study presents an N-Policy M/G/n queue model having multiple servers with possible increment. In this model, the server is turned off as soon as the queue is empty. However, as customers arrive the network according to a homogeneous Poisson process with rate λ , the server is not immediately turned on until when the number of customers reaches a pre-determined threshold value, N . Service of customers is on first-come, first-served basis. The model has two categories of servers viz: active and reserved servers. In a situation where all active servers are busy and a customer arrives, two options are available. It is either the arriving customer wait until one of the active servers becomes idle or request for an additional server from among reserved servers. The decision of the customer to wait for one of the active servers to become idle or request an additional server in order to reduce wait time is taken by an Expert System rather than making such on the basis of network performance metrics.

Keywords: [Queue size, Expert System, Network performance metrics, Threshold value, Throughput]

1. INTRODUCTION

Queues are part of everyday's life. This is so because people wait in cars, banks, hotels, supermarkets, box offices, airports, hospitals and so on. These are examples of visible queues. In fact, queues of voice calls or data packets in communication channels are common but invisible. The large size of traffic in communication lines or computer networks is also a reason why queues cannot be easily avoided [1]. Queues are often undesirable because they cost time, money and resources. They exist because the service resources are not sufficient to satisfy demand. This is because of a number of reasons. Servers may be unavailable because of space or cost limitations, or it may not always pay to provide the level of service necessary to prevent waiting.

Queuing theory is the branch of mathematics which deals with the study of waiting lines. A queue is formed when customers arrive to a service location expecting to be served with limited resources. If the server is not immediately available, the customers need to join a waiting line. The use of queuing theory allows the study of different processes associated with queues including arrivals, waiting and service[2]. The applications of queuing theory in traffic flow, telecommunications and facility design, provides a clear usage of the method in solving a wide range of industrial and domestic problems.

Queuing theory uses mathematical tools to predict the behaviour of queuing systems. Predictions deal with the probability to have n customers in the system, mean length of queues, mean waiting time, throughput and so on. A queuing system consists of a stream of arriving customers, a queue and a service process as well as the number of servers. Generally, a queue has the following components:

- a. A stochastic process describing the arrivals of customers;
- b. A stochastic process describing the service system of customers;

- 40 c. The system capacity;
- 41 d. The size of customer population; and
- 42 e. The queue discipline such as First In, First Out (FIFO); Last In, First Out and so on.

43 Traditionally, queuing theory considers models with a fixed number of servers. In
 44 most of these cases, the main performance metrics considered are queue length and waiting
 45 time [3], consequently posing great restrictions on the performance of such system.
 46 Advancements in service requirement as well as flexibility in service's delivery had changed
 47 this pattern. Consequently, it is more optimal to consider queuing systems with a changing
 48 number of servers depending on the queue length.

49 Queueing models have wider range of applications in service organizations as well
 50 as in manufacturing firms, where customers receive service by different kinds of servers in
 51 accordance with the queue discipline. In particular, the inter-arrival times and service times
 52 are restricted to follow specific probability distributions [4]. In some queuing systems, it is
 53 required that a certain level of queuing performance, such as the mean queuing delay or the
 54 blocking (queuing) probability, be guaranteed for its customers. In classical queuing
 55 systems, meeting stringent performance requirements usually results in inefficient server
 56 utilization. In some cases, such as in traditional telephone networks, frequently adjusting the
 57 number of servers may not be economically justifiable. In order to improve servers' utilization
 58 in this situation, the number of servers can be adjusted over a relatively large time scale
 59 such as on a daily or weekly basis, according to the forecast of future demand.

60 An N-policy queue refers to a queuing system in which the server does not start its
 61 service until there are N customers waiting in the queue. This policy is often used to avoid
 62 excessively frequent setups and to minimize servers' cost. The need to adequately
 63 determine the number of servers to provide required services in a queue network is
 64 paramount as it ensures that expected services are not only offered, but that such are
 65 offered within the shortest possible time. In addition, it is not only important to ensure
 66 adequate availability of required server(s), it is equally important to ensure that available
 67 ones are put to optimal use. In static queue networks, consideration is given to models with
 68 fixed number of servers. This poses great restrictions on the performance of such system
 69 [3].

70 One of the important methods to resolving conflict between meeting stringent
 71 performance requirements and achieving optimal server utilization is to adjust the number of
 72 servers dynamically over time rather than keeping a fixed number of servers all the time.
 73 This problem was formulated by [5]. In this study, the authors associated the number of
 74 servers $S(t)$ in a queue network as a function of time. This is minimized subject to the
 75 constraint that the probability of a delay never exceeds a target probability, given the
 76 characteristics of the time-dependent arrival process as a function of time. When the change
 77 in the number of servers in a queuing system is economically feasible, server utilization
 78 could be improved by adjusting the number of servers according to the number of
 79 customer(s) in the system at time t [6].

80 In an N-Policy system, the turning on of server depends on the number of customers
 81 in the system. When the number of customers in the system reaches a threshold of $N(N \geq 1)$,
 82 the server is turned on but not immediately accessible to waiting customers until start-up is
 83 completed. After this, the server immediately begins serving waiting customers [7]. A
 84 common type of N-Policy is called (v, N) -policy, with $0 \leq v \leq N < +\infty$, according to which the
 85 server is turned on when N customers are present and the server is turned off when it
 86 terminates a service with v customers left in the system [8]. This duration of the 'start-up' are
 87 independent and identically distributed random variables of the general distribution function

88 $U(t)$, where $t \geq 0$ with a mean startup time μ_U and a finite $\frac{\partial^2 U}{\partial t^2}$. Similarly, the service times for

89 a customer are independent and identically distributed random variables for arbitrary
90 distribution function S_t , where $t \geq 0$, a mean service time μ_S and a finite variance $\frac{2}{\sigma_S^2}$.

91 Although efficient methods have been developed for analyzing queueing system
92 when arrival rate and service rate of customers are known, however in practical applications
93 when the arrival rate and service rate are described using linguistic terms rather than
94 numerical values, it becomes complicated to evaluate the performance measures of such a
95 queueing system using statistical theory [4]. In this situation, [9] introduced the concept of
96 "fuzziness". Consequently, fuzzy queueing model was first introduced by [10]. As an
97 extension of this, [11], [12] and [13] improved the concept of "fuzziness" as introduced by [9].

98 Research findings by [14], the author analyzed fuzzy queueing models using
99 Day/Stout/Warren (DSW) algorithm while [15] analyzed fuzzy N-Policy queues with infinite
100 capacity. Similar to this, [16] implemented DSW algorithm for the brief description of his
101 fuzzy queueing model while [17] studied fuzzy queueing model with multiple servers using
102 the same algorithm and also executed its performance measures.

103 Automated systems based on fuzzy logic have been used widely in control
104 systems, household appliances, decision-making systems, the medical and automobile
105 industries [9]. While Boolean algebra set values only include "1" and "0" or "True" and
106 "False", it is believed through fuzzy logic that there are other values between "1" and "0" or
107 "True" and "False", which are sometimes referred to as in-between values. In other words,
108 Boolean logic engages the principles of totally inclusive and exclusive rules on its set of "1"
109 and "0" while the principles of totally inclusive, exclusive and 'in between values' rules is
110 engaged in fuzzy logic [18].

111 Related works to this study can be broadly grouped into two as follows:

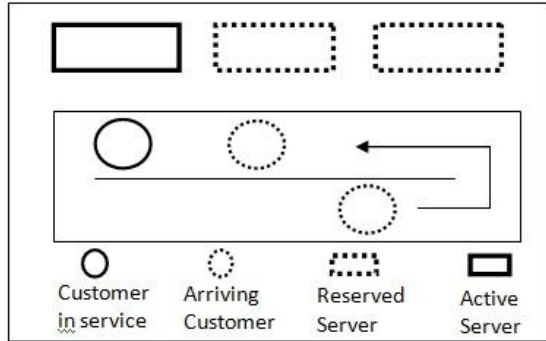
112 a. A queue network with servers having different service rates and researchers aim at
113 allocating incoming customers to optimize network performance. This gives opportunity to
114 customers to move from long queues to shorter ones or even leave the queue. In this
115 case, [19] proposes that the optimal number of servers is of the form $\lambda + \gamma \sqrt{\lambda}$
116 depending on the total arrival rate λ for a given grade of service γ . In order to ensure
117 performance optimality, multi-threshold strategies could be adopted as it gives customers
118 opportunity to take decisions when the queue to a given server exceeds a certain
119 threshold [20]; and

120 b. A queue network with identical servers and researchers aim to distribute customers
121 among available servers which can become active or inactive [5].

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123 This study is aimed at the use of a fuzzy-based Expert System in servers'
124 management in a queue network. In essence, flexibility in the management of servers
125 becomes highly inevitable. The study becomes necessary taking cognizance of the level
126 of flexibility needed in today business environment involving the application of queues in
127 service delivery. Unlike previous studies in which dynamic servers' management is
128 premised on increasing the number of servers correspondingly as the number of
129 customers arriving the network increases in order to save time, this study aims at the use
130 of dynamic management of servers using a fuzzy - based Expert System.

131 The Expert system manages servers in the queue network by the application of
132 fuzzy rules on input variables to produce an output and consequently applying other fuzzy
133 procedures to arrive at decisions as far as servers' management is concerned in the
134 queue network. A practical application of the proposed model is a customers' service unit
135 of a telecommunication firm. Customers make calls and also use various unstructured
136 supplementary service data codes on their phones to make inquiries on services, request
137 for service upgrade, migrate from one service plan to another, buy airtime and data,
138 among others. In most cases, these requests and services are managed using automated

139 systems which are limited in number. The queuing system considered in this application
 140 is illustrated in figure 1.
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144 Figure 1: Queuing system in a typical customer care unit of a telecommunication firm

145 Active and reserved servers which are the customers' care service providers as in
 146 our case, share the same queue and are shown as rectangles. If the active server is busy
 147 attending to a subscriber and a request is made by another subscriber, the system decides
 148 whether or not to take an additional server from among reserved servers to attend to the
 149 arriving request or the new request wait in queue until an active server is idle. In figure 1,
 150 there is one active server and two reserved servers, a request in service and an arriving
 151 request. The proposed model is flexible such that the fuzzy-based Expert System manages
 152 the number of servers in the network taking cognizance of input variables and fuzzy rules
 153 which are applied in order to arrive at a decision.

154 In a multi-server queue system, customers arrive at rate λ . Each customer is served
 155 by one server and an arriving customer waits in queue when all servers are busy. There are
 156 s servers so that the maximum service rate of the queue is μ , where μ is the service rate of
 157 individual servers. If the number of customers in the queue, n , is less than the number of
 158 servers, s , the service rate equals $n\mu$. Similarly, in order to ensure queue stability, it is
 159 required that the amount of work that arrives per unit time ρ is less than the maximum
 160 service rate, i.e. $\rho = \lambda E[S] < s$ [21]. In this case, the equilibrium distribution is obtained
 161 using (1) as follows:

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$$\lambda P_0 = \mu P_1$$

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$$(\lambda + n\mu) P_n = \lambda P_{n-1} + (n+1)\mu P_{n+1} \text{ for } n < s,$$

 164
$$(\lambda + s\mu) P_n = \lambda P_{n-1} + (s+1)\mu P_{n+1} \text{ for } n \geq s.$$

165 Consequently,
$$P_n = \frac{\rho^n}{m(n)} P_0, \quad (1)$$

166 where
$$m(n) = \begin{cases} n! & 0 \leq n < s \\ s! & n \geq s \end{cases}$$

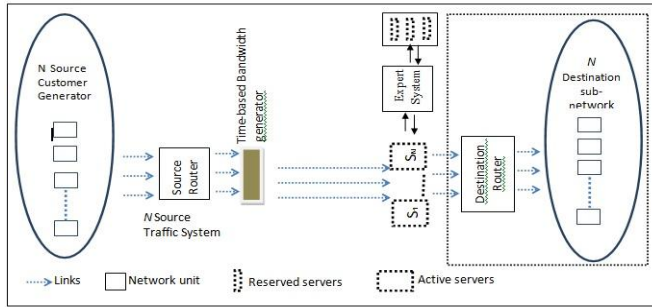
 167 for $0 \leq n$
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170 2. MATERIAL AND METHODS

171 This is discussed under the following sub-headings: schematic structure of the proposed
 172 system and fuzzy approach to servers' increment in the proposed model.
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174 2.1. Schematic structure of the proposed system

175 The proposed model considers a case of N customers traveling through and contending for
 176 service in a queue network as depicted in figure 2.



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 178 Figure 2: Proposed queue and servers' system

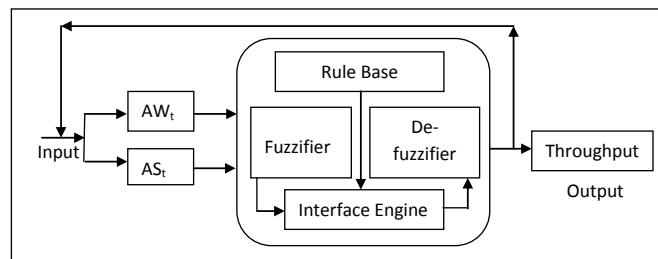
- 179 The proposed model in figure 2 has the following components:
- 180 a. N source customers' generator: This generates sequence of customers and transmits
 181 them into the network over available links;
 - 182 b. N source traffic system: This is a source router used to transmit customers to the idle
 183 servers;
 - 184 c. Time-based bandwidth generator: This generates random numbers typical of bandwidth
 185 sizes or capacity over a known range of time $0 \leq t \leq T$;
 - 186 d. Active servers: These serve available customers as they arrive the network;
 - 187 e. Expert System: When an arriving customer gets into the system and found no idle server,
 188 the Expert System chooses whether the numbers of active servers be increased by a unit
 189 from among reserved servers in order to serve the arriving customer or to allow it wait in
 190 the system until one of the busy servers becomes idle;
 - 191 f. Reserved servers: These are reserved servers from among which the system chooses to
 192 increase the number of active servers whenever the need arises; and
 - 193 g. N destination sub-network: This route served customers to their respective destinations.

194 **2.2. Fuzzy approach to servers' increment in the proposed model**

195 A fuzzy control system is a rule-based system in which a set of rules, called fuzzy rules,
 196 define a control mechanism to adjust the system [22]. Generally, a fuzzy logic controller for
 197 queues comprises of four principal components: a fuzzification interface, a knowledge base,
 198 an inference engine as well as a de-fuzzification interface [23]. The output of the fuzzy logic
 199 controller is used to tune the system parameters according to some predefined program
 200 which is based on the state of the system and it is adaptive in nature.

202 **2.2. Simulation**

203 Matlab trial version was used to simulate the model. As customers arrive, the system
 204 computes corresponding values of the average wait time (AW_t) and average service time
 205 (AS_t). At a point, the values of AW_t and AS_t were 59.5 and 44.9 respectively giving the
 206 current throughput based on the set of rules using the proposed fuzzy controller depicted in
 207 figure 3.



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Figure 3: Fuzzy controller used

214 Each of the two crisp inputs, i.e. AW_i and AW_s were classified into five linguistic variables of
215 "Extremely Low", "Low", "Normal", "High" and "Extremely High" represented as "EL", "L", "N",
216 "H" and "EH" respectively. The throughput which is the output was also classified into five
217 linguistic variables as applicable to the input variables. When a customer arrives, the current
218 values of AW_i and AW_s were obtained and the corresponding throughput is calculated based
219 on the two inputs and the set of rules. There are rules on the basis of which the system
220 operates. The rule function, f is defined as follows:

221
$$f = \{F, G, V, E\}$$

222 where "F" is "Fair", "G" is "Good", "V" is "Very Good" and "E" is "Excellent". The
223 corresponding fuzzy rules table is given in table 1.
224

		Average Wait Time (AW_i)				
		EL	L	N	H	EH
Average Service Time (AS_i)	EL	F	F	G	G	V
	L	F	G	G	V	E
	N	G	G	V	V	E
	H	G	V	V	V	E
	EH	G	V	V	E	E

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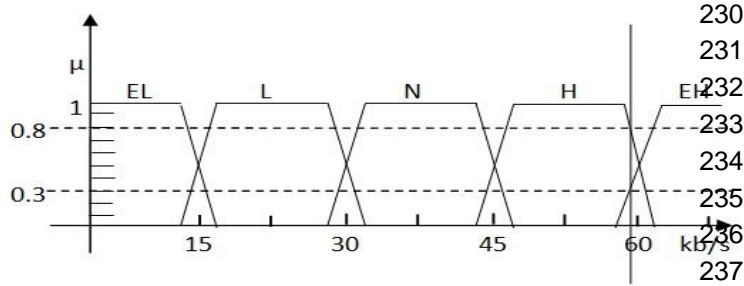
Table 1: Fuzzy rules table adopted

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Consequently, the membership of AW_i is given in figure 4.

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Figure 4: Membership of AW_i

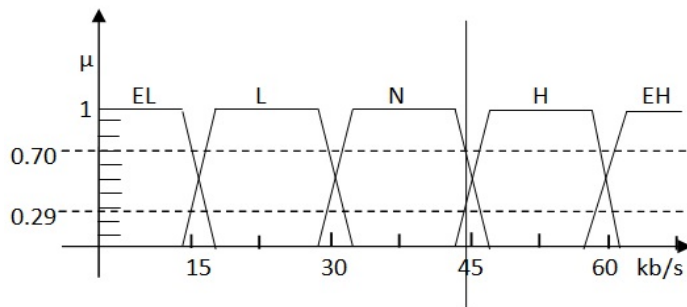
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In a similar way, the membership of AS_i is given in figure 5.

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Figure 5: Membership of AS_i

256 The final decision (FD) is generated based on the minimum operations as indicated in
 257 figures 4 and 5 after the minimum value operation. The values are 0.3, 0.8, 0.29 and 0.70.
 258 The computation of the FD was made using centroid method as indicated in (2) below:
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$$260 \quad FD = \frac{\mu_1 D_1 + \mu_2 D_2 + \dots + \mu_n D_n}{\mu_1 + \mu_2 + \mu_3} \quad (2)$$

261
 262 Substituting the minimum values in this equation gives:

$$263 \quad FD = \frac{(0.29 \times 0.2) + (0.70 \times 0.4) + (0.29 \times 0.6) + (0.3 \times 0.8)}{0.29 + 0.70 + 0.29 + 0.3}$$

$$264 \quad FD = 0.5$$

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 266 The FD value of 0.5 is plotted to derive the decision index as indicated in figure 6.

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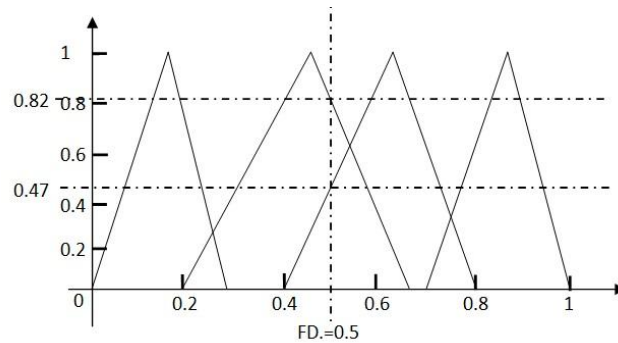


Figure 6: Fuzzified decision index

2.3. Discussion

From figure 6, it is obvious that at 0.82, the 'wait time' of customers is high. This implies that there is a significant number of customers waiting for service turns. Consequently the Expert System request for additional one server from among the reserved servers to complement the services of the active ones in order to reduce the wait time of the customers in the buffer. Similarly, at 0.47, the 'wait time' of customers is reasonable. This implies that the rate of customers' arrival has not exceeded the capacity of the active servers. Consequently, if a customer arrives and does not find any idle server to service it, it waits until one of the busy servers become idle. This decision is taken by the Expert System within the system and not taken on the basis of any network performance metric. This implies that every time a customer arrives the system and found no idle server, the current value of AW_t and AS_t are obtained and the throughput is calculated based on the two inputs and the set of rules and decision is taken on the basis of the output.

3. CONCLUSION

The study describes an N-Policy M/G/n queue model with possible server increment in which customers are served on first-come, first-served basis by active servers. As customers arrive the system and found no idle server, the Expert System determines the action to take using fuzzy logic approach, consequently making the management of servers dynamic. The contribution of the study is that the model is able to dynamically manage the number of servers in queue network using fuzzy logic approach as against the usual idea of correspondingly increasing the number of servers in a queue system as the number of

302 customers increases or by considering certain network performance metrics. The proposed
303 model does not necessarily increment servers correspondingly to service requests of arriving
304 customers but also ensures that available servers are put to optimal use.

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306 **ACKNOWLEDGEMENTS**

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311 **COMPETING INTERESTS**

312 Authors have declared that no competing interests exist.

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315 **AUTHORS' CONTRIBUTIONS**

316 All the authors prepared the manuscript, carried out the simulation, read the manuscript and
317 approved it for submission.

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320 **4. REFERENCES**

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