A Communication Network with Dynamic Bandwidth Allocation having Bulk Arrivals using Zero Truncated Binomial Distribution

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ABSTRACT-

In most of the recent communication systems such as Telecommunications, Satellite Communications, ATM Networks and the Internet, the data switching nodes plays a vital role in forwarding the data from the sender to the receiver located at different places. In this paper we assumed that the communication system is in equilibrium, the performance measures like the joint probability generating function of the buffer size distribution, the average content of the buffers, the mean delays in transmission, the throughput of the nodes and utilization are derived explicitly. We observed the dynamic bandwidth allocation (DBA), bulk arrival consideration have a great impact on performance measures and the results are very close to the reality. We considered zero truncated binomial distribution for the batch arrival of input data.

Keywords- Communication networks, Dynamic bandwidth allocation, Bulk arrivals, Performance measures.

1. INTRODUCTION

Much work has been reported regarding communication networks and their performance evaluation. Martin Reiser (1982) and Jaime Jungok Bae(1991) have reviewed the communication networks and analytical methods for their evaluation. Several authors developed various communication network models with several considerations in order to analyze the situation close to the reality. One of the important considerations in communication network model is transporting data/voice more effectively with a guaranteed Quality of Service (QoS). For efficient communication, different service models have been proposed a scalable traffic management mechanism to ensure QoS. However, in some situations at broadband integrated services, the digital network has a synchronized transmission mode. The output of one transmitter is usually the input of another transmitter.

Due to the unpredicted nature of the transmission lines, congestion occurs in communication systems. In order to analyze the communication network efficiently, one has to consider the analogy between communication networks and waiting line models. Generally, the analysis in a communication system is mainly concerned with the problem of allocation and distribution of data or voice packetization, statistical multiplexing, flow control, bit dropping, link assignment, delay and routing etc. For efficient utilization of the resources, mathematical modeling provides the basic frame work in communication networks. The communication networks are modeled as interconnected queues by viewing the message as the customer, communication buffer as waiting line and all activities necessary for transmission of the message as service. This representation is the most natural with respect to the actual operation of such systems. This leads a communication network to view as a tandem or serial queuing network. Several authors have studied the communication networks as tandem queues (Kleinrock L. 1976; Yukuo hayshida, 1993; Paul Dupis et al, 2007).

Because of the unpredicted nature of demand at transmission lines congestion occurs in communication systems. Statistical multiplexing is one of the major considerations for efficient utilization of the resources. With the statistical multiplexing load dependent communication network models have been generated to accommodate the bit dropping methodologies (Kin K. Leung, 2002). Bit dropping method can be classified as IBD (Input Bit Dropping) and OBD (Output Bit Dropping). Depending on the implementation of the actual algorithms, IBD or OBD performance is measured. As a result of the bit dropping or flow control strategies voice quality is expected to degrade gracefully when overload occurs. The extent of degradation of service quality is a function of the fraction of voice calls lost, which in turn depends on the load. To have an efficient transmission with high quality, it is needed to consider the variation on transmission rates based on the contents of the buffers. This is often referred as dynamic bandwidth allocation.

Some algorithms have been developed with various protocols and allocation strategies for optimal utilization of bandwidth (Emre and Ezhan, 2008; Gundale and Yardi, 2008; Hongwang and Yufan, 2009; Fen Zhou et al. 2009; Stanislav, 2009). These strategies are developed based on arrival process of the packets through bit dropping and flow control techniques. It is needed to utilize the bandwidth maximum possible by developing strategies of transmission control based on buffer size. One such strategy is dynamic bandwidth allocation. In dynamic bandwidth allocation, the transmission rate of the packet is

adjusted instantaneously depending upon the content of the buffer. Recently P.Suresh Varma et al (2007) has developed some communication network models using dynamic bandwidth allocation. However, they considered that the arrivals of packets to the buffer are single. But, in store-and-forward communication the messages are packetized and transmitted. When a message is packetized, the number of packets of that message is random having bulk in size. Hence, considering single packet arrival to the initial node may not accurately evaluate the performance of the communication network. Therefore, in this paper, a communication network with dynamic bandwidth allocation having bulk arrivals is developed and analyzed under Markovian environment. This model wick characterize the two node communication networks arising at places like Telecommunications, Satellite communications, computer communications etc. more close to the realistic situation.

The statistical multiplexing of the communication network is characterized by considering that the arrival and transmission process follow Poisson and the number of arrivals to the initial node is in batch with a random size having uniform distribution. The performance of the communication network is carried through deriving the joint probability generating function of the buffer size distribution, the mean content of the buffers, the mean delay in transmission, the throughput etc. A numerical illustration for sensitivity of the network with respect to the input parameters is given.

2. A COMMUNICATION NETWORK WITH DYANMIC BANDWIDTH ALLOCATION AND BULK ARRIVALS

Tandem communication systems received lot of attention in literature. These communication networks consisting of two nodes, having two buffers transmits data over a common shared system. In these communication systems, the messages are packetized at the source and stored in buffers for transmission. When a message arrives, it is converted into data/voice packets of a pre-fixed length. The number of packets that a message can be converted is random depending upon the length of the message. Hence, the arrivals of packets to the initial buffer is in bulk having a variable size. After being transmitted in the first node, it is transmitted through second node. In both the nodes the transmission is carried with dynamic bandwidth allocation strategy. In DBA, the transmission rate is a linear function of number of packets in the buffer. Here, it is assumed that the arrival of packets follow a compound Poisson process with arrival rate $\lambda E(X)$, where X is the number of packets a message can be converted and having probability mass function as "C_k" {C_k, k=1, 2, 3,,}. The number of transmissions at each node also follows Poisson with parameters " μ_1 " and " μ_2 " depending upon the number of packets in first and second buffers respectively. The queue discipline is First-in-First-Out (FIFO). Let n₁ and n₂ are the random variable which denotes the number of packets in first and second buffer respectively. The schematic diagram representing a communication network model with bulk arrival under equilibrium is shown in figure 1.



Figure 1. Communication network with dynamic bandwidth allocation and bulk arrivals

Let $P_{n_1,n_2}(t)$ be the probability that there are n1 packets in the first buffer and n2 packets in the second buffer at time t. The difference - differential equations of the Communication network are $\partial P_{n_1}(t)$

$$\frac{\partial P_{n_1,n_2}(t)}{\partial t} = -(\lambda + n_1\mu_1 + n_2\mu_2)P_{n_1,n_2}(t) + (n_1 + 1)\mu_1P_{n_1 + 1,n_2 - 1}(t) + (n_2 + 1)\mu_2P_{n_1,n_2 + 1}(t) + \lambda \left\lfloor \sum_{k=1}^{n_1} P_{n_1 - k,n_2}(t)C_k \right\rfloor$$
(1)

$$\frac{\partial P_{n_{1},0}(t)}{\partial t} = -(\lambda + n_{1}\mu_{1})P_{n_{1},0}(t) + \mu_{2}P_{n_{1},1}(t) + \lambda \sum_{k=1}^{n_{1}} P_{n_{1}-k,0}(t)C_{k} \quad (2)$$

$$\frac{\partial P_{0,n_{2}}(t)}{\partial t} = -(\lambda + n_{2}\mu_{2})P_{0,n_{2}}(t) + \mu_{1}P_{1,n_{2}-1}(t) + (n_{2}+1)\mu_{2}P_{0,n_{2}+1}(t) \quad (3)$$

$$\frac{\partial P_{0,0}(t)}{\partial t} = -\lambda P_{0,0}(t) + \mu_{2}P_{0,1}(t) \quad (4)$$

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$$\frac{\partial P_{1,0}(t)}{\partial t} = -\lambda \mu_1 P_{1,0}(t) + \mu_2 P_{1,1}(t) + \lambda P_{0,0}(t) C_1(5)$$

$$\frac{\partial P_{0,1}(t)}{\partial t} = -(\lambda + \mu_2) P_{0,1}(t) + \mu_1 P_{1,0}(t) + 2\mu_2 P_{0,2}(t)$$
(6)

with initial conditions

 $P_{00}(0) = 1; P_{n_1,n_2}(0) = 0 \text{ for } n_1, n_2 > 0$

Let $P(z_1, z_2; t)$ be the joint probability generating function of $P_{n_1, n_2}(t)$ then

$$P(z_{1,}z_{2};t) = \sum_{n_{1}=0}^{\infty} \sum_{n_{2}=0}^{\infty} Z_{1}^{n_{1}} Z_{2}^{n_{2}} P_{n_{1},n_{2}}(t)$$
(7)

multiplying the equations (1) to (6) with corresponding $Z_1^{n_1}, Z_2^{n_2}$ and summing overall $n_1=0, 1, 2, 3, ...$ and $n_2=0, 1, 2, 3, ...$ one can get

$$\begin{split} \sum_{n_{1}=1}^{\infty} \sum_{n_{2}=1}^{\infty} \frac{\partial}{\partial t} P_{n_{1},n_{2}}(t) Z_{1}^{n_{1}} Z_{2}^{n_{2}} &= -\left[\sum_{n_{1}=1}^{\infty} \sum_{n_{2}=1}^{\infty} (\lambda + n_{1}\mu_{1} + n_{2}\mu_{2}) P_{n_{1},n_{2}}(t) Z_{1}^{n_{1}} Z_{2}^{n_{2}} \right] \\ &+ \sum_{n_{1}=1}^{\infty} \sum_{n_{2}}^{\infty} (n_{1} + 1)\mu_{1} P_{n_{1}+1,n_{2}-1}(t) Z_{1}^{n_{1}} Z_{2}^{n_{2}} \\ &+ \sum_{n_{1}=1}^{\infty} \sum_{n_{2}}^{\infty} (n_{2} + 1)\mu_{2} P_{n_{1},n_{2}+1}(t) Z_{1}^{n_{1}} Z_{2}^{n_{2}} \\ &+ \sum_{n_{1}=1}^{\infty} \sum_{n_{2}}^{\infty} \left[\lambda \sum_{k=1}^{n_{1}} P_{n_{1}-k,n_{2}}(t) . C_{k} \right] Z_{1}^{n_{1}} Z_{2}^{n_{2}} \\ &+ \sum_{n_{1}=1}^{\infty} \sum_{n_{2}}^{\infty} \left[\lambda \sum_{k=1}^{n_{1}} P_{n_{1}-k,n_{2}}(t) . C_{k} \right] Z_{1}^{n_{1}} Z_{2}^{n_{2}} \\ &\text{After simplification,} \quad \frac{\partial P(Z_{1}, Z_{2}; t)}{\partial t} = \left[\lambda (c(Z_{1}) - 1) \right] P(Z_{1}, Z_{2}; t) + \left[\mu_{1}(Z_{2} - Z_{1}) \right] . \frac{\partial P(Z_{1}, Z_{2}; t)}{\partial Z_{1}} + \left[\mu_{2}(1 - Z_{2}) \right] . \frac{\partial P(Z_{1}, Z_{2}; t)}{\partial Z_{2}} \end{split}$$

Rearrange the terms

$$\frac{\partial P(Z_1, Z_2; t)}{\partial t} - \left[\mu_1(Z_2 - Z_1)\right] \cdot \frac{\partial P(Z_1, Z_2; t)}{\partial Z_1} - \left[\mu_2(1 - Z_2)\right] \cdot \frac{\partial P(Z_1, Z_2; t)}{\partial Z_2} = \left[\lambda(c(Z_1) - 1)\right] P(Z_1, Z_2; t)$$
(9)

Using the Lagrangian's method, the auxiliary equations of the equation (9) are

$$\frac{\partial t}{l} = \frac{-\partial Z_1}{\mu_1(Z_2 - Z_1)} = \frac{-\partial Z_2}{\mu_2(1 - Z_2)} = \frac{\partial P(Z_1 Z_2; t)}{\left[\lambda(c(Z_1) - 1)\right] P(Z_1, Z_2; t)}$$
(10)
Solving the equation (10) which has the solution with initial conditions

Solving the equation (10) which has the solution with initial conditions $P_{00}(0) = 1; P_{n_1,n_2}(0) = 0$ for all $n_1, n_2 > 0$

$$\mathbf{u} = (\mathbf{Z}_2 - \mathbf{1})\mathbf{e}^{-\mu_2 \mathbf{t}} \tag{11}$$

$$\mathbf{v} = \left[(\mathbf{Z}_1 - 1) + \frac{\mu_1}{\mu_2 - \mu_1} (\mathbf{Z}_2 - 1) \right] \mathbf{e}^{-\mu_1 t}$$
(12)

$$w = P.exp\left[-\lambda \sum_{k=1}^{\infty} \sum_{r=1}^{k} \sum_{J=0}^{r} (-1)^{2r-J} C_{k} ({}^{k}C_{r}) ({}^{r}C_{J}) \left(\frac{u\mu_{1}}{\mu_{2}-\mu_{1}}\right)^{J} v^{(r-J)} \frac{e^{[J\mu_{2}+(r-J)\mu_{1}]t}}{J\mu_{2}+(r-J)\mu_{1}}\right]$$
(13)

Where u, v and w are arbitrary integral constants. Therefore

$$P = w.exp \left[\lambda \sum_{k=1}^{\infty} \sum_{r=1}^{k} \sum_{J=0}^{r} (-1)^{2r-J} C_{k} ({}^{k}C_{r}) ({}^{r}C_{J}) \left(\frac{u\mu_{1}}{\mu_{2} - \mu_{1}} \right)^{J} v^{(r-J)} \frac{e^{[J\mu_{2} + (r-J)\mu_{1}]t}}{J\mu_{2} + (r-J)\mu_{1}} \right]^{2r-J}$$

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Substituting the value of 'w'

$$P = \left(\exp\left[\lambda \sum_{k=1}^{\infty} \sum_{r=1}^{k} \sum_{J=0}^{r} (-1)^{2r-J} C_{k} ({}^{k}C_{r}) ({}^{r}C_{J}) \left(\frac{u\mu_{1}}{\mu_{2} - \mu_{1}} \right)^{J} v^{(r-J)} \frac{1}{J\mu_{2} + (r-J)\mu_{1}} \right] \right)$$
$$\left(\exp\left[\lambda \sum_{k=1}^{\infty} \sum_{r=1}^{k} \sum_{J=0}^{r} (-1)^{2r-J} C_{k} ({}^{k}C_{r}) ({}^{r}C_{J}) \left(\frac{u\mu_{1}}{\mu_{2} - \mu_{1}} \right)^{J} v^{(r-J)} \frac{e^{[J\mu_{2} + (r-J)\mu_{1}]t}}{J\mu_{2} + (r-J)\mu_{1}} \right] \right)$$

This implies

$$P = \exp \left[\lambda \sum_{k=1}^{\infty} \sum_{r=1}^{k} \sum_{J=0}^{r} (-1)^{2r-J} C_{k} ({}^{k}C_{r}) ({}^{r}C_{J}) \left(\frac{u\mu_{1}}{\mu_{2} - \mu_{1}} \right)^{J} v^{(r-J)} \frac{\left(e^{[J\mu_{2} + (r-J)\mu_{1}]t} - 1 \right)}{J\mu_{2} + (r-J)\mu_{1}} \right] \right]$$

Substituting the values of u and v in the above equation and simplifying, one can get the joint probability generating function of the two node communication network with bulk arrivals as

$$P(Z_{1}, Z_{2}, t) = \exp\left[\lambda \sum_{k=1}^{\infty} \sum_{r=1}^{k} \sum_{J=0}^{r} (-1)^{2r-J} C_{k}({}^{k}C_{r})({}^{r}C_{J}) \left(\frac{\mu_{1}(Z_{2}-1)}{\mu_{2}-\mu_{1}}\right)^{J} \left((Z_{1}-1) + \frac{\mu_{1}(Z_{2}-1)}{\mu_{2}-\mu_{1}}\right)^{r-J} \frac{\left(1 - e^{[J\mu_{2}+(r-J)\mu_{1}]t}\right)}{J\mu_{2} + (r-J)\mu_{1}}\right]$$
(14)

3. Performance measures using Zero Truncated Binomial Distribution for the input arrivals

Assuming that the communication system is in equilibrium t must be equal to Zero (t=0) in the joint probability generating function. For obtaining the performance of the communication network at is needed to know the functional form of the probability mass function of the number of packets that a message can be converted (C_k). Let the batch size of packets follows a zero truncated binomial distribution. Then, the probability mass function distribution of the batch size of packets in

a message is
$$C_k = \left(\frac{\frac{n!}{k!(n-k)!} \times p^k \times (1-p)^{n-k}}{1-(1-p)^n}\right), n > 0, 0 \le p \le 1$$
 for $k=1,2,...,n$.

(15) The mean number of packets in a message is $\left(\frac{n \times p}{1 - (1 - p)^n}\right)$ and its variance is $\frac{np\left[1 - p - (1 - p + np)(1 - p)^n\right]}{\left[1 - (1 - p)^n\right]^2}$. Substituting the

value of C_k in (1), we get the joint probability generating function of the number of packets in both the buffers is

$$P(Z_{1}, Z_{2}) = \exp\left[\lambda \sum_{k=1}^{n} \sum_{r=1}^{k} \sum_{J=0}^{r} (-1)^{2r-J} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}}\right) ({}^{k}C_{r}) ({}^{r}C_{J}) \left(\frac{\mu_{1}(Z_{2}-1)}{\mu_{2}-\mu_{1}}\right)^{J} \left((Z_{1}-1) + \frac{\mu_{1}(Z_{2}-1)}{\mu_{2}-\mu_{1}}\right)^{r-J} \frac{1}{J\mu_{2} + (r-J)\mu_{1}}\right]$$
(16)

The probability that the network is empty is

$$P_{00} = \exp\left[\lambda \sum_{k=1}^{n} \sum_{r=1}^{k} \sum_{j=0}^{r} (-1)^{2r} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}}\right) ({}^{k}C_{r}) ({}^{r}C_{J}) (\mu_{1})^{J} \frac{(-\mu_{2})^{r-j}}{(\mu_{2}-\mu_{1})^{r}} \left(\frac{1}{j\mu_{2}+(r-j)\mu_{1}}\right)\right]$$
(17)

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The probability generating function of the first buffer size distribution is

$$P(Z_{1}) = \exp\left[\lambda \sum_{k=1}^{n} \sum_{r=1}^{k} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}}\right)^{k} C_{r} (Z_{1}-1)^{3r} \frac{1}{r\mu_{1}}\right]$$

(18)

The probability that the first buffer is empty as

$$P_{0.} = \exp \left[\lambda \sum_{k=1}^{n} \sum_{r=1}^{k} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}} \right)^{k} C_{r} (-1)^{3r} \frac{1}{r\mu_{1}} \right]$$

(19)

The mean number of packets in the first buffer is

$$\mathbf{L}_{1} = \frac{\lambda}{\mu_{1}} \left[\sum_{k=1}^{n} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}} \right) \mathbf{k} \right]$$
(20)

The utilization of the first node is

$$U_{1} = 1 - \exp\left[\lambda \sum_{k=1}^{n} \sum_{r=1}^{k} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}}\right)^{k} C_{r} (-1)^{3r} \frac{1}{r\mu_{1}}\right]$$
(21)

The probability generating function of the second buffer size distribution is

$$P(Z_{2}) = \exp\left[\lambda \sum_{k=1}^{n} \sum_{r=1}^{k} \sum_{J=0}^{r} (-1)^{2r-J} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}}\right) ({}^{k}C_{r}) ({}^{k}C_{J}) \left(\frac{\mu_{1}(Z_{2}-1)}{\mu_{2}-\mu_{1}}\right)^{r} \frac{1}{J\mu_{2}+(r-J)\mu_{1}}\right]$$

(22)

The probability that the second buffer is empty is

$$P_{0} = \exp\left[\lambda \sum_{k=1}^{n} \sum_{r=1}^{k} \sum_{J=0}^{r} (-1)^{3r-J} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}}\right) ({}^{k}C_{r}) ({}^{r}C_{J}) \left(\frac{\mu_{1}}{\mu_{2}-\mu_{1}}\right)^{r} \frac{1}{J\mu_{2}+(r-J)\mu_{1}}\right]$$
(23)

The mean number of packets in the second buffer is

$$L_{2} = \frac{\lambda}{\mu_{2}} \left[\sum_{k=1}^{n} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}} \right) . k \right]$$

(24) The utilization of the second node is

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$$U_{2} = 1 - \exp\left[\lambda \sum_{k=1}^{n} \sum_{r=1}^{k} \sum_{J=0}^{r} (-1)^{3r-J} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}}\right) ({}^{k}C_{r}) ({}^{r}C_{J}) \left(\frac{\mu_{1}}{\mu_{2}-\mu_{1}}\right)^{r} \frac{1}{J\mu_{2}+(r-J)\mu_{1}}\right]$$
(25)

The mean number of packets in the network is

$$\mathbf{L}_{\mathrm{N}} = \mathbf{L}_{1} + \mathbf{L}_{2}$$

$$Thp_{1} = \mu_{1} \left[1 - \exp\left[\lambda \sum_{k=1}^{n} \sum_{r=1}^{k} \left(\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1 - (1-p)^{n}} \right)^{k} C_{r} (-1)^{3r} \frac{1}{r\mu_{1}} \right] \right]$$
(27)

The average delay in the first buffer is

$$W(N_{1}) = \frac{L_{1}}{Thp_{1}} = \frac{\frac{\lambda}{\mu_{1}} \left[\sum_{k=1}^{n} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}} \right) k \right]}{\mu_{1} \left[1 - \exp \left[\lambda \sum_{k=1}^{n} \sum_{r=1}^{k} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}} \right) k C_{r} (-1)^{3r} \frac{1}{r\mu_{1}} \right] \right]}$$
(28)

Throughput of the second node is

$$Thp_{2} = \mu_{2} \left[1 - \exp\left[\lambda \sum_{k=1}^{n} \sum_{r=1}^{k} \sum_{J=0}^{r} (-1)^{3r-J} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}}\right) ({}^{k}C_{r}) ({}^{r}C_{J}) \left(\frac{\mu_{1}}{\mu_{2}-\mu_{1}}\right)^{r} \frac{1}{J\mu_{2} + (r-J)\mu_{1}} \right] \right]$$

The average delay in the second buffer is

$$W(N_{2}) = \frac{L_{2}}{Thp_{2}} = \frac{\frac{\lambda}{\mu_{2}} \left[\sum_{k=1}^{n} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}} \right) .k \right]}{\mu_{2} \left[1 - \exp \left[\lambda \sum_{k=1}^{n} \sum_{r=1}^{k} \sum_{J=0}^{r} (-1)^{3r-J} \left(\frac{\frac{n!}{k!(n-k)!} \times xp^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}} \right) ({}^{k}C_{r}) ({}^{r}C_{J}) \left(\frac{\mu_{1}}{\mu_{2}-\mu_{1}} \right)^{r} \frac{1}{J\mu_{2}+(r-J)\mu_{1}} \right] \right]$$
(29)

The variance of the number of packets in the first buffer is

$$\operatorname{Var}(\mathbf{N}_{1}) = \mathbf{E}\left[\mathbf{N}_{1}^{2} - \mathbf{N}_{1}\right] + \mathbf{E}\left[\mathbf{N}_{1}\right] - \left(\mathbf{E}\left[\mathbf{N}_{1}\right]\right)^{2}$$
$$= \lambda \left[\sum_{k=1}^{n} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}}\right) \mathbf{k}(k-1) \left(\frac{1}{2\mu_{1}}\right) + \sum_{k=1}^{n} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}}\right) \mathbf{k}\left(\frac{1}{\mu_{1}}\right)\right]$$
(30)

The variance of number of packets in the second buffer is $Var(N_2) = E[N_2^2 - N_2] + E[N_2] - (E[N_2])^2$

(26)

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$$= \left\{ \frac{\lambda \times \mu_{1}}{2 \times \mu_{2}(\mu_{1} + \mu_{2})} \left[\sum_{k=1}^{\infty} \left(\frac{\frac{n!}{k!(n-k)!} \times p^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}} \right) . k(k-1) \right] + \left\{ \frac{\lambda}{\mu_{2}} \sum_{k=1}^{n} k. \left(\frac{\frac{n!}{k!(n-k)!} \times xp^{k} \times (1-p)^{n-k}}{1-(1-p)^{n}} \right) \right\}$$
(31)

The coefficient of variation of the number of packets in the first buffer is

$$\operatorname{cv}(\mathbf{N}_{1}) = \frac{\sqrt{\operatorname{Var}(\mathbf{N}_{1})}}{\mathbf{L}_{1}}$$
(32)

The coefficient of variation of the number of packets in the second buffer is

$$\operatorname{cv}(N_2) = \frac{\sqrt{\operatorname{Var}(N_2)}}{L_2}$$
(33)

PERFORMANCE EVALUATION OF THE COMMUNICATION NETWORK 5.

The performance of the proposed network is discussed through numerical data illustration. Different values of the parameters are considered for bandwidth allocation and arrival of packets. After interacting with the technical staff at the Internet service providing station, it is considered that the message arrival rate (λ) varies from 1x10⁴ messages/sec to 9x10⁴ messages/sec.

Then each message is converted into some number of packets of arbitrary size depending on the size of the message. It is assumed that the number of packets that a message can be converted into varies from 1 to 30. Hence, the number of arrivals of packets to the buffer are in batches of random size. The batch size is assumed to follow zero truncated binomial distribution with parameters (n, p).

From the equations (25), (26),(28), (30) and (32) the mean number of packets and the utilization of the network are computed for different values of n, p, λ , μ_1 , μ_2 and are given in Table1.

р	λ	μ_1	μ_2	$P_0(t)$	$\mathbf{P}_{0}(t)$	$P_{00}(t)$	Thp1	Thp2	W(N1)	W(N2)	L1	U1	L2	U2	LN
0.5	8	4	8	0.001138	0.02884	0.00316	3.95449	7.76931	2.53124	0.64419	10.00978	0.98862	5.00489	0.97116	15.0
0.5	8	4	8	0.00525	0.01072	0.00141	3.97902	7.91428	3.76989	0.94768	15.00046	0.99475	7.50023	0.98928	22.50
0.5	8	4	8	0.003	0.00518	7.89589^{*}	3.988	7.95857	5.01505	1.25651	20.00002	0.997	10.00001	0.99482	30.00
0.5	8	4	8	0.00194	0.00297	5.04975*	3.99224	7.97624	6.26214	1.56716	25	0.99806	12.5	0.99703	37.5
0.1	8	4	8	0.10933	0.30695	0.03885	3.5627	5.54442	0.068542	0.22022	2.44194	0.89067	1.20097	0.69305	3.662
0.3	8	4	8	0.06604	0.1953	0.02179	3.73583	6.43758	0.96527	0.28008	3.607	0.93396	1.80304	0.8047	5.409
0.5	8	4	8	0.03716	0.11177	0.0114	3.85135	7.105833	1.34012	0.36317	5.16129	0.96284	2.58065	0.88823	7.74
0.7	8	4	8	0.021	0.06104	0.00608	3.91598	7.51164	1.71919	0.46708	7.01705	0.979	3.50853	0.93896	10.52
0.5	1	4	8	0.66262	0.7604	0.57162	1.34953	1.91681	0.47806	0.16829	0.64516	0.33738	0.32258	0.2396	0.96
0.5	3	4	8	0.29093	0.43967	0.18672	2.83629	4.48266	0.6824	0.21589	1.93548	0.70907	0.96774	0.56033	2.903
0.5	5	4	8	0.12774	0.25422	0.06103	3.48906	5.96625	0.92455	0.2734	3.22581	0.87226	1.6129	0.74578	4.838
0.5	7	4	8	0.05608	0.14699	0.01994	3.77567	6.89407	1.19611	0.3309	4.51613	0.94392	2.25806	0.85301	6.774
0.5	8	3.0	8	0.0124	0.10471	0.00395	2.9628	7.16233	2.32271	0.36031	6.88172	0.9876	2.58065	0.89529	9.462
0.5	8	3.5	8	0.02322	0.10837	0.00725	3.41874	7.13306	1.72538	0.36179	5.89862	0.97678	2.58065	0.99163	8.479
0.5	8	4.0	8	0.03716	0.11177	0.0114	3.85135	7.10583	1.34012	0.36317	5.16129	0.96284	2.58065	0.88823	7.74
0.5	8	4.5	8	0.05358	0.11494	0.01615	4.25891	7.08044	1.07723	0.36448	4.58781	0.94642	2.58065	0.88506	7.168
0.5	8	4	6	0.03716	0.05904	0.00725	3.85135	5.64576	1.34012	0.60946	5.16129	0.96284	3.44086	0.94096	8.602
0.5	8	4	7	0.03716	0.08475	0.00941	3.85135	6.40673	1.34012	0.46035	5.16129	0.96284	2.94931	0.91525	8.110
0.5	8	4	8	0.03716	0.11177	0.0114	3.85135	7.10583	1.34012	0.36317	5.16129	0.96284	2.58065	0.88823	7.74
0.5	8	4	9	0.03716	0.13916	0.01319	3.85135	7.74754	1.34012	0.29608	5.16129	0.96284	2.29391	0.86084	7.455

Table 1. Values of mean, average delay, throughput and utilization of the network model

After node 1, the packets are forwarded to the connected buffers at the node 2, with a forward transmission rate (μ_1) of 3.0×10^4 packets/sec to 5.0×10^4 packets/sec. It is further assumed that the packets leave the second node with a transmission rate (μ_2) of $6x10^4$ packets/sec to $10x10^4$ packets/sec. In both the nodes, dynamic bandwidth allocation is considered i.e. the transmission rate of each packet depends on the number of packets in the buffer connected to it at that instant.

The relationship between mean number of packets in the buffers and in the whole network and the utilization of the nodes with respect to the input parameters is shown in Figure 2.

It is observed that after 0.1 seconds, the first buffer is having on an average of 24726 packets, after 0.3 seconds it rapidly raised to an average of 52410.4 packets. After 1 second, the first buffer is containing an average of 73626.3 packets and there after the system stabilizes and the average number of packets remains to be the same for fixed values of other parameters (5, 25, 2, 4, 8) for (n, p, λ , μ_1 , μ_2).

As the batch size distribution parameter (n) varies from $1x10^4$ packets/sec to $5x10^4$ packets/sec, the first buffer, second buffer and the network average content increase from 63809.5 packets to 73626.3 packets, 31320.4 packets to 36138.9 packets and 95129.9 packets to 109765.2 packets respectively when other parameters remain fixed. As the batch size probability distribution parameter (p) varies from $10x10^4$ packets/sec to $30x10^4$ packets/sec, the first buffer, second buffer and the network average content increase from 36813.2 packets to 85897.4 packets, 18069.5 packets to 42162.1 packets and 54882.6packets to 128059.4 packets respectively when other parameters remain fixed. As the arrival rate of messages (λ)

varies from 1×10^4 messages/sec to 3×10^4 messages/sec, the first buffer, second buffer and the network average content increase from 36813.2 packets to 110439.5 packets, 18069.5 packets to 54208.4 packets and 54882.6 packets to 164647.9 packets respectively when other parameters remain fixed at (1, 5, 25, 4, 8) for (n, p, μ_1, μ_2).



Fig 2. Relationship between Delay, Throughput and some input parameters

Similarly when the values of λ , n, and p increases, the utilization of node 1 and node 2 were also increasing. The effect of variation in the transmission rates of the nodes (μ_1 , μ_2) on the mean number of packets in the buffers and network as well on utilization of the nodes can be observed from the table1.

From the equations (28) to (29), the throughput and the average delay of the network are computed for different values of n, p, λ , μ_1 , μ_2 and are given in Table 1. The relationship between average delay and throughput of the nodes with respect to the input parameters is in figure 3. It can be observed from the table1 and figure3 that when the input parameters λ , n, p are increasing, the throughput of the nodes and average delay in both the buffers are also increasing. The influence of the transmission rates of the nodes on the performance measures can also be observed from the table1.

The values of other performance measures like the probability of emptiness of the first, second buffers and the network are also computed and analyzed. When the values of the parameters n, p and λ increase, the probabilities of emptiness of both buffers decrease. When the transmission rate of nodel (μ_1) increases, the probability of emptiness of the first buffer increase whereas the probability of emptiness of the second buffer decreases. When the transmission rate of second node (μ_2) increases, the probability of emptiness of the second node increases whereas it remains constant for node1.

Using the equations given section 4, the values for variance of number of packets in each buffer and the coefficient of variation of the number of packets in each buffer are also computed. If the variance increases then the burstness of the buffers will be high. Hence, the parameters are to be adjusted such that the variance of the buffer content in each buffer must be small. The coefficient of variation of the number of packets in each buffer will helps us to understand the consistency of the traffic flow through buffers. If coefficient of variation is large then the flow is inconsistent and the requirement to search the assignable causes of high variation. It also helps us to compare the smooth flow of packets in two or more nodes.

From this analysis it is observed that the dynamic bandwidth allocation strategy has a significant influence on all performance measures of the network. It is further observed that the performance measures are highly sensitive towards

smaller values of time. Hence, it is optimal to consider dynamic bandwidth allocation and evaluate the performance under equilibrium conditions. It is also to be observed that the congestion in buffers and delays in transmission can be reduced to a minimum level by adapting dynamic bandwidth allocation. This phenomenon has a vital bearing on quality of service (QoS).

7. CONCLUSIONS

In this paper a two node Communication network with dynamic bandwidth allocation having bulk arrivals is developed and analyzed under equilibrium condition. The dynamic bandwidth allocation strategy along with bulk arrival consideration is more suitable to the realistic situations in all types of communication networks. The novelty of this Communication network is that the arrival of packets to the initial node is in bulk with random size. The performance of the statistical multiplexing is measured by approximating the arrival process with a compound Poisson and the transmission process with Poisson process. It is observed that steady state (equilibrium) analysis of the Communication network will approximate the performance measures more close to the practical situation. This network can also be extended to the multi node communication networks. It is interesting to note that this Communication network model includes some of the earlier Communication network model given by P.S.Varma and K.Srinivasa Rao (2007), K.Nageswara Rao et.all (2011).

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