

On the Stability of Interacting Queues  
in a Multiple Access System

by

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## Abstract

We consider the standard discrete-time slotted ALOHA system with a finite number of buffered terminals. The stability (ergodicity) region for this system is known for the case of two terminals and for the case of infinite, but symmetric, terminals. In this paper we introduce a new approach of studying the stability of this system by means of a simple concept of dominance. As a result we show that the stability region for the case of two terminals can be obtained in a very simple way. Furthermore, we obtain lower (inner) bounds for the stability region of the system with an arbitrary finite number of terminals. These bounds are superior to the ones already known. Finally we point out a similarity between these stability results and the achievable region of the no-feedback collision channel that may suggest a connection between the two problems.

## 1. Introduction

Interacting queueing systems occur naturally in multiple access channel models and shared computer processor systems. As is the case with most non-standard queueing systems, these interacting queues are difficult to analyze. Their study has received much attention lately owing to their importance in applications, as well as to their theoretical interest. In [1] Fayolle and Iasnogordski displayed the inherent difficulty of the analysis of such systems. In [2] the importance of interfering queues in multiple access systems was recognized. In [3] Saadawi and Ephremides introduced an approximate model for the analysis of the slotted ALOHA system. In [4,5] Sidi and Segall introduced different approximations that led to an exact analysis of a simple 2-user system. Szpankowski in [6,7] has considered the ergodicity region of the slotted ALOHA system and obtained lower bounds. Tsybakov and Mikhailov [8] obtained sufficient or necessary conditions for the stability of this system for the case of  $M$  users and, based

on a result by Malyshev [9], were able to obtain both necessary and sufficient conditions for ergodicity. Finally, in a number of papers that have appeared in the literature since the early seventies it has been established that in the classical, bufferless, symmetric, infinite-user ALOHA system the stability threshold is  $e^{-1}$  [10,11,12].

In this paper we consider the case of discrete-time, slotted ALOHA system with  $M$  users each of which has a buffer of infinite capacity to store incoming packets. The assumptions of the model are the usual ones. Time is slotted and the transmission time of a packet is equal to one slot. Each user receives (generates) packets according to a Bernoulli process. The rate of arrivals is  $\lambda_i$  for the  $i$ th user. Arrivals to different users are independent. In each slot user  $i$  attempts to transmit the head-of-the-line packet with probability  $p_i$ , provided the buffer is not empty. Based on instantaneous ternary feedback (collision, idle, success) each user determines the outcome of the attempted transmission. Simultaneous transmission attempts by two or more users result in collision.

We are interested in determining the region of values of the arrival rates  $\lambda_i, i = 1, \dots, M$  for which this system of  $M$  queues is stable.\* We introduce a new approach in the study of the stability question. This approach consists of considering hypothetical, auxiliary systems of queues that closely parallel the operation of the system of interest but dominate it in a well-defined sense. As a result, we are able to determine the stability region for the case of  $M=2$ , that was obtained in [8], in a very simple way that illuminates the relationship and interaction between the two queues. This is done in Section 2. Furthermore, by using the same approach in a slightly more elaborate form we are able to obtain improved sufficient conditions for the stability of the system of  $M$  queues, for any finite  $M$ . This is done in stages. In Section 3 a first set of bounds is obtained from the direct extension of the dominance concept as used in Section 2 and in section 4 the more elaborate set of dominating systems is used to lead to the derivation of the final set of bounds. Finally in section 5 we relate these results to the achievable region of the collision channel without feedback [13,14]. Some of the technical proofs have been moved to appendices for the sake of clarity and continuity in the main text.

## 2. The Dominant System for the Case of 2 Queues

Let us denote by  $S$  the system of queues that is the object of our study as described in

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\* the precise definition of stability is provided in the next section

the introduction. Let  $Q_i$  denote the queue size at the  $i$ th terminal,  $i = 1, 2$ . Clearly, the queue sizes evolve as a 2-dimensional random walk in the first quadrant as shown in Fig. 1. The transition probabilities of the underlying Markov chains can be easily obtained, based on the rules of operation of the ALOHA system. In fact this chain is irreducible and aperiodic. What makes the analysis difficult is of course the infinite size of the state space and the fact that the probabilities of transition on the boundaries are different from those of their counterparts in the interior of the quadrant. This is shown, for example, in the figure for the transition  $(Q_1, Q_2) \rightarrow (Q_1 - 1, Q_2)$ . If these transition probabilities were identical, regardless of whether  $Q_2 = 0$ , the system of two queues would decompose nicely\* and each queue could be separately analyzed. This observation motivates the introduction of an auxiliary hypothetical system that has precisely this property. Consider a new system,  $S^2$ , consisting of copies of the two queues of  $S$ , with the following properties:

- 1) arrivals at queue  $i$  in the new system occur at exactly the same instants as in the original system,  $i = 1, 2$ ;
- 2) the "coin-tosses" that determine transmission attempts at queue  $i$ ,  $i = 1, 2$ , have exactly the same outcomes in both systems;
- 3) whenever  $Q_i = 0$ ,  $i = 1, 2$ , terminal  $i$  continues to transmit "dummy" packets with the same probability  $p_i$ , thus continuing its interference with the other terminal, whether it is empty or not.

It is clear that the queue sizes, at either terminal, in the new system will never be less than their counterparts in the original system, provided they start from identical initial conditions in both systems. In the new system  $S^2$  the "service" rate seen by, say, terminal 1 is always equal to  $p_1(1 - p_2)$  while in the original system it oscillates between the values of  $p_1$  and  $p_1(1 - p_2)$  depending on whether terminal 2 is empty or not.

**Definition.** The stability region of either system is the set of values of  $\lambda_i$ ,  $i = 1, 2$  for which the underlying irreducible, aperiodic Markov chain  $(Q_1, Q_2)$  is ergodic.

Clearly, the strong stochastic dominance of  $S^2$  on  $S$  implies that the stability region of  $S^2$  inner-bounds that of  $S$ . The stability region of  $S^2$  is easy to determine. Each terminal

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\* Note, however, that the queues would still be statistically dependent since the transition  $(Q_1, Q_2) \rightarrow (Q_1 - 1, Q_2 - 1)$  has zero probability.

operates as a discrete  $M | M | 1$  system and is therefore stable if and only if

$$\lambda_i < p_i \prod_{j \neq i} (1 - p_j)$$

Thus the stability region of  $S^2$  is the rectangle shown in Fig. 2. This is of course a poor bound for the stability region of  $S$  because  $S^2$  dominates the original system too strongly. We would like, to draw attention to one observation before proceeding any further. Note that each queue behaves identically in both systems so long as neither queue empties. Thus if both systems get started with identical, non-zero queue sizes, they will evolve following exactly identical trajectories at least until one of the queues empties. This observation plays a crucial role in some of the proof arguments that will follow.

Now, let us consider another hypothetical, auxiliary system  $S^1$ , in the hope of obtaining a dominant system that does not dominate the original system as strongly and that can lead, therefore, to a closer tracking of the original system's behavior and to a better bound of its ergodicity region. The system  $S^1$  is identical to both  $S$  and  $S^2$  with respect to arrivals and attempted transmissions (that is, it has properties (1) and (2)) but differs with respect to the behavior when the queues empty. In this system queue 2 behaves as in  $S$  while queue 1 behaves as in  $S^2$ ; that is, only queue 1 continues the irresponsible transmission of "dummy" packets when it empties. As a result, queue 1 sees a "service" rate that oscillates between  $p_1$  and  $p_1(1 - p_2)$  depending on whether queue 2 is empty or not, while queue 2 always sees a worst-case service rate of  $p_2(1 - p_1)$  regardless of the status of queue 1. It is clear that  $S^1$  dominates  $S$  since either queue will have a successful departure in  $S$  whenever it has one in  $S^1$ , but not necessarily vice-versa. It is also clear that in  $S^1$  the stability of queue 2 is easily determined by the fact that it operates as a discrete-time  $M | M | 1$  system. Thus it will be stable iff

$$\lambda_2 < p_2(1 - p_1) \tag{1}$$

Let us therefore assume that (1) is satisfied and let us determine the criterion for the stability of queue 1. Terminal 1 sees a "service" rate that has the value  $p_1$  when terminal 2 is empty (and that happens with probability  $1 - \frac{\lambda_2}{p_2(1 - p_1)}$ ) according to the well-known  $M | M | 1$

formula) and the value  $p_1(1-p_2)$  when terminal 2 is non-empty (which occurs with probability  $\frac{\lambda_2}{p_2(1-p_1)}$ ). In appendix it is shown that a necessary and sufficient condition for the stability of queue 1 is that the arrival rate  $\lambda_1$  be less than its average service rate, thus extending the well-known result for a  $G | G | 1$  queue [15]. Applying this result here yields

$$\lambda_1 < p_1 \left(1 - \frac{\lambda_2}{p_2(1-p_1)}\right) + p_1(1-p_2) \frac{\lambda_2}{p_2(1-p_1)}$$

which can be written as

$$\lambda_1 < p_1 \left(1 - \frac{\lambda_2}{1-p_1}\right) \tag{2}$$

Thus the stability region of  $S^1$  is as shown in Figure 3.

What is important is that the boundary of the stability region given by Eq. (2) is not only a bound for the stability region of  $S$  but coincides with it. The proof relies on the observation made earlier that, so long as the queues do not empty, systems  $S$  and  $S^1$  are indistinguishable. Here is the argument: Given that  $\lambda_2 < p_2(1-p_1)$ , if for some  $\lambda_1$  queue 1 is stable in  $S^1$ , it is also stable in  $S$  by virtue of the dominance. Conversely, if for some  $\lambda_1$  queue 1 is unstable in  $S^1$ , then the queue size  $Q_1(n)$  will grow to  $\infty$  without emptying with finite (non-zero) probability.\* Thus, not all sample paths of  $Q_1(n)$  correspond to transient behavior with infinitely often visits to 0. However, so long as queue 1 does not empty,  $S^1$  and  $S$  behave identically if started from the same initial conditions. Thus the same sample paths that go to  $\infty$  without visiting 0 belong to the evolution of queue 1 in system  $S$ . Therefore queue 1 is unstable in system  $S$  as well.

Notice that by reversing the roles of the two queues in system  $S^1$ , that is by assuming that queue 2 is the one that transmits dummy packets when it empties, we obtain a stability region given by

$$\lambda_1 < p_1(1-p_2)$$

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\* This follows from the fact that queue 2 is stable in its own right and decoupled from queue 1; nevertheless an independent proof can be supplied as in [16].

and

$$\lambda_2 < p_2 \left(1 - \frac{\lambda_1}{1 - p_2}\right) \quad (3)$$

The branch of the boundary that corresponds to Eq. (3) can be shown to be part of the stability region of  $S$  in a similar fashion. Thus the union of the branches corresponding to Eqs. (2) and (3) defines the stability region for  $S$  as shown in Fig. 4. This is identical to the region obtained by Tsybakov and Mikhailov in [8] through use of Malyshev's lemma. Note that by taking the envelope of these regions as  $p_1$  and  $p_2$  vary in  $[0, 1]$  we obtain the curve C, shown in Fig. 4, which is analytically described by

$$\sqrt{\lambda_1} + \sqrt{\lambda_2} = 1 \quad (4)$$

or parametrically, by

$$\lambda_1 = p_1(1 - p_2)$$

$$\lambda_2 = p_2(1 - p_1) \quad (5)$$

$$p_1 + p_2 = 1$$

The idea of this simple derivation as well as the choice of dominant systems was first briefly reported in [16].

### 3. The Dominant System for the case of $M$ queues

Now consider the case of  $M > 2$ . The approach described in section 2 is insufficient to yield the ergodicity region of the system. It does help however in producing bounds that are superior to the ones known to date. In this section we will develop this dominance approach and show how these bounds can be obtained.

Let

$Q_i(n)$  = queue size at terminal  $i$  at time  $n$

$D_i(n)$  = number of departures from terminal  $i$  at time  $n, i = 1, \dots, M, n = 1, 2, \dots$

and let

$$P_{i,k}^{(n)} = Pr[Q_i(n) = k] \quad (6)$$

$$\mu_i(n) = Pr[D_i(n) = 1] \quad (7)$$

Thus  $\mu_i(n)$  is the “service” rate at terminal  $i$  at time  $n$ . It clearly depends on the status of the queues at the different terminals and assumes values between  $p_i$ , when all other queues are empty and  $p_i \prod_{j \neq i} (1 - p_j)$ , when none of the other queues are empty. Let  $\tilde{\mu}_i(n)$  denote the average value of  $\mu_i(n)$ . We further define two types of events of interest:

$$\xi_k(n) = \{Q_k(n) = 0\}, \text{ i.e. the event that at time } n \text{ the } k\text{th queue is empty} \quad (8)$$

and

$$E_{k_1 k_2 \dots k_e}(n) = \bigcap_{i \in \{k_1, \dots, k_e\}} \xi_i(n) \bigcap_{i \notin \{k_1, \dots, k_e\}} \bar{\xi}_i(n) \quad (9)$$

i.e. the event that at time  $n$  the only empty queues are the ones corresponding to terminals  $k_1, \dots, k_e$ . Also note that the family of events  $E$  forms a partition of the sample space. Furthermore we have

$$\xi_k(n) = E_k(n) \bigcup_{k_1 \neq k} \{ \bigcup E_{k k_1}(n) \} \bigcup \dots \bigcup_{\substack{k_1, \dots, k_e \\ \text{all distinct}}} \{ \bigcup E_{k k_1 \dots k_e}(n) \} \bigcup \dots \bigcup E_{1,2,3,\dots,k,\dots,M}(n), \quad (10)$$

In Appendix 1 it is established that the limits of  $\mu_i(n)$  as  $n \rightarrow \infty$  exist; we denote them by  $\mu_i$  and their average values by  $\tilde{\mu}_i$ . Let us consider also the system  $S^1$ , defined in a manner similar to that of the preceding section. Namely,  $S^1$  is identical to  $S$  in all respects (arrival instants, transmission attempts, etc.) except that terminal 1 continues to transmit “dummy” packets, even when it is empty, with the same probability  $p_1$  as when it is not empty. All other



terminals behave as in  $S$ . We define the quantities  $Q^1, D^1, P^1$ , and  $\mu^1$  as above but with the use of the superscript “1” to denote reference to system  $S^1$ . Similarly we also consider for reference purposes the system  $S^M$ , in which all terminals continue to transmit when empty. This system is the generalization of  $S^2$  as defined in Section 2.

Our goal is to derive sufficient conditions for the stability of  $S^1$  and to conclude, by virtue of the dominance relationship, that these are also sufficient conditions for the stability of  $S$ . We shall focus on queue 1. To “decouple” its stability status from that of the other queues we need to assume that the other queues are stable. Unfortunately, unlike the case of  $M = 2$ , we do not know what is the stability criterion for the set of queues  $2, \dots, M$  and thus we are forced to assume a “strong” stability status for them, namely that

$$\lambda_i < \mu_i^M \triangleq p_i \prod_{j \neq i} (1 - p_j), \quad i = 2, \dots, M \quad (11)$$

The validity of Eq. (10) implies stability of this group of queues under system  $S^M$  and thus under  $S^1$  and  $S$  as well.

Let  $B_1^1(n)$  denote the event that in system  $S^1$ , terminal 1 at time  $n$  successfully transmits a packet, whether real or “dummy”; its probability is  $\mu_1^1(n)$ . We shall presently show that a sufficient condition for the stability of queue 1 in  $S^1$  is

$$\lambda_1 < \phi(\lambda_2, \dots, \lambda_M; p_1, \dots, p_M) \quad (12)$$

where  $\phi$  is a specific function of the other arrival rates and all retransmission probabilities, that lower-bounds the asymptotic service rate  $\mu_1^1$  of terminal 1. In Appendix 2 we show that this bound is already an improvement over those derived in [6-8]. We present the results in the form of two lemmas that lead to a theorem. The first lemma lower-bounds  $\mu_1^1(n)$ , the second lemma establishes the bound as  $n \rightarrow \infty$ , and the theorem establishes the sufficient condition by making use of the results of Appendix 1.

### Lemma 3.1

In the system  $S^1$ , the probability of a successful transmission by queue 1 at the instant  $n$  satisfies the following lower bound inequality for all  $n \geq 1$ :

$$\mu_1^1(n) \geq p_1 \prod_{j=2}^M (1 - p_j) + p_1 \sum_{j=2}^M Pr[\xi_j^1(n)] \cdot p_j \prod_{\substack{j=2 \\ j \neq i}}^M (1 - p_j) \quad (13)$$

**Proof**

By definition:

$$\mu_1^1(n) = Pr[B_1^1(n)] \quad (14)$$

We apply the theorem of total probability to Eq. (14) conditioning the event  $B_1^1(n)$  on the  $2^{M-1}$  mutually exclusive events of Eq. (9). This results in the following equation:

$$\begin{aligned} \mu_1^1(n) = Pr[B_1^1(n)] &= Pr[B_1^1(n)/\{E_\phi^1(n)\}] \cdot Pr[E_\phi^1(n)] + \\ &+ \sum_{k_1=2}^M Pr[B_1^1(n)/E_{k_1}^1(n)] \cdot Pr[E_{k_1}^1(n)] + \dots \\ &\dots + \sum_{\substack{k_1, \dots, k_o \in \{2, \dots, M\} \\ \text{all } k_i \text{ distinct}}} Pr[B_1^1(n)/E_{k_1, \dots, k_o}^1(n)] \cdot Pr[E_{k_1, \dots, k_o}^1(n)] + \dots \\ &+ Pr[B_1^1(n) | E_{2, \dots, M}^1(n)] Pr[E_{2, \dots, M}^1(n)] \end{aligned} \quad (15)$$

The following further decomposition is now possible.

$$\begin{aligned} &Pr[B_1^1(n)/E_{k_1, \dots, k_o}^1(n)] \\ &= Pr[B_1^1(n)/\{Q_1^1(n) > 0\}, E_{k_1, \dots, k_o}^1(n)] \cdot Pr[\{Q_1^1(n) > 0\}/E_{k_1, \dots, k_o}^1(n)] + \\ &+ Pr[B_1^1(n)/\{Q_1^1(n) = 0\}, E_{k_1, \dots, k_o}^1(n)] \cdot Pr[\{Q_1^1(n) = 0\}/E_{k_1, \dots, k_o}^1(n)] \end{aligned} \quad (16)$$

Now, if queue 1 is non-empty and of the remaining ones only queues  $k_1$  through  $k_e$  are empty then the probability of a successful transmission by terminal 1 is  $p_1 \prod_{i \neq \{k_1, \dots, k_e\}} (1 - p_i)$ . Note that in this case the transmitted packet is a real information packet. On the other hand, if queue 1 is empty and the status of the other queues is as before then the probability of a successful transmission is still given by  $p_1 \prod_{i \notin \{k_1, \dots, k_e\}} (1 - p_i)$ . The difference is that in this case the transmitted packet is a dummy packet. However, we are not concerned yet with this distinction since at this stage we seek a lower bound for the probability of successful transmission of information or dummy packets  $\mu_1^1(n)$ . Besides, near saturation, all packets tend to be information packets since, then, the queues are rarely empty.

Based on the above comments we deduce that

$$P_r[B_1^1(n)/E_{k_1, \dots, k_e}^1(n)] = p_1 \prod_{i \notin \{k_1, k_2, \dots, k_e\}} (1 - p_i) \quad (17)$$

Now, note that the following inequalities can be easily verified for any set of indices  $k_1, k_2, \dots, k_e$ :

$$\begin{aligned} 1 &= \prod_{k \in \{k_1, k_2, \dots, k_e\}} (p_k + (1 - p_k)) = (p_{k_1} + 1 - p_{k_1}) \prod_{k \in \{k_2, k_3, \dots, k_e\}} (p_k + (1 - p_k)) = \\ &= [p_{k_1} + (1 - p_{k_1})] \prod_{k \in \{k_2, k_3, \dots, k_e\}} (1 - p_k) \left(1 + \frac{p_k}{1 - p_k}\right) \geq p_{k_1} \prod_{k \in \{k_2, k_3, \dots, k_e\}} (1 - p_k) + \\ &\quad + \prod_{k \in \{k_1, k_2, \dots, k_e\}} (1 - p_k) \end{aligned} \quad (18)$$

$$1 = \prod_{k=2}^M [p_k + (1 - p_k)] \geq \prod_{k=2}^M (1 - p_k) + \sum_{k_1=2}^M p_{k_1} \prod_{\substack{k=2 \\ k \neq k_1}}^M (1 - p_k) \quad (19)$$

Using eqns. (17), (18) and (19), we rewrite Eq. (15) as follows:

$$\begin{aligned}
\mu_1^1(n) &\geq p_1 \prod_{k=2}^M (1-p_k) \cdot Pr[E_\phi^1(n)] + \\
&+ \sum_{k_1=2}^M [\{p_1 \prod_{k=2}^M (1-p_k)\} + \{p_1 p_{k_1} \prod_{\substack{k=2 \\ k \neq k_1}}^M (1-p_k)\}] \cdot Pr[E_{k_1}^1(n)] + \dots \\
\dots + &\sum_{\substack{\{k_1, k_2, \dots, k_e\} \in \{2, 3, \dots, M\} \\ \text{all } k_i \text{ distinct}}} [\{p_1 \prod_{k=2}^M (1-p_k)\} + \{p_1 p_{k_1} \prod_{\substack{k=2 \\ k \neq k_1}}^M (1-p(k))\}] \cdot Pr[E_{k_1, k_2, \dots, k_e}^1(n)] + \dots \\
\dots + &[\{p_1 \prod_{k=2}^M (1-p_k)\} + \{\sum_{k_1=2}^M p_1 p_{k_1} \prod_{\substack{k=2 \\ k \neq k_1}}^M (1-p_k)\}] Pr[E_{2,3,\dots,M}^1(n)] \tag{20}
\end{aligned}$$

Rearranging Eq. (20) and using eqs. (9) and (10) we obtain the following inequality

$$\mu_1^1(n) \geq p_1 \prod_{j=2}^M (1-p_j) + \sum_{i=2}^M p_1 \{p_i \prod_{\substack{j=2 \\ j \neq i}}^M (1-p_j)\} Pr[\xi_i^1(n)] \tag{21}$$

This concludes the derivation.

In the next result, we derive the asymptotic value of the bound of Lemma 3.1. The bound is calculated on the assumption that queues 2 through M are stable.

### Lemma 3.2

If

$$\lambda_i < \mu_i^M = p_i \prod_{j \neq i} (1-p_j), \quad i = 1, \dots, M$$

then

$$\lim_{n \rightarrow \infty} \mu_1^1(n) \geq p_1 \prod_{j=2}^M (1-p_j) + \sum_{i=2}^M p_1 p_i \left(1 - \frac{\lambda_i}{\mu_i^M}\right) \prod_{\substack{j=2 \\ j \neq i}}^M (1-p_j)$$

**Proof:**

It is sufficient to show that

$$\lim_{n \rightarrow \infty} Pr[\xi_i^1(n)] \geq 1 - \frac{\lambda_i}{\mu_i^M} \quad 2 \leq i \leq M$$

Note that by definition

$$Pr[\xi_i^1(n)] = Pr[Q_i^1(n) = 0]$$

Since  $S^M$  dominates  $S^1$  we know that

$$P[Q_i^1(n) = 0] \geq Pr[Q_i^M(n) = 0]$$

Therefore, taking limits and using the decoupling properties of  $S^M$  we obtain

$$\lim_{n \rightarrow \infty} Pr[\xi_i^1(n)] \geq \lim_{n \rightarrow \infty} Pr[Q_i^M(n) = 0] = \left(1 - \frac{\lambda_i}{\mu_i^M}\right) \quad (22)$$

Taking limits on eq. (21) and employing the inequality (22) we obtain the desired result, namely

$$\lim_{n \rightarrow \infty} \mu_1^1(n) \leq p_1 \prod_{j=2}^M (1 - p_j) + p_1 \sum_{i=2}^M \left(1 - \frac{\lambda_i}{\mu_i^M}\right) (1 - p_j) = \mu_1^1 \quad (23)$$

The existence of these limits is again justified in Appendix 1.

Now we proceed to establish the main theorem.

### Theorem 3.1

The system  $S$  is stable if

$$\lambda_1 < p_1 \prod_{j=2}^M (1 - p_j) + p_1 \sum_{j=2}^M p_j \left(1 - \frac{\lambda_j}{\mu_j^M}\right) \prod_{\substack{k=2 \\ k \neq j}}^M (1 - p_k) \quad (24)$$

and

$$\lambda_j < \mu_j^M, \quad 2 \leq j \leq M \quad (25)$$

where

$$\mu_j^M = p_j \prod_{\substack{k=1 \\ k \neq j}}^M (1 - p_k)$$

### Proof

From Lemma 3.2 we know that the R.H.S. of Eq. (24) is a lower bound for  $\mu_1^1$ , hence  $\lambda_1 < \mu_1^1$  and  $\lambda_1 < \tilde{\mu}_1^1$ . According to the results in Appendix 1, this is sufficient to ensure that

$$Pr[Q_1^1 = 0] > 0$$

The inequalities (25) are also sufficient to imply that

$$Pr[Q_i^1 = 0] > 0, \quad 2 \leq i \leq M$$

Thus under the hypothesis of the theorem the system  $S^1$  is stable and so is  $S$  since it is dominated by  $S^1$ .

### Comments

1. Recall that in System  $S^1$  we required that terminal 1 transmit dummy packets when empty; the choice of terminal was clearly arbitrary. If we interchange the roles of terminal 1 and terminal  $i$  and repeat the proof we will derive a similar result for a different ordering of the terminals. In fact the union of the regions that are obtained in this manner for all possible orderings produces a better bound.
2. Szpankowski's results on the stability of system  $S$  [6,7] are of similar nature. In appendix 2 we show that the stability region established in [6,7] is a strict subset of the region obtained here. We should like to record that Szpankowski's results apply to systems for which the arrival rates are not necessarily Bernoulli, and in that sense his results are more general. Nevertheless, we believe that our results can also be easily generalized to accommodate non-Bernoulli arrival processes because

the dominating systems  $S^1$  and  $S^M$ , crucial to our analysis, retain their essential properties for arbitrary arrival processes.

3. The bound obtained in Theorem 3.1 is not satisfactory. As shown in Fig. 5, it fails to provide values for regions of values of the  $\lambda_i$ 's that are outside the strict strong stability ones implied by Eq. (25). These regions correspond to the areas marked by question marks in Fig. 5. Also it is clearly not very tight since on the  $\lambda_i$ -axes the actual stability surface intersects at the points  $p_1, p_2, p_3$  etc. In the case of  $M = 3$ , as shown in Fig. 5, the bound is below these values by an amount equal to  $p_1 p_2 p_3$ .

#### 4. A Series of Dominant Systems that Yield Improved Bounds

In the preceding section we obtained an inner bound to the stability region of  $S$  by essentially using the same approach as in section 2 in which we were successful in determining exactly the stability region for the 2-user system. In this section we refine our approach by considering a series of dominant systems that are able to track the behavior of  $S$  a little closer than  $S^1$  could. These systems permit us to by-pass the shortcomings of the bound of Theorem 3.1 and to relax the sufficient conditions for the stability of  $S$ .

Retaining the notation we have used so far we introduce now an extended set of systems. For  $j = 1, 2, \dots, M$  we define  $S^j$  as follows:

- 1) Arrivals at the  $i$ th queue of  $S^j, i = 1, \dots, M$ , are identical to those at the  $i$ th queue of  $S$ .
- 2) "Coin tosses" that determine transmission attempts at the  $i$ th queue of  $S^j, i = 1, \dots, M$ , have identical outcomes to those at the  $i$ th queue of  $S$ .
- 3) For  $i > j$ , terminal  $i$  behaves exactly as in  $S$ , that is it does not attempt to transmit "dummy" packets when empty.
- 4) For  $i \leq j$ , terminal  $i$  attempts to transmit "dummy" packets when empty according to the following rules: with the aide of a "genie", terminal  $i$  is informed whether any terminal  $k$ , with  $k < i$ , will attempt a transmission in the slot; if yes, terminal  $i$  refrains from attempting to transmit; if no, it attempts to transmit a "dummy" packet with probability  $p_i$ .

As a result we have a series of systems in each of which terminal 1 is afforded the first choice to transmit a "dummy" packet, if empty. If it fails to transmit, "dummy" or real

packet, terminal 2, if empty, is afforded the same choice. If it doesn't transmit, terminal 3 is given the chance, and so on, up to terminal  $j$  in the system  $S^j$ . Terminals with identity number greater than  $j$  do not attempt to transmit "dummy" packets in  $S^j$ .

As a result, this series of systems has the following properties.

- i)  $S^j$  dominates  $S^{j-1}$ ,  $j = 2, \dots, M$ , and, of source,  $S^1$  dominates  $S$ . Note that  $S^1$  coincides with the system  $S^1$  considered in section 3 and so does  $S^M$  with its counterpart in that section. The reason for the dominance is this: Terminal  $j$  in  $S^{j-1}$  is always silent when empty, while it may transmit a "dummy" packet in  $S^j$ . Furthermore, all other terminals in  $S^{j-1}$  will have a successful transmission whenever they do have one in  $S^j$ , but not vice-versa. Thus, it is clear that

$$Q_i^j(n) \geq Q_i^{j-1}(n) \geq Q_i(n), \quad n = 1, 2, \dots; j = 2, \dots, M; i = 1, \dots, M$$

- ii) The joint queue sizes of the  $M$  users in each of these systems evolve according to an  $M$ -dimensional, aperiodic, irreducible Markov chain.
- iii) In system  $S^j$  the cumulative contention from terminals  $1, 2, \dots, j$  for the use of the channel remains fixed and equal to

$$1 - \prod_{i=1}^j (1 - p_i)$$

irrespective of whether any of these queues are empty. Thus any terminal  $k$ ,  $k > j$ , faces competition from the group of the first  $j$  terminals that stays always fixed. This is ensured by the rules of transmission explained above in (4).

We present the derivation of the new bounds in the form of some lemmas that lead to a theorem. We start by introducing some notation similar to that of Eqs. (8), (9), and (10).

For any set of distinct indices  $k_i$ ,  $(j + 1) \leq k_i \leq M$ , let



$$\begin{aligned}
\xi_{k_1}^j(n) &= \{Q_{k_1}^j(n) = 0\} \\
E_{k_1}^j(n) &= \{Q_{k_1}^j(n) = 0\} \bigcup_{\substack{i=1+j \\ i \neq k_1, 2}}^M \{Q_i^j(n) > 0\} \\
E_{k_1, k_2, \dots, k_e}^j(n) &= \bigcap_{i \in \{k_1, k_2, \dots, k_e\}} \{Q_i^j(n) = 0\} \bigcap_{\substack{i \in \{j+1, \dots, M\} \\ i \notin \{k_1, k_2, \dots, k_e\}}} \{Q_i^j(n) > 0\} \\
E_\phi^j(n) &= \bigcap_{i=j+1}^M \{Q_i^j(n) > 0\}
\end{aligned} \tag{26}$$

Thus  $\xi_{k_1}^j(n)$  is the event that the  $k_1$ th queue of  $S^j$  is empty at time  $n$ .  $E_{k_1}^j(n)$  is the event that only the  $k_1$ th queue of  $S^j$  in the group from  $j+1$  to  $M$  is empty at time  $n$  (all others being non-empty).  $E_{k_1, k_2, \dots, k_e}^j(n)$  is the event that only queues  $k_1, k_2, \dots, k_e$  are empty at time  $n$  and all the others in the group are non-empty. Since the group of terminals from 1 to  $j$ , collectively, present the same behavior in competing for the channel, regardless of the status of their queues, there is no interest in whether they are empty or not. Thus we consider only the events that pertain to the group  $\{j+1, \dots, M\}$ .

There are  $\binom{M-j}{1}$  events of the type  $E_{k_1}^j(n)$  and  $\binom{M-j}{e}$  events of the type  $E_{k_1, k_2, \dots, k_e}^j(n)$ . The total number of events is  $2^{M-j}$ , and they are all mutually exclusive. Their union exhausts the sample space. The event  $\xi_{k_1}^j$  may be decomposed into mutually exclusive events as follows.

$$\xi_{k_1}^j(n) = E_{k_1}^j(n) \bigcup \left\{ \bigcup_{\substack{k_2 \in \{j+1, \dots, M\} \\ k_2 \neq k_1}} E_{k_1, k_2}^j(n) \right\} \dots \bigcup E_{j+1, \dots, M}^j(n) \tag{27}$$

We now prove a lemma that generalizes lemma 3.1.

#### Lemma 4.1

In the system  $S^j$ , the probability of a successful transmission by terminal  $j$  at time  $n$  satisfies the following lower bound inequality for all  $n \geq 1$ ,

$$\mu_j^j(n) \geq p_j \prod_{\substack{i=1 \\ i \neq j}}^M (1 - p_i) + p_j \sum_{i=j+1}^M Pr[E_i^j(n)] p_i \prod_{\substack{k=1 \\ k \neq i, j}}^M (1 - p_k)$$

### Proof

By definition,

$$\mu_j^j(n) = Pr[B_j^j(n)] \quad (28)$$

where  $B_j^j(n)$  is the event that at time  $n$  terminal  $j$  has a successful transmission of a real or a “dummy” packet in system  $j$ . We apply the theorem of total probability to Eq. (28) conditioning the event  $B_j^j(n)$  on the  $2^{M-j}$  mutually exclusive events of Eq. (26). This results in the following equation.

$$\begin{aligned} Pr[B_j^j(n)] &= Pr[B_j^j(n)/\{E_\phi^j(n)\}] \cdot Pr[E_\phi^j(n)] + \sum_{k_1=j+1}^M Pr[B_j^j(n)/E_{k_1}^j(n)] \cdot Pr[E_{k_1}^j(n)] + \dots \\ &\dots + \sum_{\substack{k_1, k_2, \dots, k_e \in \{j+1, \dots, M\} \\ k_i \text{ all distinct}}} Pr[B_j^j(n)/E_{k_1, k_2, \dots, k_e}^j(n)] \cdot Pr[E_{k_1, k_2, \dots, k_e}^j(n)] + \dots \\ &+ Pr[B_j^j(n) | E_{j+1, \dots, M}^j(n)] \cdot Pr[E_{j+1, \dots, M}^j(n)] \end{aligned} \quad (29)$$

A successful transmission by terminal  $j$  is impossible if any of the stations 1 through  $j-1$  transmit. Let us define by  $T_i^j(n)$  the event that terminal  $i$  attempts a transmission in slot  $n$  in system  $j$ . Then we decompose the general term of Eq. (29) according to the following:

$$\begin{aligned} Pr[B_j^j(n)/E_{k_1, k_2, \dots, k_e}^j(n)] &= Pr[B_j^j(n)/\bigcap_{i=1}^{j-1} \{T_i^j(n)\}^c \bigcap \{Q_j^j(n) > 0\} \bigcap E_{k_1, k_2, \dots, k_e}^j(n)] \cdot \\ &\cdot Pr[\bigcap_{i=1}^{j-1} \{T_i^j(n)\}^c / \{Q_j^j(n) > 0\} \bigcap E_{k_1, \dots, k_e}^j(n)] \cdot Pr[Q_j^j(n) > 0 / E_{k_1, k_2, \dots, k_e}^j(n)] + \end{aligned}$$

$$+Pr[B_j^j(n)/\bigcap_{i=1}^{j-1}\{T_i^j(n)\}^c \cap \{Q_j^j(n) = 0\} \cap E_{k_1, k_2, \dots, k_e}^j(n)].$$

$$\cdot Pr[\bigcap_{i=1}^{j-1}\{T_i^j(n)\}^c / \{Q_j^j(n) = 0\} \cap E_{k_1, \dots, k_e}^j(n)] \cdot Pr[\{Q_j^j(n) = 0\} / E_{k_1, k_2, \dots, k_e}^j(n)] \quad (30)$$

Since the competition from the first group of  $j-1$  terminals is fixed we have

$$Pr[\bigcap_{i=1}^{j-1}\{T_i^j(n)\}^c / \{Q_j^j(n) = 0\} \cap E_{k_1, \dots, k_e}^j(n)] = \prod_{k=1}^{j-1} (1 - p_k) \quad (31)$$

The same result holds for the case of  $Q_j^j(n) > 0$ . Now let us consider the first term of the first summand of Eq. (30). The conditioning event implies that

- i) terminals 1 through  $j-1$  remain silent at time  $n$ .
- ii) terminal  $j$  has a non-empty queue and hence will independently attempt transmission of an information packet with probability  $p_j$ .
- iii) of the remaining terminals only  $k_1$  through  $k_e$  are empty and hence they will remain silent (according to the rules of system  $S^j$ ).

Based on this we deduce that

$$Pr[B_j^j(n)/\bigcap_{i=1}^{j-1}\{T_i^j(n)\}^c \cap \{Q_j^j(n) > 0\} \cap E_{k_1, k_2, \dots, k_e}^j(n)]$$

$$= p_j \prod_{\substack{i \in \{j+1, \dots, M\} \\ i \notin \{k_1, k_2, \dots, k_e\}}} (1 - p_i) \quad (32)$$

Consider now the first term of the second summand of Eq. (30). The conditioning event implies that

- i) terminals 1 through  $j-1$  remain silent at time  $n$ .

- ii) terminal  $j$  has an empty queue; therefore, by the rules of in  $S^j$ , it will broadcast a dummy packet with probability  $p_j$ .
- iii) Of the remaining terminals only  $k_1$  through  $k_e$  are empty and hence they will remain silent. Based on this we have

$$Pr[B_j^j(n) / \bigcap_{i=1}^{j-1} \{T_i^j(n)\}^c \cap \{Q_j^j(n) = 0\} \cap E_{k_1, \dots, k_e}^j(n)] = p_j \prod_{\substack{i=j+1, \dots, M \\ i \notin \{k_1, k_2, \dots, k_e\}}} (1 - p_i) \quad (33)$$

Note that when the  $j^{\text{th}}$  queue is empty terminal  $j$  broadcasts a dummy packet, whereas when the queue is non-empty it broadcasts an information packet. However, this distinction is again of no concern since we seek a lower bound for the probability of successful transmission of information or dummy packets, and since the distinction becomes meaningless near saturation.

Substituting Eqs. (31), (32) and (33) in Eq. (30) we obtain

$$Pr[B_j^j(n) / E_{k_1, \dots, k_e}^j(n)] = p_j \prod_{\substack{i \in \{k_1, \dots, k_e\} \\ i \neq j}} (1 - p_i) \quad (34)$$

Also note that

$$Pr[B_j^j(n) / E_\phi^j(n)] = p_j \prod_{\substack{1 \leq i \leq M \\ i \neq j}} (1 - p_i) \quad (35)$$

Note that the special nature of  $S^j$  enabled us to obtain eqns. (34) and (35) without having to estimate  $Pr\{Q_j^j(n) > 0 / E_{k_1, \dots, k_e}^j(n)\}$  or  $Pr\{Q_j^j(n) = 0 / E_{k_1, \dots, k_e}^j(n)\}$ , which would not be possible.

Now we invoke the inequality (18) and a slightly modified version of inequality (19), namely,

$$1 = \prod_{k=j+1}^M [p_k + (1 - p_k)] \geq \prod_{k=j+1}^M (1 - p_k) + \sum_{k_1=j+1}^M p_{k_1} \prod_{\substack{k=j+1 \\ k \neq k_1}}^M (1 - p_k) \quad (36)$$

Substituting eqns. (34), (35) and using inequalities (18) and (36) in eqn. (29) we obtain

$$\begin{aligned}
\mu_j^j(n) &\geq p_j \prod_{\substack{k=1 \\ k \neq j}}^M (1-p_k) \cdot Pr[E_\phi^j(n)] + \sum_{k_1=j+1}^M \{p_j \prod_{\substack{k=1 \\ k \neq j}}^M (1-p_k) + p_j p_{k_1} \prod_{\substack{k=1 \\ k \neq k_1, j}}^M (1-p_k)\} Pr[E_{k_1}^j(n)] + \\
&+ \sum_{\substack{\{k_1, k_2, \dots, k_s\} \subset \{j+1, \dots, M\} \\ \text{all distinct}}} ((p_j \prod_{\substack{k=1 \\ k \neq j}}^M (1-p_k)) + \{p_j p_{k_1} \prod_{\substack{k=1 \\ k \neq k_1, j}}^M (1-p_k)\}) \cdot Pr[E_{k_1, k_2, \dots, k_s}^j(n)] + \dots \\
&\dots + ((p_j \prod_{\substack{k=1 \\ k \neq j}}^M (1-p_k)) + \{ \sum_{k_1=j+1}^M p_j p_{k_1} \prod_{\substack{k=1 \\ k \neq k_1, j}}^M (1-p_k) \}) \cdot Pr[E_{j+1, \dots, M}^j(n)] \quad (37)
\end{aligned}$$

or, by further rearranging and using Eq. (27),

$$\mu_j^j(n) \geq p_j \prod_{\substack{j=1 \\ i \neq j}}^M (1-p_i) + \sum_{i=j+1}^M p_j \{p_i \prod_{\substack{k=1 \\ k \neq i, j}}^M (1-p_k)\} Pr[\xi_i^j(n)] \quad (38)$$

This concludes the derivation.

By letting  $n \rightarrow \infty$  in Eq. (38) we obtain the following asymptotic bound.

$$\tilde{\mu}_j^j > \mu_j^j = \lim_{n \rightarrow \infty} \mu_j^j(n) \geq p_j \prod_{\substack{i=1 \\ i \neq j}}^M \bar{p}_i + \sum_{i=j+1}^M (p_j p_i \prod_{\substack{k=1 \\ k \neq i, j}}^M \bar{p}_k) Pr[Q_i^j = 0] \quad (39)$$

It is clear now that lower bounds for  $\mu_j^j$  can be found if lower bounds for  $Pr[Q_i^j = 0]$  can be determined. The latter are derived in the next lemma by examination of the departure statistics at queue  $i$  in  $S^j$  for  $j < i$ , and assuming that  $\lambda_i < \tilde{\mu}_i^i$ , for  $i > j$ , where  $\tilde{\mu}_i^i$  is the average asymptotic service rate of queue  $i$  in system  $S^i$ .

#### Lemma 4.2

If  $\lambda_i < \tilde{\mu}_i^i$  for all  $i$ , then in steady state  $Pr[Q_i^j > 0] \leq \frac{\lambda_i}{\tilde{\mu}_i^i}$  for any  $j < i$ . \*

\* This result is similar to the one for the  $G | G | 1$  queue, see Kleinrock [15], with average arrival rate  $\lambda$  and average service time  $\bar{x}$ . If the queue is stable for  $\lambda < \frac{1}{\bar{x}}$ , then  $Pr[Q > 0] = \lambda \bar{x}$ .

**Proof:**

Let  $X_i^j(n)$  denote the number of packets in the  $i$ th queue of  $S^j$  after the  $n^{\text{th}}$  departure.

Let  $A_i^j(n+1)$  denote the number of packets that arrive at queue  $i$  of  $S^j$  during the  $(n+1)^{\text{st}}$  service interval. Then:

$$X_i^j(n+1) = X_i^j(n) + A_i^j(n+1) - 1[X_i^j(n)] \quad (40)$$

where

$$1[x] = \begin{cases} 1, & x > 0 \\ 0, & \text{else} \end{cases}$$

We know from Appendix 1 that if  $\lambda_i < \tilde{\mu}_i^j$  then the  $i$ th queue of  $S^j$  is stable. Hence there exists a steady state distribution for the system statistics. Furthermore, the results of Appendix 1 imply that the average queue sizes are finite in that case. Taking expectations in Eq. (40) with respect to this distribution we obtain the following:

$$\begin{aligned} E[X_i^j(n+1)] &= E[X_i^j(n)] + E[A_i^j(n+1)] - Pr[X_i^j(n) > 0] = \\ &= E[X_i^j(n)] + \lambda_i E[\text{length of } (n+1)^{\text{th}} \text{ service interval for queue } i] \\ &\quad - Pr[X_i^j(n) > 0] \end{aligned} \quad (41)$$

Now, in steady state,  $E[X_i^j(n+1)] = E[X_i^j(n)]$ ; hence

$$Pr[X_i^j > 0] = \lambda_i E[\text{length of service interval of queue } i \text{ in } S^j]$$

Since for  $\lambda_i < \tilde{\mu}_i^j$  the  $i^{\text{th}}$  queue is stable in  $S^j$ , we have

$$1 \geq Pr[X_i^j > 0] = \lambda_i f^j(\lambda_i) \quad (42)$$

where  $f^j(\lambda_i)$  denotes the expected length of the service interval of queue  $i$  in  $S^j$ . Obviously this service time is also a function of the other arrival rates  $\lambda_k, k > j$ , but not of the ones for  $k \leq j$ , since arrivals to the first  $j$  queues have no bearing on how often there are transmission attempts from the group of the first  $j$  queues in  $S^j$  (they occur with probability  $1 - \prod_{k=1}^j (1 - p_k)$  in every slot). In any event, the function  $f^j$  is non-decreasing in  $\lambda_i$ ; this is quite obvious since more frequent arrivals will create more frequent collisions and less frequent successes by all queues, including the  $i^{\text{th}}$  one. Thus, for  $\lambda_i < \tilde{\mu}_i^j$ ,

$$f^j(\lambda_i) \leq f^j(\tilde{\mu}_i^j) \quad (43)$$

Using inequality (43) in Eq. (42) we obtain

$$1 \geq \tilde{\mu}_i^j f^j(\tilde{\mu}_i^j)$$

or

$$f^j(\tilde{\mu}_i^j) \leq \frac{1}{\tilde{\mu}_i^j}$$

Hence

$$Pr[X_i^j > 0] = \lambda_i f^j(\lambda_i) \leq \lambda_i f^j(\tilde{\mu}_i^j) \leq \frac{\lambda_i}{\tilde{\mu}_i^j} \quad (44)$$

Now we argue that

$$Pr[X_i^j > 0] = Pr[Q_i^j > 0] \quad (45)$$

Recall that  $Q_i^j$  is the steady state size of queue  $i$  in  $S^j$  at any arbitrary instant in time, whereas  $X_i^j$  is the steady state queue size at any arbitrary departure point in time. They could conceivably be different random variables. However since the joint queue size  $[Q_i^j(n) \dots Q_n^j(n)]$  is an irreducible aperiodic Markov Chain, ergodicity implies that eq. (45) is valid.

Finally, by virtue of the dominance of  $S^j$  by  $S^{j+1}$  we conclude that

$$\tilde{\mu}_i^j \geq \tilde{\mu}_i^{j+1}, \quad 1 \leq j \leq M-1$$

Consequently

$$Pr[Q_i^j > 0] \leq \frac{\lambda_i}{\tilde{\mu}_i^j} \leq \frac{\lambda_i}{\tilde{\mu}_i^i}, \quad j+1 \leq i \leq M$$

This completes the proof.

Now we state the main theorem of this section.

**Theorem 4.1**

The system S is stable if

$$\lambda_j < b_j^j, \quad 1 \leq j \leq M$$

where

$$b_j^j \triangleq p_j \prod_{\substack{i=1 \\ i \neq j}}^M (1 - p_i) + \sum_{i=j+1}^M p_j \left(1 - \frac{\lambda_i}{b_i^i}\right) p_i \prod_{\substack{k=1 \\ k \neq i, j}}^M (1 - p_k) \quad (46)$$

and

$$b_M^M \triangleq \mu_M^M = p_M \prod_{i=1}^{M-1} (1 - p_i) \quad (47)$$

**Proof**

By lemma 4.2 we know that if  $\lambda_i < \tilde{\mu}_i^i$  we have

$$Pr[Q_i^j = 0] \geq 1 - \frac{\lambda_i}{\tilde{\mu}_i^i}, \quad i > j$$

Substituting this in Eq (39) we obtain



$$\tilde{\mu}_j^j \geq p_j \prod_{\substack{i=1 \\ i \neq j}} \bar{p}_i + \sum_{i=j+1}^M (p_j p_i \prod_{\substack{k=1 \\ k \neq i, j}} \bar{p}_k) (1 - \frac{\lambda_i}{\tilde{\mu}_i^i}) \quad (48)$$

Note that the  $b_j^j$ 's, as defined in Eqs (46), (47), are less than or equal to the  $\tilde{\mu}_j^j$ 's. Therefore substituting the  $b_j^j$ 's in the RHS of inequality (48) we strengthen the bound. Thus if  $\lambda_j < b_j^j$ , it is also true that  $\lambda_j < \tilde{\mu}_j^j$  and, hence, the queue is stable in  $S^j$  and, by the dominance property, in  $S$  as well. This completes the proof.

Notes:

1. This is an improved bound as easily seen from the comparison of  $b_1^1$  to the value of the RHS of Eq. (24). In fact, it is easily verified that  $b_j^j > \mu_j^M$ .
2. The ordering of the terminals has been arbitrary, but the bound depends on that ordering. The union of the bounds of the stability regions over all possible orderings yields an improved bound.

## 5. The Connection to Information Theory

The results obtained so far for the stability of the slotted-ALOHA system present a striking similarity to those concerning the achievable region of the no-feedback collision channel [13,14]. First of all let us refer to the problem we studied here as the "stability" problem and to that considered in [13,14] as the "capacity" problem. Let us discuss their similarities and differences:

**"Stability" problem** - Determine the regions of values of the arrival rates  $\lambda_i$  of the Bernoulli arrival processes for which the discrete-time, slotted-ALOHA system of  $M$  buffered users is stable in the sense of ergodicity of the underlying chain, or finiteness of the average queue size (the two are equivalent here); that is, find the values of  $\lambda_i$  for which  $\exists$  values of  $p_i$ , such that the system is stable. There is no concern about Shannon capacities, and feedback is assumed to be provided to the terminals.

**"Capacity" problem** - The object of study is the same; however, a different question is asked about it, namely at what rates  $R_i$  can the terminals simultaneously transmit to the common receiver reliably (in the Shannon sense). There is no concern about stability of queues; in fact the queues may very well be assumed to be infinite in order to provide an inexhaustible supply of packets, when needed, for transmission. There is no feedback, and the terminals are

not constrained to use the ALOHA protocol for transmission, but only some (any) protocol sequence of packet transmission and retransmission.

To summarize:

Both problems consider the same channel (the collision channel), that requires a single packet transmission to ensure successful reception. However they differ in terms of the question of interest (queue stability vs. reliable transmission rate), and in terms of the assumptions on feedback, arrivals, and transmission protocol.

Interestingly the results are remarkably similar for the two problems. In fact,

- 1) for  $M = \infty$  (and  $\lambda_1 = \lambda_2 = \dots$ ) the “stability” problem requires that  $\lim_{\substack{M \rightarrow \infty \\ \lambda_i \rightarrow 0}} M\lambda_i < e^{-1}$  and the “capacity” problem states that  $\lim_{\substack{M \rightarrow \infty \\ \lambda_i \rightarrow 0}} MR_i < e^{-1}$  is achievable.
- 2) for  $M = 2$  the “stability” problem, as shown here and in [8], requires that

$$\sqrt{\lambda_1} + \sqrt{\lambda_2} \leq 1$$

and the “capacity” problem as shown in [13,14] states that

$$\sqrt{R_1} + \sqrt{R_2} \leq 1$$

defines an achievable region.

- 3) for  $2 < M < \infty$  the “stability” problem leads to the bounds derived here and elsewhere [6,7]. These bounds can be shown to be consistent with the results of the “capacity” problem, which state that

$$R_i = p_i \prod_{j \neq i} (1 - p_j)$$

$$\sum_{i=1}^M p_i = 1$$

defines an achievable region (capacity region, actually, under the constraint of using a transmission protocol sequence); namely that region contains the region of inner bounds obtained for the stability problem (as can be shown quite easily).

These observations lead to the following two questions.

- 1) Is there a non-superficial connection between the informationtheoretic and the queueing theoretic versions of the problem, or is the identity of the results a mere coincidence?
- 2) In either case, is the conjecture that for  $2 < M < \infty$  the stability region is given by

$$\lambda_i = p_i \prod_{j \neq i} (1 - p_j)$$

$$\sum_{i=1}^M p_i = 1$$

true?

Whatever the answers to these questions we may close with two, hopefully, useful remarks.

- i) Feedback is known to, in general, enlarge the capacity region of a multi-user channel; the results seem to imply that, at least for  $M=2$ , the ALOHA transmission protocol makes poor use of this feedback in that the achievable region is not enlarged, but, rather, stays puzzlingly the same.
- ii) The meaning of the requirement that  $\sum_{i=1}^M p_i = 1$  in the parametric description of the capacity region remains unclear, in connection with the role of the  $p_i$ 's in the transmission protocol of the "stability" problem. The intriguing nature of the sum of the values of the  $p_i$ 's was first noticed by Szpankowski in [6,7]. Here we simply note that for the "stability" problem one can easily show that if

$$\sum_{i=1}^M p_i < 1$$

all users can improve their delay (or queue-size) performance if they uniformly increase their respective  $p_i$ 's by an amount given by

$$\Delta p_i = p_i(1 - p_i) \cdot \Delta t \tag{49}$$

where  $\Delta t$  is a common auxiliary variable, until the sum hits the value 1. Specifically one can show that any increment in the values of the  $p_i$ 's according to Eq. (49) produces a system dominated by the previous one in the dominance sense that was used throughout this paper. We do not know how to show the improvement by decreasing the  $p_i$ 's when  $\sum_{i=1}^M p_i > 1$ .

## 6. Conclusions

In this paper we have used a new technique involving stochastic dominance to study the ergodicity region of the discrete-time slotted-ALOHA system with a finite number of buffered users. We have obtained the region exactly for  $M=2$  and derived inner bounds to it for  $M > 2$ , that improve upon earlier ones. The results look remarkably similar to those of the no-feedback collision channel and thus lead to speculation about a deeper relationship between the queueing-theoretic and information-theoretic approaches to the problem. They also suggest a conjecture about the precise boundaries of the stability region for  $M > 2$ . The dominance approach introduced in this paper represents a natural and simple concept that may be of further value to the study of other related problems.

## Appendix 1

In this appendix we establish the dynamics of the  $i$ th queue in system  $S^j$ , as defined in the general formulation of section 4, and obtain the stability criterion that is used repeatedly in all sections of the paper. Specifically the objective is to show that in  $S^j$ , if  $\lambda_i < \bar{\mu}_i^j$ , the  $i$ th queue is stable in the sense that

$$P_{i,0}^j = Pr[Q_i^j = 0] > 0.$$

Consider the  $i$ th queue of  $S^j$  at time  $n$  and the system equations for the queue-size probabilities. These are

$$P_{i,0}^j(n) = \bar{\lambda}_i P_{i,0}^j(n-1) + \bar{\lambda}_i \mu_{i,1}^j(n-1) P_{i,1}^j(n-1) \quad (A1.1)$$

$$\begin{aligned} P_{i,1}^j(n) &= [\bar{\lambda}_i \bar{\mu}_{i,1}^j(n-1) + \lambda_i \mu_{i,1}^j(n-1)] P_{i,1}^j(n-1) \\ &\quad + \lambda_i P_{i,0}^j(n-1) \\ &\quad + \bar{\lambda}_i \mu_{i,1}^j(n-1) P_{i,2}^j(n-1) \end{aligned} \quad (A1.2)$$

$$\begin{aligned} P_{i,k}^j(n) &= \bar{\lambda}_i \bar{\mu}_{i,k-1}^j(n-1) P_{i,k-1}^j(n-1) \\ &\quad + \bar{\lambda}_i \bar{\mu}_{i,k}^j(n-1) + \lambda_i \mu_{i,k}^j(n-1) P_{i,k}^j(n-1) \\ &\quad + \bar{\lambda}_i \mu_{i,k+1}^j(n-1) P_{i,k+1}^j(n-1) \end{aligned} \quad (A1.3)$$

Obviously the “death-rates”  $\mu_{i,k}^j(n)$  are functions of the status of the other queues, which is why the queues are coupled and difficult to analyze. However, the above equations are valid conditioned on the status of the other queues. There is, in our notation, an explicit dependence shown on the size of the  $i$ th queue. Thus

$$\mu_{i,k}^j(n) = Pr[D_i^j(n) = 1 \mid Q_i^j(n) = k]$$

Suppose now that  $P_{i,k}^j(n)$  and  $\mu_{i,k}^j(n)$  possess limits as  $n \rightarrow \infty$ ; that they indeed do will be established a little later. By considering the limits of the system equations we obtain

$$P_{i,0}^j = P_{i,0}^j \bar{\lambda}_i + \bar{\lambda}_i \mu_{i,1}^j P_{i,1}^j \quad (\text{A1.4})$$

$$\begin{aligned} P_{i,1}^j &= \lambda_i P_{i,0}^j \\ &+ \bar{\lambda}_i \bar{\mu}_{i,1}^{-j} + \lambda_i \mu_{i,1}^j P_{i,1}^j \\ &+ \bar{\lambda}_i \mu_{i,2}^j P_{i,2}^j \end{aligned} \quad (\text{A1.5})$$

$$\begin{aligned} P_{i,k}^j &= \lambda_i \bar{\mu}_{i,k-1}^j P_{i,k-1}^j \\ &+ \bar{\lambda}_i \mu_{i,k+1}^j P_{i,k+1}^j \end{aligned} \quad (\text{A1.6})$$

These equations can be solved to yield

$$P_{i,1}^j = P_{i,0}^j \frac{\lambda_i}{\bar{\lambda}_i \mu_{i,1}^j} \quad (\text{A1.7})$$

$$P_{i,k}^j = P_{i,0}^j \frac{\lambda_i^k}{\bar{\lambda}_i^k} \frac{1}{\mu_{i,k}^j} \prod_{\ell=1}^{k-1} \frac{\bar{\mu}_{i,\ell}^j}{\mu_{i,\ell}^j} \quad (\text{A1.8})$$

and hence

$$P_{i,0}^j = \frac{1}{1 + \frac{\lambda_i}{\bar{\lambda}_i \mu_{i,1}^j} + \sum_{k=2}^{\infty} \frac{\lambda_i^k}{\bar{\lambda}_i^k} \frac{1}{\mu_{i,k}^j} \prod_{\ell=1}^{k-1} \left( \frac{\bar{\mu}_{i,\ell}^j \lambda_i}{\mu_{i,\ell}^j \bar{\lambda}_i} \right)} \quad (\text{A1.9})$$

Based on Eq. (A1.9) we conclude that:

1. Queue i is stable in the sense that

$$Pr[Q_i^j = 0] = P_{i,0}^j > 0$$

if

$$\lambda_i < \lim_{k \rightarrow \infty} \mu_{i,k}^j \quad (\text{A1.10})$$

2. Queue  $i$  is unstable in the sense that

$$Pr[Q_i^j = \infty] = \lim_{k \rightarrow \infty} P_{i,k}^j = 1$$

if

$$\lambda_i > \lim_{k \rightarrow \infty} \mu_{i,k}^j \quad (\text{A1.11})$$

Note also that Eqs. (A1.8) and (A1.9) imply that under the stability condition the moments of the queue sizes exist as well.

We now wish to show that  $\lambda_i < \tilde{\mu}_i^j$  implies stability. If  $\lambda_i < \tilde{\mu}_i^j$ , one of the following is true:

$$\text{Case 1 - } \lambda_i < \tilde{\mu}_i^j \leq \lim_{k \rightarrow \infty} \mu_{i,k}^j$$

$$\text{Case 2 - } \lambda_i < \lim_{k \rightarrow \infty} \mu_{i,k}^j \leq \tilde{\mu}_i^j$$

$$\text{Case 3 - } \lambda_i = \lim_{k \rightarrow \infty} \mu_{i,k}^j < \tilde{\mu}_i^j$$

$$\text{Case 4 - } \lim_{k \rightarrow \infty} \mu_{i,k}^j < \lambda_i < \tilde{\mu}_i^j$$

Cases 1 and 2 satisfy Eq. (10) and hence the queue is stable.

Case 4 cannot occur, because if it does then by Eq. (A1.11)  $\lim_{k \rightarrow \infty} P_{i,k}^j = 1$ , which implies that

$$\tilde{\mu}_i^j = \sum_{k=0}^{\infty} \mu_{i,k}^j P_{i,k}^j = \lim_{k \rightarrow \infty} \mu_{i,k}^j$$

which contradicts the inequality

$$\lim_{k \rightarrow \infty} \mu_{i,k}^j < \tilde{\mu}_i^j.$$

Under Case 3 it is not immediately clear whether or not queue  $i$  is stable. Suppose it is not stable, then using Eq. (A1.11) we obtain

$$\tilde{\mu}_i^j = \sum_{k=0}^{\infty} \mu_{i,k}^j P_{i,k}^j = \lim_{k \rightarrow \infty} \mu_{i,k}^j$$

which contradicts the inequality

$$\lim_{k \rightarrow \infty} \mu_{i,k}^j < \tilde{\mu}_i^j$$

In conclusion,  $\lambda_i < \tilde{\mu}_i^j$  is sufficient for queue  $i$  to be stable, in system  $S^j$ , in the sense that

$$Pr[Q_i^j = 0] > 0$$

It remains for us to justify the existence of the limits:

$$\lim_{n \rightarrow \infty} P_{i,k}^j(n) = P_{i,k}^j$$

$$\lim_{n \rightarrow \infty} \mu_{i,k}^j(n) = \mu_{i,k}^j$$

The joint queue size in  $S^j$  is an aperiodic irreducible Markov Chain on a countable state space. Such a chain is either ergodic in which case

$$\lim_{n \rightarrow \infty} Pr\left[\bigcap_{i \in I} \{Q_i^j(n) = k_i\}\right] = Pr\left[\bigcap_{i \in I} \{Q_i^j = k_i\}\right] > 0$$

for any index set  $I \subset \{1, 2, \dots, M\}$  and any set of non-negative integers  $k_i$ , or the chain is not ergodic in which case

$$\lim_{n \rightarrow \infty} Pr\left[\bigcap_{i \in I} \{Q_i^j(n) = k_i\}\right] = 0$$

for any index set  $I \subset \{1, 2, \dots, M\}$  and any set of non-negative integers  $k_i$ .



The existence of the limit of  $P_{i,k}^j(n)$  is a direct consequence of this. We shall now establish the existence of the limit of  $\mu_{i,k}^j(n)$ . At no loss of generality we shall do so for  $j = 1$ . Let us use the notation introduced in sections 3 and 4 concerning the events that involve emptiness of the queues. Namely, for any set of distinct indices  $k_i, 2 \leq k_i \leq M$ , let

$$\xi_{k_1}^1(n) = \{Q_{k_1}^1(n) = 0\} \quad (\text{A1.12})$$

$$E_{k_1}^1(n) = \{Q_{k_1}^1(n) = 0\} \bigcap_{\substack{i=1 \\ i \neq k_1}}^M \{Q_i^1(n) > 0\} \quad (\text{A1.13})$$

$$E_{k_1, k_2, \dots, k_e}^1(n) = \bigcap_{i \in \{k_1, \dots, k_e\}} \{Q_i^1(n) = 0\} \bigcap_{\substack{i \in \{1, 2, \dots, M\} \\ i \notin \{k_1, k_2, \dots, k_e\}}} \{Q_i^1(n) > 0\}$$

and

$$E_\phi^1(n) = \bigcap_{i=2}^M \{Q_i^1(n) > 0\} \quad (\text{A1.14})$$

Thus  $\xi_{k_1}^1(n)$  is the event that the  $k_1$ th queue of only the  $k_1$ th queue of  $S^1$  is empty at time  $n$  (all others being non empty).  $E_{k_1, k_2, \dots, k_e}^1(n)$  is the event that only queues  $k_1, k_2, \dots, k_e$  are empty at time  $n$  and all the others are non-empty.

There are  $\binom{M-1}{1}$  events of the type  $E_{k_1}^1(n)$  and  $\binom{M-1}{e}$  events of the type  $E_{k_1, k_2, \dots, k_e}^1(n)$ . The total number of events is  $2^{M-1}$ , and they are all mutually exclusive. Their union exhausts the sample space

Now the following decomposition is possible for  $\mu_{i,j}^1(n)$ .

$$\begin{aligned} \mu_{i,j}^1(n) &= Pr[D_i^1(n) = 1/Q_i^1(n) = j] = Pr[D_i^1(n) = 1/Q_i^1(n) = j, E_\phi^1(n)] \cdot Pr[E_\phi^1(n)/Q_i^1(n) = j] + \\ &+ \sum_{k_1=2}^M Pr[D_i^1(n) = 1/Q_i^1(n) = j, E_{k_1}^1(n)] \cdot Pr[E_{k_1}^1(n)/Q_i^1(n) = j] + \dots \end{aligned}$$

$$\begin{aligned}
& \dots + \sum_{\substack{\{k_1, \dots, k_e\} \in \{2, 3, \dots, M\} \\ \text{all } k_i \text{ distinct}}} Pr[D_i^1(n) = 1 / \{Q_i^1(n) = j\}, E_{k_1, \dots, k_e}^1(n)] \cdot Pr[E_{k_1, \dots, k_e}^1(n) / Q_i^1(n) = j] + \\
& \dots + Pr[D_i^1(n) = 1 / \{Q_i^1(n) = j\}, E_{2, \dots, M}^1(n)] \cdot Pr[E_{2, \dots, M}^1(n) / Q_i^1(n) = j] \quad (A1.20)
\end{aligned}$$

Each of the  $2^{M-1}$  terms has a limit as  $n \rightarrow \infty$  since

$$Pr[D_i^1(n) = 1 / Q_i^1(n) = j, E_{k_1, \dots, k_e}^1(n)]$$

$$= p_1 \prod_{\substack{k=2 \\ k \neq \{k_1, \dots, k_e\}}}^M \bar{p}_k$$

which is not a function of  $n$ , and since the limit of

$$\begin{aligned}
& Pr[E_{k_1, \dots, k_e}^1(n) / Q_i^1(n) = j] \\
& = \frac{Pr\{E_{k_1, \dots, k_e}^1(n), Q_i^1(n) = j\}}{Pr\{Q_i^1(n) = j\}}
\end{aligned}$$

exists as  $n \rightarrow \infty$ , because as outlined before, the joint queue size Markov Chain is aperiodic and irreducible.

It can be shown in a similar manner that the limits of  $\mu_{i,k}^j$  exist for any other value of the index  $j$ . The proof depends only on the Markovian nature of the total system and thus it applies to any of the systems  $S^j$  including  $S$ .

## Appendix 2

In this appendix we show that for the system  $S$ , the stability region obtained by Szpankowski in [6,7] is a strict subset of the region obtained here. First we describe briefly the model used in [6,7] and state the results using our notation. Then we compare the two bounds.

The model employed in [6,7] is identical to ours except in one respect. The arrival processes are not assumed to be Bernoulli but, rather, more general processes with means  $\lambda_i$  and variances  $\mu_i^2 < \infty$ .

Let

$$A_n \triangleq p_n \prod_{\substack{k=1 \\ k \neq n}}^M (1 - p_k) - \lambda_n \quad (\text{A2.1})$$

and

$$d_n(m) \triangleq A_n \bar{p}_n + A_m p_n \quad (\text{A2.2})$$

It is shown in [6,7] that the system is stable if for every pair  $(n, m)$ ,  $1 \leq n, m \leq M$ , the following hold;

$$\text{i) } d_n(m) > 0 \text{ and } d_m(n) > 0, \text{ if } p_n + p_m \leq 1 \quad (\text{A2.3})$$

$$\text{ii) } d_n(m) > 0 \text{ or } d_m(n) > 0, \text{ if } p_n + p_m > 1. \quad (\text{A2.4})$$

We can see from (A2.3) and (A2.4) that at most one of the  $A_n$ 's can be negative. If a pair of them, say  $A_n$  and  $A_m$ , were negative then neither A2.3 nor A2.4 would be satisfied, since  $d_n(m)$  and  $d_m(n)$  would both be negative. So without loss of generality let us assume that only  $A_1$  is negative.

Since all  $A_n$ 's except  $A_1$  are positive, both A2.3 and A2.4 will be satisfied for all  $m, n$  for which neither  $n$  nor  $m$  equals 1. Thus, A2.3 and A2.4 are equivalent to the following:

$$\text{i) } A_n > 0, \quad 2 \leq n \leq M \quad (\text{A2.5})$$

and

$$\text{ii) } d_1(m) > 0 \text{ and } d_m(1) > 0, \quad \text{if } p_1 + p_m \leq 1 \quad (\text{A2.6})$$

$$d_1(m) > 0 \text{ or } d_m(1) > 0, \quad \text{if } p_1 + p_m > 1 \quad (\text{A2.7})$$

Using our notation the above questions may be rewritten as follows after some rearrangement

$$\text{i) } \lambda_n < p_n \prod_{\substack{k=1 \\ k \neq n}}^M (1 - p_k), \quad 2 \leq n \leq M \quad (\text{A2.8})$$

and

ii) if  $p_1 + p_m \leq 1$ :

$$\lambda_1 < \min \left\{ p_1 \prod_{k=2}^M (1 - p_k) + (1 - p_m) \prod_{\substack{k=1 \\ k \neq m}}^M (1 - p_k) - \lambda_m \frac{(1 - p_m)}{p_m} \right\};$$

$$p_1 \prod_{k=2}^M (1 - p_k) + \frac{p_1}{(1 - p_1)} p_m \prod_{\substack{k=1 \\ k \neq m}}^M (1 - p_k) - \lambda_m \frac{p_1}{(1 - p_1)}, \quad 2 \leq m \leq M \quad (\text{A2.9})$$

if  $p_1 + p_m > 1$ :

$$\lambda_1 < \max \left\{ p_1 \prod_{k=2}^M (1 - p_k) + (1 - p_m) \prod_{\substack{k=1 \\ k \neq m}}^M (1 - p_k) - \lambda_m \frac{(1 - p_m)}{p_m} \right\};$$

$$p_1 \prod_{k=2}^M (1 - p_k) + \frac{p_1}{(1 - p_1)} p_m \prod_{\substack{k=1 \\ k \neq m}}^M (1 - p_k) - \lambda_m \frac{p_1}{(1 - p_1)}, \quad 2 \leq m \leq M \quad (\text{A2.10})$$

Eqn. (A2.10) can be further simplified. If  $p_1 + p_m > 1$  we have

$$\left( \frac{p_1}{1 - p_1} - \frac{1 - p_m}{p_m} \right) > 0 \quad (\text{A2.11})$$

Recalling that all  $A_m$ 's except  $A_1$ 's are positive, we have

$$\lambda_m < p_m \prod_{\substack{k=1 \\ k \neq m}}^M (1 - p_k), \quad 2 \leq m \leq M \quad (\text{A2.12})$$

Therefore

$$\lambda_m \left[ \frac{p_1}{1 - p_1} - \frac{1 - p_m}{p_m} \right] < \left[ \frac{p_1}{1 - p_1} - \frac{1 - p_m}{p_m} \right] p_m \prod_{\substack{k=1 \\ k \neq m}}^M (1 - p_k), \quad 2 \leq m \leq M \quad (\text{A2.13})$$

or

$$(1 - p_m) \prod_{\substack{k=1 \\ k \neq m}}^M (1 - p_k) - \lambda_m \frac{(1 - p_m)}{p_m} < \frac{p_1}{(1 - p_1)} p_m \prod_{\substack{k=1 \\ k \neq m}}^M (1 - p_k) - \lambda_m \frac{p_1}{(1 - p_1)},$$

$$2 \leq m \leq M \quad (\text{A2.14})$$

Using (A2.14) we see that (A2.10) is equivalent to

$$\lambda_1 < p_1 \prod_{k=2}^M (1 - p_k) + \frac{p_1}{(1 - p_1)} p_m \prod_{\substack{k=1 \\ k \neq m}}^M (1 - p_k) - \lambda_m \frac{p_1}{(1 - p_1)}, \quad 2 \leq m \leq M \quad (\text{A2.15})$$

Thus the stability region is specified by Eqs. (A2.8), (A2.9) and (A2.15). Our results of sec. 3 led to the following region of stability

$$\lambda_n < p_n \prod_{\substack{k=1 \\ k \neq n}}^M (1 - p_1) \text{ for every } n, \quad 2 \leq n \leq M \quad (\text{A2.16})$$

and

$$\lambda_1 < p_1 \prod_{k=2}^M (1 - p_k) + \sum_{i=2}^M p_1 p_i \left(1 - \frac{\lambda_i}{\mu_i^M}\right) \prod_{\substack{k=2 \\ k \neq i}}^M (1 - p_k) \quad (\text{A2.17})$$

Now let us compare the two regions. Notice that (A2.8) and (A2.16) are identical; therefore it remains to be shown that the R.H.S. of Eqs. (A2.9) and (A2.15) are strictly less than the R.H.S. of (A2.17). To show this we retain only the  $i = m$  term in the summation of (A2.17) and obtain the following inequality

$$\begin{aligned} \text{RHS of (A2.17)} &> p_1 \prod_{k=2}^M (1 - p_k) + p_1 p_m \left(1 - \frac{\lambda_m}{\mu_m^M}\right) \prod_{\substack{k=2 \\ k \neq m}}^M (1 - p_k) = \\ &= p_1 \prod_{k=2}^M (1 - p_k) + p_1 p_m \prod_{\substack{k=2 \\ k \neq m}}^M (1 - p_k) - \frac{p_1}{(1 - p_1)} \lambda_m \\ &= p_1 \prod_{k=2}^M (1 - p_k) + \frac{p_1}{(1 - p_1)} p_m \prod_{\substack{k=1 \\ k \neq m}}^M (1 - p_k) - \frac{p_1}{(1 - p_1)} \lambda_m \end{aligned} \quad (\text{A2.18})$$

A comparison of (A2.18) and the R.H.S. of (A2.9) and (A2.15) reveals that the RHS of (A2.9) and (A2.15) are strictly less than the RHS of (A2.17). This completes the comparison.

## References

1. G. Fayolle, R. Iasnagorodski, "Two Coupled Processors: The Reduction to a Riemann-Hilbert Problem", *Wahrscheinlichkeitstheorie*, pp. 1-27, 1979.
2. L. Kleinrock, Y. Yemini, "Interfering Queueing Processes in Packet-Switched Broadcast Communication, Proc. of IFIP Congress, Tokyo, 1980.
3. T. Saadawi, A. Ephremides, "Analysis, Stability, and Optimization of Slotted ALOHA with a Finite Number of Buffered Users", *IEEE Trans. on Automatic Control*, Vol. AC-26, No. 3, pp. 680-689, June 1981.
4. M. Sidi, A. Segall, "Two Interfering Queues in Packet-Radio Networks", *IEEE Trans. on Communications*, Vol. 31, No. 1, pp. 123-129, January 1983.
5. A. Segall, M. Sidi, "Priority Queueing Systems: with Applications to Packet Radio Networks", *Proc. International Seminar on Modeling and Performance Evaluation*, pp. 159-177, Paris 1983.
6. W. Szpankowski, "A Multiqueue Problem: Bounds and Approximations", *Performance of Computer-Communication Systems* (H. Rudin, W. Bux, editors), IFIP Congress, 1984.
7. W. Szpankowski, "Performance Evaluation of a Protocol for Multiaccess Systems", *Proc. Performance '83*, pp. 337-395, College Park, MD, 1983.
8. B. Tsybakov, W. Mikhailov, "Ergodicity of Slotted ALOHA Systems", *Problems of Information Transmission*, Vol. 15, No. 4, pp. 73-87, 1979.
9. W. Malyshev, "A Classification of two-dimensional Markov Chains and Piece-Linear Martingales", *Doklady Akademii Nauk USSR*, 3, pp. 526-528, 1972.
10. N. Abramson, "The ALOHA System-Another Alternative for Computer Communications", *AFIPS Conf. Proc.*, Vol. 37, pp. 281-285, 1970.
11. G. Fayolle, Labetoulle, E. Gelenbe, "Stability and Optimal Control of the Packet Switching Broadcast Channels", *JACM*, Vol. 24, pp. 375-380, July 1977.
12. W.A. Rosenkrantz, D. Towsley, "On the Instability of the Slotted ALOHA Multiaccess Algorithm", *IEEE Trans. Automatic Control*, Vol. AC-28, pp. 994-996, 1983.

13. J. Massey, P. Mathys, "The Collision Channel Without Feedback," **IEEE Trans. on Info. Theory**, Vol. 31, No. 2, pp. 192-204, March, 1985.
14. J.Y. Hui, **Fundamental Issues of Multiple Accessing**, Ph.D. Thesis, MIT, 1983.
15. L. Kleinrock, **Queueing Systems**, John Wiley & Sons, New York, 1976.
16. T. Nakasis, A. Ephremides, "Steady-State Behavior of Interacting Queues - Stability Results and a Numerical Approach", **Proc. of the Johns Hopkins Conference on Information Sciences and Systems**, Baltimore, MD, March 1985.



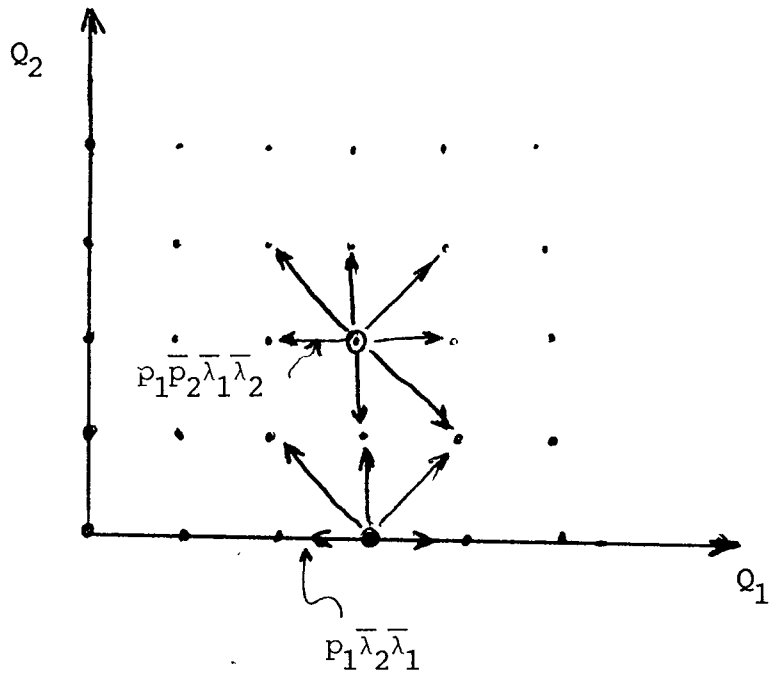


Figure 1: The 2-queue state space

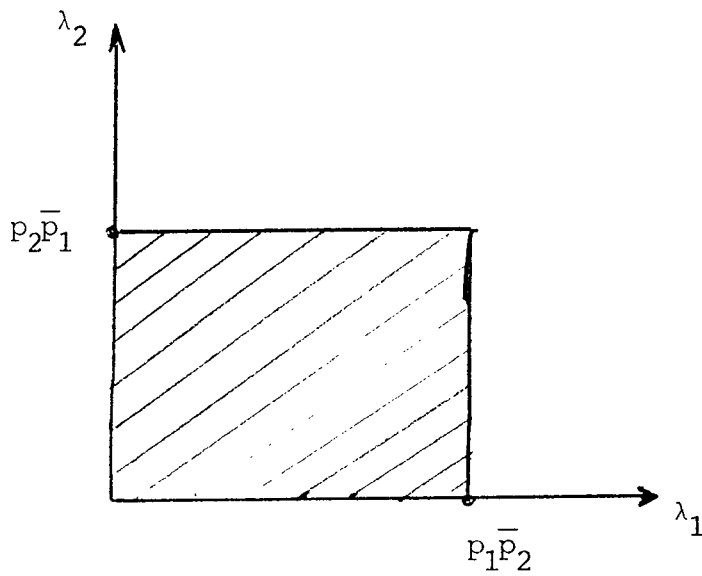


Figure 2: Stability region of  $S^2$

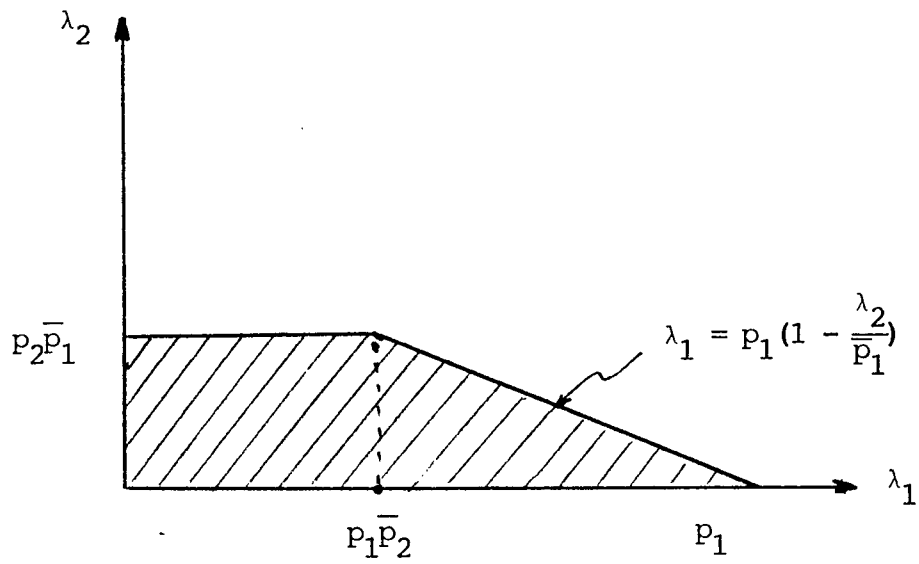


Figure 3: Stability region of  $S^1$ .

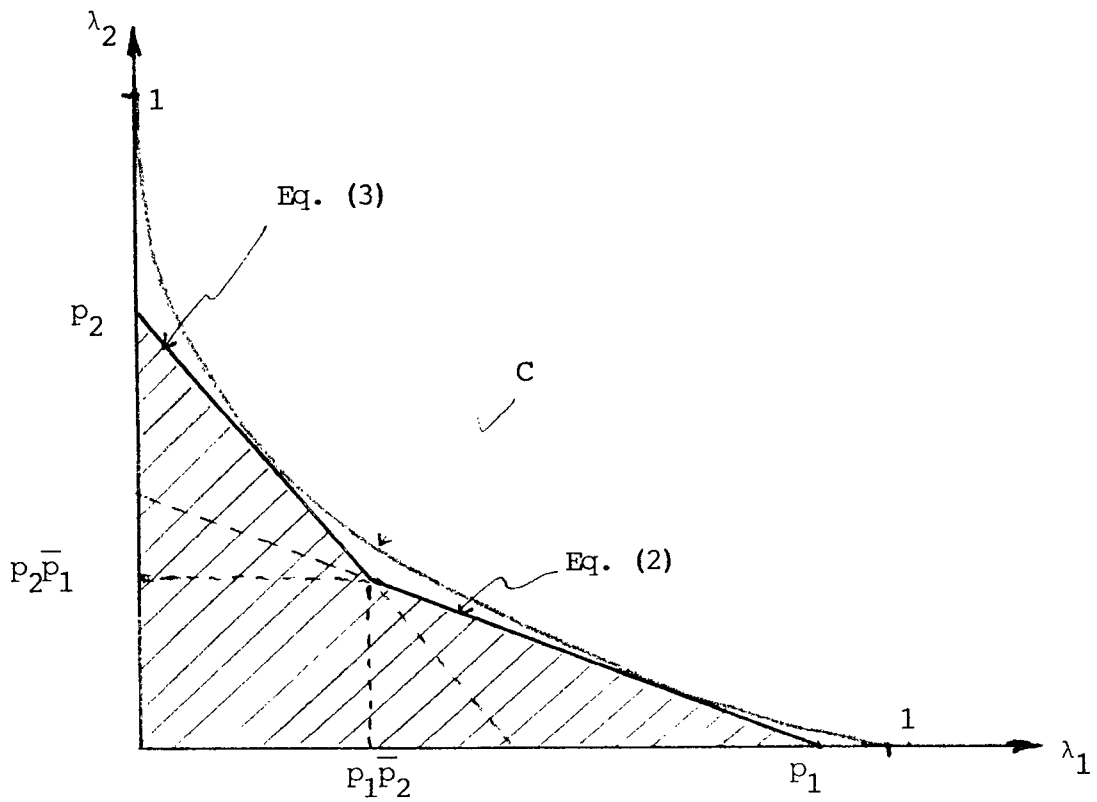


Figure 4: Stability region of  $S$ .

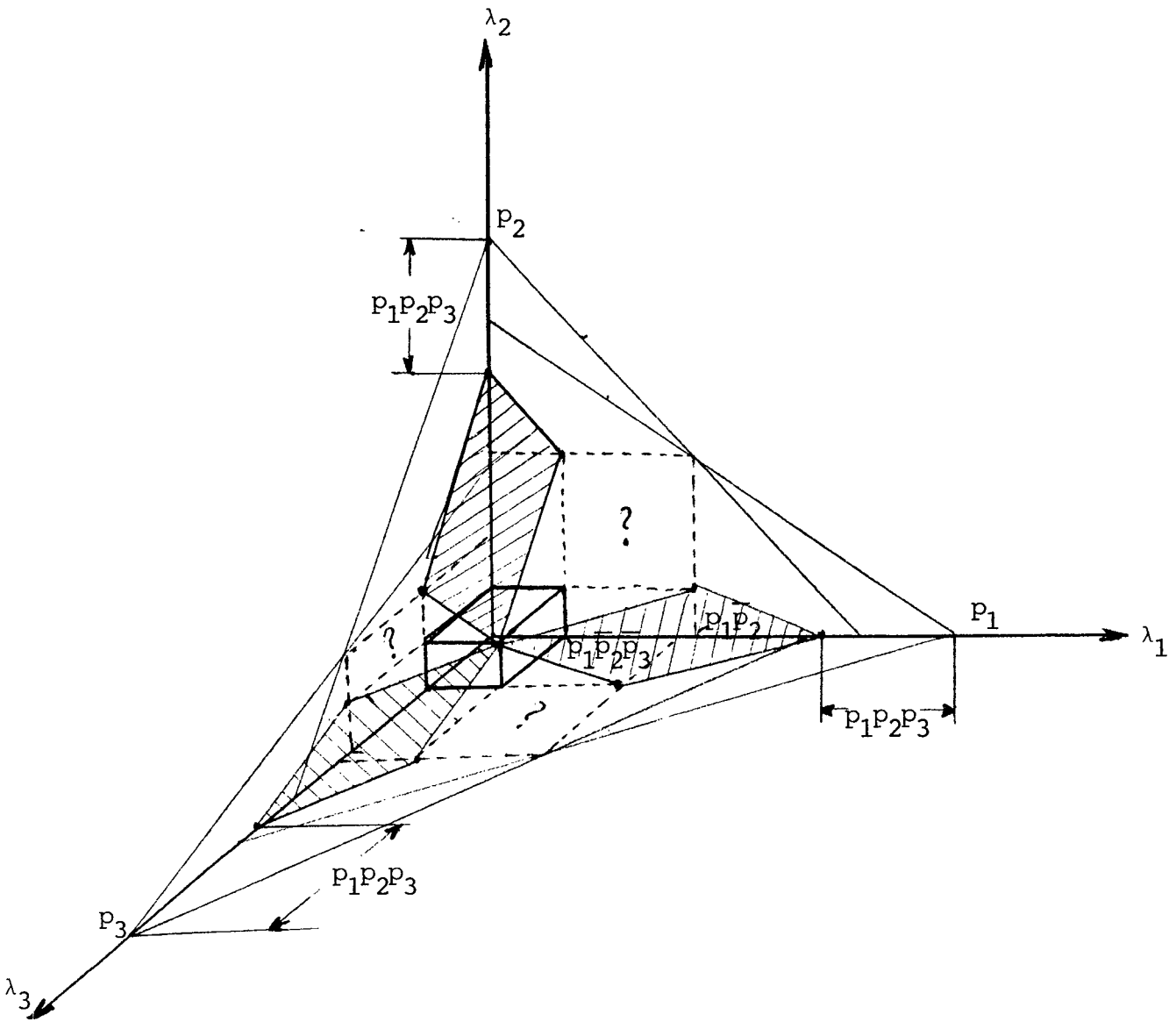


Figure 5. Stability Bounds for  $M=3$ .