Bounds on the Utilization of Aloha-Like Multiple-Access Broadcast Channels

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BOUND ON THE UTILIZATION OF ALOHA-LIKE MULTIPLE-ACCESS BROADCAST CHANNELS*

by

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ABSTRACT

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Introduction

Consider the following model of the generalized ALOHA system. Geographically separated but time synchronized transmitters send and receive messages on a common channel. If no transmitter is active, this fact is recognized by all within $t_0$ seconds. If exactly one transmitter sends a message, the message is received successfully and this is known to all within $t_1$ seconds. Finally, if two or more transmitters are active simultaneously, then a collision is said to occur and it is detected by all within $t_2$ seconds. All messages involved in the collision must be retransmitted at a later time.

This model represents a variety of systems. The slotted ALOHA channel [1] has $t_0 = t_1 = t_2$. Carrier sense multiple access radio systems [2] can detect idles quickly (carrier not present) while they distinguish between collisions and successes by using error detecting codes. Thus they have $t_0 << t_1 = t_2$. Some broadcast cable systems (e.g., the Ethernet [3]) have a "listen while transmit" feature that allows the quick abortion of transmission when a collision is detected. Thus typically $t_0 = t_2 << t_1$. Finally "reservation" systems use short messages to reserve time for longer data messages. The short messages can be seen as an idle/collision detection mechanism, and again $t_0 = t_2 << t_1$ [4].

We define the utilization of a channel access scheme as the fraction of the time during which messages are successfully transmitted. We define the "capacity" of this channel as the supremum, over all schemes, of the utilization. If the number of transmitters is finite, then the capacity is 1. Simple schemes like synchronous time division multiplexing or round robin transmission (cyclic polling) avoid collisions and can achieve this capacity. Unfortunately
they cause relatively long message delays when the generation rate of the messages is much smaller than $1/t_1$. In that case "random" transmission schemes are preferred. They allow collisions in the hope of reducing delay. Such random schemes are customarily analyzed assuming that they are infinitely many transmitters, each generating at most one message during its life-time, and that the global generation process of the messages is Poisson with rate $\lambda$.

The capacity of the channel under those conditions is still unknown. An early scheme, the slotted Aloha [1] strategy, has been said to have an utilization of $1/e$ (when $t_0 = t_1 = t_2$), but has been shown to be unstable, i.e., with probability one its utilization decreases to 0 as time goes by. A new class of protocols has recently been proposed [5], [6]. Each of those has a maximum utilization $\lambda_o$ with the property that the number of messages which have been generated but not yet successfully transmitted will be bounded with probability 1 as long as $\lambda t_1 < \lambda_o$. If $\lambda t_1 \geq \lambda_o$, the utilization of the channel is $\lambda_o$, but the expected message delay is infinite. The largest $\lambda_o$ found to this day is .4877 [7].

Note that the definition of capacity given above is not the only one that has been proposed. Pippenger [8] defines capacity as the supremum of the $\lambda$'s for which message delays can remain finite with probability one. It is clear that the value of the capacity under this definition is not larger than under the original definition, and we conjecture that they are equal. Pippenger [8] has shown that the capacity is bounded away from 1, in fact is not more than .744 ($t_0 = t_1 = t_2$). He also generalized the model to include channels when the number of transmitted messages can be determined up to some maximum $d$, and has found a bound on the utilization that is strictly increasing function of $d$, converging to 1. Moreover he showed the existence of strategies achieving utilization arbitrarily close to 1 when $d = \infty$. 
This paper generalizes Pippenger's results. An upper bound on the capacity is derived for the case of different \( t_i \)'s, its value is \( 0.704 \) when \( d=2 \) and \( t_0 = t_1 = t_2 \), and it increases rather slowly with \( d \).

Before proceeding with the precise description of the model and the derivation of the bound we will examine the implication of these results. First, an algorithm that is efficient for infinitely many sources will also be efficient for \( M < \infty \) sources as long as the typical intergeneration time at a source (\( M/\lambda \) for symmetric systems) is longer than the typical message delay. In that case, each transmission at a source is independent of the previous one, and one might as well assume that all messages have distinct sources.

Secondly, the previously mentioned results show the existence of some number \( C \), (Pippenger's Capacity) \( 0.4877 < C < 0.704 \) such that if \( \lambda t_1 < C \), the average message delay can remain bounded no matter the value of \( M \). However, if \( \lambda t_1 > C \), the average message delay must increase with \( M \). It is readily seen that the increase is linear for synchronous time division multiplexing and cyclic polling.

Determining the values of \( C \) and of the capacity remains a challenging proposition.

2. The Precise Model

To understand the following model, note that a conflict resolution protocol is a sequential decision process, thus it can be described as a tree. Every node corresponds to an "experiment", i.e., the transmission of messages. Branches correspond to outcomes, i.e., numbers of messages transmitted. Associated with each experiment is a set of times, typically a time
interval. Only those messages generated during the set corresponding to an experiment are transmitted when the experiment is made. Other conflict resolution algorithms rely on random choices, both ways are probabilistically equivalent when the generation times are Poisson.

A protocol for \([0,T]\) is an infinite d-ary tree in which there is an initial node called the root, and in which each node \(k\) is connected by branches to offsprings \(k^i\), \(i = 0, 1, 2, \ldots d\), that can be other nodes or leaves. Every node \(k\) is labelled with a measurable subset \(\gamma(k)\) of \([0,T]\). \(u(k)\) denotes the Lebesque measure of \(\gamma(k)\) divided by \(T\).

Let the random variables \(E\) denote a set of Poisson message arrival times in \([0,T]\), with expected cardinality \(v\). The execution of a protocol with respect to a finite set \(E\) in \([0,T]\) is a path through the tree defined as follows. Let \(k_0\), the first node on the path, be the root and let \(E_0\) be \(E\). Suppose that \(k_m\) and \(E_m\) have been determined, then \(k_{m+1} = k^j_m\) where \(j\) is the minimum of \(d\) and the cardinality of \(\gamma(k_m) \cap E_m\); \(E_{m+1} = E_m\) if \(k_{m+1} = k^j_m\), and \(E_{m+1} = E_m/\gamma(k_m)\) otherwise. In other words, \(j\) is the number of non-transmitted messages whose arrival times are in \(\gamma(k_m)\), or \(d\) if there are more than \(d-1\) such messages, and \(E_{m+1}\) is \(E_m\) minus any successfully transmitted message.

The set of nodes \(k\) in an execution \(\mathcal{E}\) that have offsprings \(k^{(1)}\) is denoted by \(S\), the set of successful experiments in \(\mathcal{E}\).

A protocol will be called valid if, for almost every subset \(E\) of \([0,T]\), the execution \(\mathcal{E}\) of the protocol with respect to \(E\) terminates after finitely many steps with \(E \subset \bigcup \gamma(k)\), i.e., if every message has been successfully transmitted.
The set of nodes \( k \) in an execution \( Z \) that have offsprings \( k^{(0)} \) or \( k^{(1)} \) is denoted by \( T_Z \), the set of experiments in \( Z \) not resulting in collisions.

A valid protocol will be called \textit{minimal} if for all executions \( Z \),
\[
y(k) \cap y(k') = \emptyset, \ k \neq k', \ k, k' \in T_Z.
\]
Thus, in a minimal protocol, a subset of \([0,T]\) is never tested again once it has been determined not to contain a message, or when the only message present has been successfully transmitted. Any valid protocol can be made minimal by iteratively changing the \( y(k)'s \), starting from the root, so as to satisfy the null intersection property. The execution of the protocol with respect to a set \( E \) is not affected by the change.

The execution of a protocol is a random path through the tree. \( P(k) \) denotes the probability that node \( k \) is included in an execution, and \( q(k,i) \) denotes the conditional probability that \( k^{(i)} \) follows \( k \) in the execution of a protocol.

The expected number of experiments, \( \sigma \), in an execution of a protocol has value
\[
\sigma = \sum_k P(k)
\]

The expected fraction \( q_i \) of experiments resulting in outcome \( i \) is given by (assuming \( \sigma < \infty \))
\[
q_i = \frac{1}{\sigma} \sum_k P(k) q(k,i) \tag{1}
\]
Note that \( \sum_{i=0}^{d} q_i = 1 \).
For valid protocols

\[ q_1 = \frac{v}{\sigma} \]  

(2)

We will denote \((q_0, q_1, \ldots, q_d)\) by \(q\).

The efficiency \(e\) of a protocol is simply

\[ e = \frac{\nu t_1}{\sigma \sum_{i=0}^{d} q_i t_i} \]

where \(t_i > 0\) is the time it takes to observe outcome \(i\). Note that for valid protocols

\[ e = \frac{\nu t_1}{\nu t_1 + \sigma \sum_{i \neq 1} q_i t_i} = \frac{t_1}{t_1 + f} \]

where \(f\) is defined by \(f = \frac{\sigma}{\nu} \sum_{i \neq 1} q_i t_i\) and can be thought of as the expected time overhead per message. Note that efficiencies close to 1 are achieved when \(t_1 \gg f\).

The previous relation between \(e\) and \(f\) allows us to lowerbound \(f\) (which does not depend on \(t_1\)) in order to upperbound \(e\). This is the object of the next section.
3. **Derivation of the Results**

Our goal is to lowerbound $f$ for any valid protocol.

As mentioned in the previous section, it is enough to consider minimal protocols. We will show that $q$ lies in some closed convex region $S$ of the unit simplex. The minimum over that region of $f$ considered as a function of $q$ will be our lower bound.

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We first note that for any execution $\ell$ of a minimal protocol

$$\sum_{k \in \ell} q(k) \leq 1.$$  

Averaging over $\ell$ yields

$$\sum_{k} P(k) \mu(k)(q(k,0) + q(k,1)) \leq 1$$

(3)

Next the entropy $h$ (i.e., minus the mean of the log of the probabilities) of the executions of a protocol can be written

$$h = \sum_{k} P(k) H(q(k))$$

(4)

where $H(q(k)) = - \sum_{i=0}^{d} q(k,i) \log q(k,i)$.

The probability of an execution $\ell$ of a minimal protocol is no more than $e^{-\nu} \prod_{b \in \ell} \nu a(b)$, as one arrival must have occurred in every $\gamma(b)$, $b \in S_{\ell}$, which are disjoint, and no arrival could have occurred outside such a subset.

Thus

$$h \geq \mathbb{E}(- \log \prod_{k \in S_{\ell}} \mu(k)e^{-\nu}) = \nu \log e - \sum_{k} P(k)q(k,1)\log(\mu(k))$$

(5)
The right hand side of (5) is not less that
\[
\log e - \frac{1}{k} P(k) q(k, 1) \log \frac{q(k, 1)}{q(k, 0) + q(k, 1)}
\]
as can be seen by using the inequality \(\ln(x) \leq x - 1 \) and (3).

Subtracting this last expression from the right hand side of (4), dividing by \(a\) and using (2) one obtains
\[
\frac{1}{a} \sum_k P(k) g(q(k)) \geq 0,
\]
where
\[
g(x) = -x_1 \log (x_0 + x_1) - \sum_{i=0}^{d} x_i \log(x_i) - x_i \log e
\]

It is shown in the appendix that \(g\) is a strictly concave function, thus by Jensen's inequality and (1),
\[
g(q) \geq 0
\]

To obtain a lower bound \(\alpha\) on \(f\), we find
\[
\alpha = \min f = \min_{q \in S} \min_{q \in S} \sum_{i \neq 1} q_i t_i
\]
where \(S = \{ q \in \mathbb{R}^{d+1}, q_i \geq 0, \sum q_i = 1, g(q) \geq 0 \}\)

\(\alpha\) is finite, as \(S\) contains the point \(\left(\frac{1}{d+1}, \ldots, \frac{1}{d+1}\right)\). Note that \(q\) achieving the above minimum also achieves
\[
o = \min_{q \in S} \left( \sum_{i \neq 1} q_i t_i - q_1 a \right) \tag{6}
\]
A $q^*$ in $S$ minimizing (6) must satisfy $g(q^*) = 0$, as if $g(q^*) > 0$, then $q_i^* \neq 1$ and one can decrease the objective function by increasing $q_i^*$ and decreasing some $q_i^*$ ($i \neq 1$) by the same amount. A solution exists as $S$ is compact in $\mathbb{R}^{d+1}$ and the objective function (6) is continuous. The solution is unique as $g$ is strictly concave. The problem of minimizing (5) subject to equality constraints is standard. Necessary and sufficient conditions for the optimality of $q^*$ are

\[ g(q^*) = 0 \]
\[ \sum q_i^* = 1 \]  \hspace{1cm} (7)
\[ \ln q_o + \frac{q_i}{q_o + q_i} + \mu t_o + (\lambda + 1) = 0 \]  \hspace{1cm} (8)
\[ \ln (q_o + q_i) + \frac{q_i}{q_o + q_i} - \mu a + (\lambda + 1) = 0 \]  \hspace{1cm} (9)
\[ \ln q_i - \mu t_i + (\lambda + 1) = 0 \]  \hspace{1cm} i = 2, 3 \ldots d. \]  \hspace{1cm} (10)
\[ \sum_{i \neq 1} t_i q_i - \alpha q_1 = 0 \]  \hspace{1cm} (11)

where $\lambda$ and $\mu$ are Lagrange multipliers.

One checks that $g(q^*) = 0$ implies $\lambda + 1 = 0$. To find $q^*$ and $\mu$, one must proceed numerically. An iterative way is to first guess a value of $\mu$, thus determining $q_i^* \ 2 \leq i \leq d$ by (10). $q_o^* = q_1^*$ is obtained from (7), then $q_o^*$ and $q_1^*$ from (8). If (9) and (11) are not verified, the value of
u should be changed and the process repeated. Figure 1 showing the upper bound on the overhead when \( d = 2 \), as a function of \( t_1/t_2 \) was so obtained.

The solution is simpler when \( t_i = t \) \((i = 0, 2, \ldots, d)\). Defining \( L = \frac{q_1}{q_0 - q_1} \), one obtains from (8) and (10) that \( q_i = q_0 e^{iL} \) \(2 \leq i \leq d\), and from (7) that \( q_0 = \frac{1}{1-L + (d-1) e^L} \). Subtracting (8) from (9) yields

\[
\log(1-L) + u(t+\alpha) = 0. \quad \text{As from (8) and (11),} \quad u(t+\alpha) = \frac{1-L}{L q_0} (-L \ln q_0), \quad L
\]

satisfies the equation:

\[
\ln(1-L) = \frac{(1+(1-L)(d-1)e^L)(L + \ln(1-L) - \ln(1+(1-L)(d-1)e^L))}{L}
\]

Once \( L \) is computed, one can find \( q_1^* \), which here is an upperbound on the relative frequency of experiments resulting in a success, by the formula \( q_1^* = \frac{L}{1+(1-L)(d-1)e^L} \cdot \frac{t(1-q_1^*)}{q_1^*} \) is equal to \( \alpha \), our lowerbound on the time overhead per message, \( f \). Numerical results appear in Table 1.

One sees that they are not very sensitive to \( d \). Indeed one can derive from (12) that \( 1-L = \frac{1}{e(d-1)\ln d} \left(1 + 0 \left(\frac{1}{(\ln d)^2}\right)\right) \),

where \( (\ln d)^2 0 \left(\frac{1}{\ln (d))2}\right) \) is bounded for \( d \geq 1 \), and consequently

\[
q_1^* = \frac{\ln d}{1 + \ln d} (1 + 0 ((\ln(d))^{-2}). \quad \text{This result indicates that determining how many messages are involved in a collision does not greatly pay off.}
\]
<table>
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<th>$q_1^*$</th>
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</tr>
<tr>
<td>10000</td>
<td>.0781</td>
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</tr>
</tbody>
</table>

**TABLE 1**

Lowerbound $\alpha$ on the time overhead per message and upperbound $q_1^*$ on the relative frequency of success as $d$ varies ($t_i = 1$, $i = 0,2,\ldots,d$)
Appendix

We will show here that the function

\[ g(x) = -x_0 \log x_0 - x_1 \log (x_0 x_1) - \frac{1}{2} \sum_{i=2}^{n} x_i \log x_i - x_1 \log e \]

is strictly concave for \( x > 0 \). By inspection, this is immediately true for all terms, except the second one. We prove now that \( -x_0 \ln x_0 - x_1 \ln (x_0 x_1) \) is strictly concave by showing that the matrix of second partial derivatives is negative definite. This matrix is equal to

\[
\begin{pmatrix}
\frac{-(x_0^2 + x_0 x_1 x_1^2)}{x_0 (x_0 x_1)^2} & \frac{-x_0}{(x_0 x_1)^2} \\
\frac{-x_0}{(x_0 x_1)^2} & \frac{-(2x_0 x_1)}{(x_0 x_1)^2}
\end{pmatrix}
\]

The upper diagonal term is always negative, while the determinant, \( \frac{1}{x_0 (x_0 x_1)} \), is always positive. Thus Sylvester's test is verified.
References


