Probabilistic Analysis of a Crossbar Switch

B.A. Makrucki
T.N. Mudge

March 1981

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DEPARTMENT OF ELECTRICAL AND COMPUTER ENGINEERING
SYSTEMS ENGINEERING LABORATORY
THE UNIVERSITY OF MICHIGAN, ANN ARBOR 48109
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Abstract

This report presents a probabilistic analysis of a crossbar switch interconnection network. A crossbar switch can be used to interconnect various combinations of computer subsystems. In the analysis below it is assumed, without loss of generality, that the crossbar is being used to connect \( N \) processors to \( M \) memories. The crossbar is termed an \( N \times M \) crossbar (read "\( N \) to \( M \) crossbar"). General expressions are developed for a variety of performance figures for an \( N \times M \) crossbar including: the probability of a memory request being accepted (i.e. not being blocked by another request to the same memory), the expected bandwidth of the crossbar, and the average wait time of a request before it is accepted. Closed form solutions to these expressions are given for the uniform request case and for the favorite memory case (i.e. where processor \( i \) requests memory \( i \) with a higher probability than others' memories). The closed form solutions are tested against simulations.
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1. Introduction

This report presents a probabilistic analysis of a crossbar switch interconnection network. A crossbar switch can be used to interconnect various combinations of computer subsystems including processors to processors, processors to memories, processors to I/O devices, and memories to I/O devices. In the analysis below it is assumed, without loss of generality, that the crossbar is being used to connect \( N \) processors to \( M \) memories. This multiprocessor system is depicted in Figure 1. The crossbar is termed an \( N \)-\( M \) crossbar (read "\( N \) to \( M \) crossbar"). In the operation of the system it is further assumed that there is a system wide clock, and that read and write memory requests made by the processors can only occur in synchronism with this clock.

The designer of a multiprocessor system has a wide variety of interconnection networks to choose from. An obvious candidate for an interconnection

![Figure 1. Multiprocessor System.](image-url)
network is a shared bus—several processors and memories connected to a shared time-multiplexed bus. However, a shared bus only provides high speed interconnection if it is very fast relative to processors and memories. If it is desired to connect N processors to M memories a shared bus' performance decays as N increases. To improve performance over the shared bus a crossbar switch may be used to establish multiple bus connections between processors and memories. A crossbar allows any set of simultaneous interconnections in which at most one input bus is connected to each output bus. Unlike the shared bus its performance does not decrease with increase in N. However, the component complexity of the crossbar grows as \(O(NM)\). For this reason past proposals for tightly coupling multiple processors to multiple memories have steered away from crossbar switches. In their place a whole range of ingenious networks have been proposed ranging from the early ideas of Clos and Benes [Clo53, Ben65] through the Omega network of Lawrie [Law75], the indirect binary n-cube of Pease [Pea77], the Delta network of Patel [Pat79], and the Data Manipulator network of Feng [WuF80], which have many of the connectivity properties of a crossbar without the component complexity. For a good survey of the state-of-the-art in networks see Siegel [Sie80]. The component complexity of these networks all grow as \(O(N \log_2 N)\) rather than \(O(N^2)\) (assume \(N = M\)).

In the context of VLSI technology it is no longer clear that reduced component complexity is an advantage within a single IC. For example, preliminary layouts for a Delta network and a crossbar carried out by us [MaM81] suggest that the reduced complexity networks do not appear to translate into more efficient space utilization in an IC layout. The unimportance of component complexity or device count as a measure of design efficiency in ICs has been discussed by Thompson and Franklin [Tho80, Fra80] among others. Furthermore, although the reduced complexity networks preserve some of the connectivity properties of a crossbar they do not preserve bandwidth [Pat79]. At the

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present time the limiting parameter is the number of pins required of any package that the IC might be put into. For these reasons we have decided to explore the design and analysis of crossbar switches more fully as a basic building block for interconnection networks. This is in contrast to other proposals for reduced complexity networks (for example [CiS81]). As mentioned above this report presents a probabilistic analysis, design is discussed in a companion report [MaM81].

Previous work on the probabilistic analysis of crossbars can be found in [Rav72, Str70, Bha75, BaS76, CKL77]. The analysis presented below is an extension of this work. Analyses presented in the above references assume that each processor makes a memory request through the crossbar every memory cycle. Furthermore, these requests are assumed to be uniformly distributed among the memories. Our analysis relaxes both of these conditions. In particular, we allow the possibility of a processor not making a memory request during a memory cycle. Also, we allow the requests to be non-uniformly distributed among the memories. In the case of each processor having a favorite memory, i.e., being more likely to request one particular memory than any other, a new closed form solution for the probability of request acceptance is developed. In contrast to some of the analyses presented in [BaS76, Bha75, CKL77] we have not attempted to model dependencies between successive memory requests and little attempt has been made to model the effect of being allowed to queue memory requests, although an expression for the expected wait time of a memory request is developed in section 4. Including queues in the analysis results in very complex models that, although more accurate, can only be solved for very small values of $N$ and $\lambda$ (see [Bha75]). For the analysis presented in this report, i.e., when the above mentioned memory request conditions are relaxed, including queues complicates matters further. Including dependencies makes the analysis even more intractable.

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2. Analysis of an N×N Crossbar

**Assumption:** Consistent with the system behavior outlined in the Introduction it is assumed that events in the system occur at discrete time intervals defined by the system clock. In particular, memory requests made by the processors occur synchronously with the system clock. We shall term this assumption the synchronous request assumption (SRA).

**Definition 1:** Let \( R_i \) be the event that processor \( i \) requests any memory.

**Definition 2:** Let \( S_{ij} \) be the event that processor \( i \) requests memory \( j \).

**Assumptions:** Assume the request behavior of a processor is an independent process. Assume also that the request behavior of each processor is independent of that of any other processor. We shall term these assumptions the independent request property (IRP).

One consequence of the IRP is that our analysis does not model resubmission of denied requests. In other words if several processors request the same memory those that do not get the memory (all but one) "lose" their request, and in the next time period their requests will be made without regard to this lose since the request behavior of a processor is an independent process. In a realistic situation the request would most likely be resubmitted. The effect of this shortcoming of the analysis is measured experimentally, and the results are prescribed in a later section.

One other consequence of the IRP is that the following holds:

\[
\Pr\{S_{ab} \cap S_{cd}\} = \Pr\{S_{ab}\}\Pr\{S_{cd}\} \quad \text{for all } a \neq c
\]

**Assumptions:** Assume that it is possible to measure the following two statistics empirically.

(1) The probability that processor \( i \) requests any memory: \( r_i \).
(2) The probability that processor i requests memory j given that it makes any request at all: \( p_i(j) \).

The values of \( r_i \) and the \( p_i(j) \) for a particular processor i will be termed the request distributions for that processor.

The statistic \( r_i \) and the \( p_i(j) \) may be estimated from memory reference counts obtained from typical programs. This technique was used by Baskett and Smith in [BaS76].

From definitions 1 and 2 and the above assumptions it follows that:

\[
    r_i = \Pr\{R_i\} \tag{1}
\]

\[
    p_i(j) = \Pr\{S_{ij}|R_i\} \tag{2}
\]

From the definition of conditional probability we have:

\[
    \Pr\{S_{ij}|R_i\} = \frac{\Pr\{S_{ij} \cap R_i\}}{\Pr\{R_i\}} \quad \text{(Pr}\{R_i\}>0)
\]

However, from definitions 1 and 2 it follows that:

\[
    S_{ij} \cap R_i = S_{ij}
\]

(See event space diagrams in Figure 2.)

Therefore:

\[
    \Pr\{S_{ij}\} = \Pr\{S_{ij}|R_i\} \Pr\{R_i\}
\]

Using (1) and (2) this can be written:

\[
    \Pr\{S_{ij}\} = r_i p_i(j) \tag{3}
\]

**Definition 3:** Let \( A_{ij} \) be the event that processor i requests memory j and is accepted.
Definition 4: Let $B_{ijk}$ be the event that processor $i$ and $k-1$ others request memory $j$.

Assumption: If exactly $k$ processors request memory $j$ assume that the one processor whose request is granted is selected from the set of $k$ with a probability of $\frac{1}{k}$. We shall term this the uniform selection rule (USR). In the notation developed so far the USR can be expressed as follows:

$$P_r(A_{ij} | B_{ijk}) = \frac{1}{k}$$

From definitions 2 and 3 it follows that:

$$A_{ij} \cap S_{ij} = A_{ij}$$

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(See event space diagrams in Figure 2.)

Furthermore, from definitions 2 and 4 it follows that:

\[ S_{ij} = \bigcup_{k=1}^{N} B_{ijk} \]

(See event space diagrams in Figure 2.)

Combining these two observations gives the following:

\[ A_{ij} = A_{ij} \cap S_{ij} = A_{ij} \cap \left[ \bigcup_{k=1}^{N} B_{ijk} \right] \]

Distributing gives:

\[ A_{ij} = \bigcup_{k=1}^{N} \left[ A_{ij} \cap B_{ijk} \right] \] (4)

From definition 4 it follows that the events \( B_{ijk} \) are mutually exclusive with respect to \( k \), i.e. \( B_{ijk_1} \cap B_{ijk_2} = \emptyset \) for all \( k_1 \neq k_2 \). Therefore, the events \( A_{ij} \cap B_{ijk} \) are mutually exclusive with respect to \( k \) also. This fact allows us to express the relationship between the probabilities associated with the events in equation 4 as follows:

\[ \Pr\{A_{ij}\} = \sum_{k=1}^{N} \Pr\{A_{ij} \cap B_{ijk}\} \] (5)

Using the definition of conditional probability again, we have:

\[ \Pr\{A_{ij} \cap B_{ijk}\} = \Pr\{A_{ij} | B_{ijk}\} \Pr\{B_{ijk}\} \]

Substituting this in 5 gives:

\[ \Pr\{A_{ij}\} = \sum_{k=1}^{N} \Pr\{A_{ij} | B_{ijk}\} \Pr\{B_{ijk}\} \]

Applying the USR assumption to this gives:

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\[
\Pr\{A_{il}\} = \sum_{k=1}^{N} \frac{1}{k} \Pr\{B_{lk}\}
\]  

Definition 5: Define a binary vector, \( v_{jk} = \langle v_{jk}(1), \ldots, v_{jk}(N) \rangle \) to represent the request pattern seen by memory \( j \) when processors \( i \) and \( k-1 \) others request memory \( j \). The elements of \( v_{jk} \) are defined as follows:

\[
v_{jk}(h) = 1
\]

iff processors \( h, i, \) and \( k-2 \) others request memory \( j \).

\[
v_{jk}(h) = 0
\]

iff processor \( h \) makes no request or requests a memory other than \( j \), and processors \( i \) and \( k-1 \) others request memory \( j \).

There are \( \binom{N-1}{k-1} \) distinct vectors \( v_{jk} \). When necessary they will be distinguished by an additional subscript as follows:

\[
v_{ijkl} \quad l = 1, \ldots, \binom{N-1}{k-1}
\]

From definition 5 it follows that:

\[
w_{ijkl}(i) = 1
\]

\[
\sum_{h=1}^{N} w_{ijkl}(h) = k
\]

and

\[
\{w_{ijkl}(h) = 1\} \Rightarrow \text{event } S_{lj}
\]

However, the converse of relation 7 does not necessarily hold since the event that processor \( h \) requests memory \( j \) says nothing about processor \( i \) requesting memory \( j \) nor does it indicate how many processors request memory \( j \).
Definition 6: Denote the occurrence of event $S$ and its non-occurrence by $S^1$ and $S^0$ respectively.

From definitions 5 and 6 and relation 7 we have:

$$w_{ijkl} \Rightarrow \bigcap_{h=1}^{N} S_{hj}^{w_{ijkl}(h)}$$

Furthermore, the converse holds since the right hand side completely specifies the request pattern seen by memory $j$. Therefore, we can write:

$$w_{ijkl} \iff \bigcap_{h=1}^{N} S_{hj}^{w_{ijkl}(h)}$$ \hspace{1cm} (8)

Definition 7: Let $w_{ijkl}$ be the event corresponding to the request pattern $\nu_{ijkl}$.

From relation 8 and definition 7 it follows that:

$$w_{ijkl} = \bigcap_{h=1}^{N} S_{hj}^{w_{ijkl}(h)}$$

Applying the IRP assumptions to this allows us to express the relationship between the probabilities associated with the events as follows:

$$\Pr\{w_{ijkl}\} = \prod_{h=1}^{N} \Pr\{S_{hj}^{w_{ijkl}(h)}\}$$ \hspace{1cm} (9)

From equation 3 and definition 6 it follows that:

$$\Pr\{S_{hj}^1\} = r_{h}p_{h}(j)$$

and

$$\Pr\{S_{hj}^0\} = 1 - r_{h}p_{h}(j)$$

Substituting these results into equation 9 gives:

$$\Pr\{w_{ijkl}\} = \prod_{h=1}^{N} [r_{h}p_{h}(j)]^{w_{ijkl}(h)} [1 - r_{h}p_{h}(j)]^{1-w_{ijkl}(h)}$$

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However, as noted earlier from definition 5 it follows that:

\[ w_{ijkl}(i) = 1 \]

Thus the above equation can be rewritten to give:

\[
\Pr\{W_{ijkl}\} = r_iP_j(j) \prod_{h=1}^{N_i} (r_hP_h(j))^{w_{ijkl}(h)}[1 - r_hP_h(j)]^{1-w_{ijkl}(h)}
\]  
(10)

From definition 4, 5 and 7 it follows that:

\[
B_{ijk} = \bigcup_{l=1}^{N} W_{ijkl}
\]  
(11)

(See event space diagrams in Figure 2.)

Furthermore, from definitions 5 and 7 it may be concluded that events \( W_{ijkl} \) are mutually exclusive with respect to \( l \), i.e. \( W_{ijkl_1} \cap W_{ijkl_2} = \emptyset \) for all \( l_1 \neq l_2 \). This allows us to express the relationship between the probabilities associated with the events in equation 11 as follows:

\[
\Pr\{B_{ijk}\} = \sum_{i=1}^{N} \Pr\{W_{ijkl}\}
\]  
(12)

Substituting from equation 10 into 12 gives:

\[
\Pr\{B_{ijk}\} = r_iP_j(j) \sum_{l=1}^{N} \prod_{h=1}^{N_i} (r_hP_h(j))^{w_{ijkl}(h)}[1 - r_hP_h(j)]^{1-w_{ijkl}(h)}
\]  
(13)

Substituting equation 13 into equation 6 gives us the following expression for the probability of processor \( i \) successfully requesting memory \( j \):

\[
\Pr\{A_{ij}\} = r_iP_j(j) \sum_{k=1}^{N} \frac{1}{k} \sum_{l=1}^{N} \prod_{h=1}^{N_i} (r_hP_h(j))^{w_{ijlk}(h)}[1 - r_hP_h(j)]^{1-w_{ijlk}(h)}
\]  
(14)
From the point of view of individual processors an interesting quantity is the probability of processor i successfully requesting memory j once i has requested j, i.e. \( \Pr\{A_{ij} | S_{ij}\} \). This gives a measure of a request not being blocked by a conflict. To obtain \( \Pr\{A_{ij} | S_{ij}\} \) we need to modify equation 14 slightly. This can be done in a straightforward manner as follows. Above it was observed that definitions 2 and 3 lead to:

\[ A_{ij} \cap S_{ij} = A_{ij} \]

Combining this with the definition of conditional probability gives:

\[
\Pr\{A_{ij} | S_{ij}\} = \frac{\Pr\{A_{ij} \cap S_{ij}\}}{\Pr\{S_{ij}\}} = \frac{\Pr\{A_{ij}\}}{\Pr\{S_{ij}\}}
\]

Substituting for \( \Pr\{S_{ij}\} \) from equation 3 and for \( \Pr\{A_{ij}\} \) from equation 14 gives:

\[
\Pr\{A_{ij} | S_{ij}\} = \sum_{k=1}^{N} \frac{1}{k} \sum_{i=1}^{N} \prod_{h=1, h \neq i}^{N} \left[ r_{h} p_{h}(j) \right]^{w_{j,h}(h)} \left[ 1 - r_{h} p_{h}(j) \right]^{w_{j,h}(h)}
\]  

(15)

In general, it is desirable that an interconnection network provide as many simultaneous channels as needed from its N inputs to its M outputs. In the case of an N-M crossbar the number of simultaneous channels needed is given by:

\[
\text{Req}[BW] \triangleq \sum_{i=1}^{N} r_{i}
\]

(16)

The left hand side reads "requested bandwidth". It is measured in "channels", however, if the data rate per channel can be expressed in Hertz then it is possible to represent \( \text{Req}[BW] \) in Hertz; units more commonly used to measure bandwidth. Two other bandwidth measures are useful for characterizing an N-M crossbar. They are defined as follows:

\[
\text{E}[BW] \triangleq \sum_{i=1}^{N} \sum_{j=1}^{M} \Pr\{A_{ij}\}
\]

(17)

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Max[BW] \overset{\Delta}{=} \min(N,M) \tag{18}

The first of these is just the expected bandwidth, i.e. the expected number of channels in use between the N processors and the M memories. The second definition is a measure of the maximum number of channels that can exist between the processors and memories if request conflicts never occur. It can be seen that Max[BW] depends only on the structure of the N-M crossbar, whereas Req[BW] and E[BW] are also dependent on the request distributions that define each processor's request behavior (i.e. r_i and p_i(j)).

Using the definitions given in equations 16, 17, and 18 we can define two figures of merit for an N-M crossbar operating under a particular set of request distributions as follows:

\[ E \overset{\Delta}{=} \frac{E[BW]}{Req[BW]} \tag{19} \]

\[ U \overset{\Delta}{=} \frac{E[BW]}{Max[BW]} \tag{20} \]

Where E is a measure of the effectiveness of the crossbar in fulfilling the demands of the particular set of request distributions, and U is a measure of the utilization of the crossbar given the particular set of request distributions. In general, E and U are both functions of N, M, r_i and p_i(j) for all i and j. Also, \( 0 \leq E \leq 1 \) and \( 0 \leq U \leq 1 \).
3. Closed Form Solutions

For certain functional forms for \( r_i \) and \( p_i(j) \) it is possible to derive closed form expressions for equation 14, and as a consequence equations 15 through 20 too. In this section we shall illustrate this possibility with several examples.

3.1. Uniform Request Distributions

In this case:

\[
\text{let } \quad r_i = r \quad \text{for all } \quad i \quad \text{and} \quad j
\]
\[
\text{and } \quad p_i(j) = \frac{1}{M} \quad \text{for all } i
\]

Then equation 14 simplifies to the following:

\[
\Pr\{A_{ij}\} = \frac{r}{M} \sum_{k=1}^{N} \frac{1}{k} \sum_{l=1}^{k-1} \prod_{h=1, h \neq k}^{N} \left[ \frac{r}{M} \right]^{w_{ijh}[1-\frac{r}{M}]^{1-w_{ijh}}}
\]  
(21)

Recall that for all values of \( l \) exactly \( k \) elements of \( w_{ijh} \) are 1 (\( k-1 \) elements if the \( h=i \) case is omitted), and \( N-k \) elements are 0. Therefore, equation 21 reduces to:

\[
\Pr\{A_{ij}\} = \frac{r}{M} \sum_{k=1}^{N} \frac{1}{k} \sum_{l=1}^{N-1} \left[ \frac{r}{M} \right]^{k-l}[1-\frac{r}{M}]^{N-k}
\]  
(22)

Since \( l \) is no longer a parameter of any of the terms in the scope of the rightmost summation, equation 22 reduces to:

\[
\Pr\{A_{ij}\} = \frac{r}{M} \sum_{k=1}^{N} \frac{1}{k} \left( \frac{N-1}{k-1} \right) \left[ \frac{r}{M} \right]^{k-l}[1-\frac{r}{M}]^{N-k}
\]

This can be written as:

\[
\Pr\{A_{ij}\} = \frac{1}{N} \sum_{k=1}^{N} \left( \frac{N}{k} \right) \left[ \frac{r}{M} \right]^{k}[1-\frac{r}{M}]^{N-k}
\]  
(23)
Recalling the binomial expansion of $\left[ \frac{r}{M} + (1 - \frac{r}{M}) \right]^N$ we can rewrite 23 as follows:

$$\Pr\{A_y\} = \frac{1}{N} \left[ 1 - \left(1 - \frac{r}{M}\right)^N \right]$$

(24)

Also:

$$\Pr\{A_y \mid S_y\} = \frac{M}{rN} \left[ 1 - \left(1 - \frac{r}{M}\right)^N \right]$$

(25)

Equations 16 through 18 simplify to:

$$\text{Req}[BW] = rN$$
$$E[BW] = M\left[1 - \left(1 - \frac{r}{M}\right)^N\right]$$
$$\text{Max}[BW] = \min(N,M)$$

(26)

Therefore equations 19 and 20 simplify to:

$$E = \frac{M}{rN} \left[1 - \left(1 - \frac{r}{M}\right)^N\right]$$
$$U = \text{Max}\{\left[1 - \left(1 - \frac{r}{M}\right)^N\right], \frac{M}{N}\left[1 - \left(1 - \frac{r}{M}\right)^N\right]\}$$

(27)

Further simplifying assumptions lead to even more concise closed forms as follows:

Let $\alpha = \frac{N}{M}$

Then equation 25 has a limiting form as follows:

$$\Pr\{A_y \mid S_y\} = \frac{1}{r\alpha} \left[1 - \left(1 - \frac{r\alpha}{N}\right)^N\right]$$

If $N$ and $M$ grow large this last equation has a limiting form as follows:

$$\lim_{N \to \infty} \frac{1}{r\alpha} \left[1 - \left(1 - \frac{r\alpha}{N}\right)^N\right] = \frac{1}{r\alpha} \left[1 - e^{-r\alpha}\right]$$

(28)

Some or all of the equations 24 through 28 have previously been derived in [Str70, CKL77, BrD77] for the case $r=1.0$. 

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Figures 3 through 7 illustrate some of the results obtained by assuming uniform request distributions. Figure 3 shows \(\Pr(A | S)\) and \(E[BW]\) for a 4-M crossbar as a function of \(M\) when \(r = 0.1\) (i.e. a 10% duty cycle for memory requests). As one would expect the probability of a submitted request being accepted approaches 1 as \(M\) increases, since increasing \(M\) while keeping \(N\) constant reduces the likelihood of requests conflicting for memories. For the

\[
\begin{align*}
p(AIS) & = 1.0 \\
E(BW) & = 0.4
\end{align*}
\]

\(\text{Figure 3.}\)

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same reason the expected bandwidth in use, \( E[BW] \), approaches the expected bandwidth requested, \( \text{Req}[BW] \). (Recall that \( \text{Req}[BW] = rN = 0.4 \)). Figure 4 again shows \( \text{Pr}[A_{ij} \mid S_{ij}] \) and \( E[BW] \) for a 4-\( M \) crossbar as a function of \( M \), however, the request has been increased to the point where every processors makes a request at every memory cycle, i.e. \( r = 1.0 \). Again \( \text{Pr}[A_{ij} \mid S_{ij}] \) asymptotically approaches 1 with increase in \( M \), but at a slower rate than before. In addition, \( E[BW] \) approaches \( \text{Req}[BW] \) (=4), however, since \( r = 1.0 \), \( E[BW] \) also

\[ \begin{align*}
\text{Figure 4.}
\end{align*} \]
approaches the structural limit of the crossbar, namely $\text{Max[BW]} (=4)$. Figures 5 through 7 illustrate how $\text{E[BW]}$ varies with $r$ for a 4-4 crossbar, a 16-16 crossbar, and a 64-64 crossbar respectively. Notice that in all three cases 65-70% of the maximum number of channels available are being used when $r = 1.0$.

3.2. Favorite Memory Distributions

In this case assume that processor $i$ is more likely to request memory $i$ than other memories. To model this case let:

\[ \text{Figure 5.} \]
Figure 6.

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\[ r_i = r \quad \text{for all } i \]
\[ p_i(i) = p \quad \text{for all } i \]
\[ p_i(j) = \frac{1-p}{M-1} \quad \text{for all } i \text{ and } j, j \neq i \]

For \( i \) to be the favorite we require that \( p > \frac{1}{M} \). For brevity and to simplify the analysis let \( N = M \). The cases \( N > M \) and \( N < M \) have closed forms that can be obtained using a straightforward extension of the analysis presented below. In deriving an expression for \( \Pr\{A_{ij} | S_{ij}\} \) two cases arise:

\emph{Case 1:} \quad i = j

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Equation 15 simplifies to the following:

\[
\Pr\{A_{ii} \mid S_{ii}\} = \sum_{k=1}^{N} \frac{1}{k} \sum_{l=1}^{k-1} \left[ \frac{r(1-p)}{M-1} \right]^{k-1} \left[ 1 - \frac{r(1-p)}{M-1} \right]^{N-k}
\]

For convenience let:

\[
\frac{r(1-p)}{M-1} = t
\]

Then:

\[
\Pr\{A_{ii} \mid S_{ii}\} = \sum_{k=1}^{N} \frac{1}{k} \binom{N-1}{k-1} t^{k-1} (1-t)^{N-k}
\]

Recalling the binomial expansion of \([t+(1-t)^N]\), allows us to write:

\[
\Pr\{A_{ii} \mid S_{ii}\} = \frac{1}{Nt} [1-(1-t)^N]
\]  \hspace{1cm} (29)

As would be expected the expression on the right hand side of equation 29 reaches a maximum as \(p \to 1\). Indeed, using l'Hôpital's rule it can be shown that:

\[
\lim_{p \to 1} \Pr\{A_{ii} \mid S_{ii}\} = 1
\]

**Case 2:** \(i \neq j\)

In this case the set of possible request patterns seen by memory \(j\) partition into two mutually exclusive subsets. These subsets are distinguished by whether or not each of their patterns include a request from processor \(j\) (a favorite request). Let \(\{w_{ij}^{**}\}\) be the set of possible request patterns where \(*\) indicates that the subscript in that position ranges over its allowed values. Then the set of possible request patterns that each include a request from processor \(j\) is given by:

\[
\alpha = \{ \ w_{ij}^{**} \mid w_{ijkl}(j)=1 \ \}
\]

And the set of possible request patterns none of which include a request from

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processor \( j \) is given by:

\[
\beta = \{ \omega_{nj} \mid \omega_{ijkl}(j) = 0 \}
\]

Examination of sets \( \alpha \) and \( \beta \) reveals that:

\[
|\alpha| = \binom{N-2}{k-2} \quad 2 \leq k \leq N \tag{30}
\]

and

\[
|\beta| = \binom{N-2}{k-1} \quad 1 \leq k \leq N \tag{31}
\]

To see this, notice that for set \( \alpha \) processors \( i \) and \( j \) request memory \( j \) leaving \( k-2 \) other processors out of a total of \( N-2 \) as possible requestors. Also, notice that for set \( \beta \) processor \( i \) requests memory \( j \) leaving \( k-1 \) other processors out of a total of \( N-2 \) as possible requestors. The total of \( N-2 \) is obtained by noting that for both sets \( \alpha \) and \( \beta \) the requests from processors \( i \) and \( j \) are fixed.

An event corresponding to any \( \omega_{ijkl} \in \alpha \) occurs with the following probability:

\[
\text{rpt}^{k-1}[1-\text{t}]^{N-k}
\]

since processor \( j \) requests its favorite memory, \( j \), with probability \( \text{rp} \); \( k-1 \) other processors (including \( i \)) make non-favorite requests to memory \( j \) with probability \( \text{t}^{k-1} \); and the \( N-k \) remaining processors do not request memory \( j \) with probability \( [1-\text{t}]^{N-k} \).

Similarly, an event corresponding to any \( \omega_{ijkl} \in \beta \) occurs with probability:

\[
(1-\text{rp})\text{t}^{k}(1-\text{t})^{N-k-1}
\]

Since processor \( j \) does not request memory \( j \) with probability \( 1-\text{rp} \); \( k \) other processors (including \( i \)) make non-favorite requests to \( j \) with probability \( \text{t}^{k} \); and the \( N-k-1 \) remaining processors do not request memory \( j \) with probability \( (1-\text{t})^{N-k-1} \).
Using the results of equations 30, 31, 32, 33, together with the fact that in the case we are considering $Pr[S_{ij}] = r_{i}p_{i}(j) = t$, allows us to simplify equation 15 to the following:

\[
Pr[A_{ij} | S_{ij}] = \sum_{k=1}^{N} \frac{1}{k} \binom{N-2}{k-2} r p t^{k-2} (1-t)^{N-k} + \sum_{k=1}^{N-1} \frac{1}{k} \binom{N-2}{k-1} (1-r p) t^{k-1} (1-t)^{N-k-1}
\]  

(34)

The summation over $k$ splits into two parts corresponding to the mutually exclusive events associated with sets α and β.

Equation 34 can be reduced to a closed form using some simple algebraic identities. The steps involved are outlined below:

Recall

\[
\binom{N-2}{k-2} = \binom{N-1}{k} - \binom{N-1}{k-1}
\]  

(35)

This equation holds for all integer values of $k$ provided we adopt the convention that the binomial coefficient $\binom{N}{k} = 0$ if $k < 0$ or if $k > N$ (see Knuth p.53 [Knu88]).

Equation 35 allows us to rewrite equation 34 as follows:

\[
Pr[A_{ij} | S_{ij}] = \sum_{k=2}^{N} \frac{1}{k} \binom{N-1}{k-1} r p t^{k-2} (1-t)^{N-k} - \sum_{k=2}^{N-1} \frac{1}{k} \binom{N-2}{k-1} r p t^{k-2} (1-t)^{N-k} + \sum_{k=1}^{N-1} \frac{1}{k} \binom{N-2}{k-1} (1-r p) t^{k-1} (1-t)^{N-k-1}
\]

Further manipulation of the binomial coefficients gives:

\[
Pr[A_{ij} | S_{ij}] = \frac{r p}{N t^2} \sum_{k=2}^{N} \binom{N}{k} t^k (1-t)^{N-k} - \frac{r p (1-t)}{(N-1) t^2} \sum_{k=2}^{N-1} \binom{N-1}{k} t^k (1-t)^{N-k-1} + \frac{(1-r p)}{(N-1) t} \sum_{k=1}^{N-1} \binom{N-1}{k} t^k (1-t)^{N-k-1}
\]

Recalling the binomial expansion of $[t+(1-t)]^N$ enables us to arrive at the following closed form:
\[ \Pr[A_{ij} \mid S_{ij}] = \frac{rp}{Nt^2} [1 - (1-t)^N - Nt(1-t)^{N-1}] \\
- \frac{rp(1-t)}{(N-1)t^2} [1 - (1-t)^N - (N-1)t(1-t)^{N-2}] + \frac{(1-rp)}{(N-1)t} [1 - (1-t)^N] \]

canceling terms reduces this to:

\[ \Pr[A_{ij} \mid S_{ij}] = \frac{rp}{Nt^2} [1 - (1-t)^N] + \frac{t-rp}{(N-1)t^2} [1 - (1-t)^N] \]  \hspace{1cm} (36)

Together equations 29 and 36 define \( \Pr[A_{ij} \mid S_{ij}] \) for all values of \( i \) and \( j \) in the case where processor \( i \) requests memory \( i \) as a favorite memory. Recall that for memory \( i \) to be a favorite \( p > \frac{1}{M} \), and that \( \lim_{p \to 1} \Pr[A_{ij} \mid S_{ij}] = 1 \). In addition, it can be shown, using l'Hopital's rule with equation 36, that \( \lim_{p \to 1} \Pr[A_{ij} \mid S_{ij}] = 0 \) for \( i \neq j \). This suggests that in the case of favorite memories the USR be abandoned and the requests from non-favorite processors be given higher priority than those from the favorite processor. For example, we could let:

\[ \Pr[A_{ij} \mid S_{ij}] = \begin{cases} 
q & \text{if } i = j \\
\frac{1-q}{k-1} & \text{if } i \neq j 
\end{cases} \]

and \( q < \frac{1}{k} \).

Furthermore, it can be shown that: \( E[BW]_{\text{favorite}} > E[BW]_{\text{uniform}} \). Finally, in the limiting case of \( p=1 \) it can be shown that: \( E[BW]_{\text{favorite}} = \min(N,M) \). This agrees with our definition of \( \text{Max}[BW] \) (see equation 15), which is consistent because in the case of \( p=1 \) all requests are exclusively between processors and their favorite memories, i.e. no conflicts occur.

Probabilistic Analysis of a Crossbar Switch
4. Simulation Results

Simulations were run in an effort to compare the analytical results with a more realistic system operation. It was found that the IRP assumptions, although not realistic, did not introduce significant differences.

The simulator is designed to model a system where processors emit requests according to their respective request distributions. Furthermore, once requests are emitted they must wait for a connection. That is, if processor i emits a request for memory j and it is not accepted the request is resubmitted on every succeeding cycle until the connection is granted. Each memory has associated with it N request fields where processors store requests. During each cycle each memory selects the requesting processor that gets the connection uniformly from all requests in its set of N request fields (empty request fields are ignored). The simulation results are shown in the Appendix.

**Definition 8:** Let \( N_W(i,j) \) be the number of cycles that processor i spends waiting for a requested connection to memory j.

**Definition 9:** Let \( N_R(i,j) \) be the number of cycles that processor i requests or uses memory j (i.e. it is the number of cycles during which processor i references memory j).

**Definition 10:** Let \( WF(i,j) \) be the waiting fraction for processor i requesting memory j. That is, it is the fraction of cycles that processor i spends waiting for memory j, relative to the number of requests processor i generates for memory j.

Then, from definitions 8, 9, and 10 it follows that:

\[
WF(i,j) = \frac{N_W(i,j)}{N_R(i,j)}
\]

One would expect that:

Probabilistic Analysis of a Crossbar Switch
\[ WF(i,j) \approx 1 - \Pr\{A_{ij} \mid S_{ij}\} \]

That is, the number of cycles spent waiting is approximately the probability of not having a request accepted, multiplied by the number of references to that memory. From simulation runs for many request distributions this approximation has been found to hold within about 10%.

Now notice that if for example \( WF(i,j) = \frac{2}{3} \) then it may be seen that the average waiting time is 2 while the number of requests is 3. That is, for each request for memory \( j \), on the average processor \( i \) must wait for 2 cycles and then use the connection on the third.

**Definition 11:** Let \( E[W(i,j)] \) be the average waiting time of processor \( i \) for memory \( j \). Then, from the above discussion it follows that:

\[
\frac{E[W(i,j)]}{E[W(i,j)] + 1} = WF(i,j) \approx 1 - \Pr\{A_{ij} \mid S_{ij}\}
\]

So,

\[
E[W(i,j)] \approx \frac{1 - \Pr\{A_{ij} \mid S_{ij}\}}{\Pr\{A_{ij} \mid S_{ij}\}}
\]

From this simple approximate analysis the average waiting time may be found.

Note some special cases:

\[ \Pr\{A_{ij} \mid S_{ij}\} = \frac{1}{2} \Rightarrow E[W(i,j)] = 1 \]

\[ \Pr\{A_{ij} \mid S_{ij}\} = \frac{1}{3} \Rightarrow E[W(i,j)] = 2. \]

The limiting behavior of the waiting time is:
\[
\lim_{N \to \infty} E[W] = \frac{1 - \frac{1}{ra} [1 - e^{-ra}]}{\frac{1}{ra} [1 - e^{-ra}]} = \frac{ra}{1 - e^{-ra}} - 1
\]

If \( ra = 1 \), then \( \lim_{N \to \infty} E(W) = \frac{1}{e-1} < 1 \)

All simulations were run with lengths of 100,000 system cycles the Appendix contains a tabular summary of the simulation data. From earlier experiments it was found that as \( r_1 \) decreases the runs become less stable unless the number of simulation cycles is increased over 100,000. Thus \( r_1 = .1 \) is about the minimum value used for our simulations.

Examination of the simulation results shows that for the most part \( E[BW] > E[BW]_{\text{sim}} \). This effect was previously noted in [CKL77] for the case when \( r = 1 \), and is caused by the simulator resubmitting blocked requests until they are finally accepted. As noted earlier, resubmission of blocked requests is more realistic than our model which operates under the IRP assumptions. Thus, we can consider our model as yielding slightly optimistic results. Notice that the above discrepancy essentially disappears when \( r \) becomes small.
6. Conclusion

In conclusion, the analysis presented above allows one to compute various performance figures for an N-M crossbar. The simulation results show that these performance figures are a close approximation to the actual behavior of the crossbar provided the request distributions are an accurate model of the request behavior of the processors.
6. References

[BaS78]  

[Bha75]  

[BrD77]  

[CiS81]  

[CKL77]  

[Clo63]  

[Fra80]  

[Knu68]  

[Law75]  

[MaM81]  

[Pat79]  

[Pea77]  

[Ravi72]  

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[Str70]  

[Tho80]  

[Ben65]  

[WuF80]  
7. Appendix

The following is a summary of simulation data for several system sizes and request rates/distributions.

<table>
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<th>N</th>
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<td>2.63</td>
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</table>

where,

- \( N \) - the number of processors
- \( M \) - the number of memory banks
- \( r \) - the processor request rate
- \( 1-pa \) - \( \text{Pr} \{ A_i | S_i \} \)
- \( WF \) - the simulation waiting fraction,
  - it is the average over three runs
- \( bw1 \) - the simulation bandwidth, for the first run
- \( bw2 \) - the simulation bandwidth, for the second run
- \( bw3 \) - the simulation bandwidth, for the third run
- \( E[BW] \) - the computed average bandwidth
- \( E[BW] \text{sim} \) - the average simulation bandwidth, it is the average of \( bw1, bw2, bw3 \)
- \( E[W] \) - the computed average waiting time

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<table>
<thead>
<tr>
<th>N</th>
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<th>p</th>
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<th>pamin</th>
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These simulations were run using request distributions in which each processor requests a favorite memory with probability "p" and the remaining uniformly.

- **pamax** = Pr{A_m | S_m} for the preferred memory, and any memory preferred by no processor
- **pamin** = Pr{A_m | S_m} for any other memory
- **wfrmin** = WF, from the three simulations, for the preferred memory
- **wfrmax** = WF, from the three simulations, for any other memory

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<table>
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Since this is an 8-4 connector, request "compression" must take place. In these distributions, two processors prefer a given memory (with magnitude "p") and their remaining probability distribution is uniform over all other memories. There are then three other pairs of processors in a similar situation. Thus each memory is preferred by two processors.

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