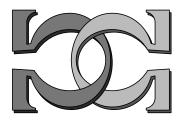
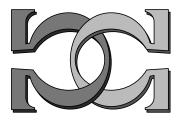


Update Networks and Their Routing Strategies

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CDMTCS-127 March 2000



Centre for Discrete Mathematics and Theoretical Computer Science

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Abstract. We introduce the notion of update networks to model communication networks with infinite duration. In our formalization we use bipartite finite graphs and game-theoretic terminology as an underlying structure. For these networks we exhibit a simple routing procedure to update information throughout the nodes of the network. We also introduce an hierarchy for the class of all update networks and discuss the complexity of some natural problems.

1 Introduction

A network can be viewed as a finite directed graph whose nodes represent a type of system (e.g., processors, data repositories, and servers) and edges represent communication channels between the nodes. The process of communication in a network is essentially an infinite duration process. The interacting network has to be robust (e.g., be fault-tolerant and free of dead-locks). We are interested in those networks that possess liveness and fairness properties. Liveness means that all nodes are actively participating in the collective system. Fairness means that critical information of the dynamic network is distributed to all of the nodes.

To give an example, suppose we have data stored on each node of a network and we want to continuously update all nodes with consistent data. For instance, we are interested in addressing redundancy issues in distributed databases. Often one requirement is to share key information between all nodes of the distributed database. We can do this by having a data packet of current information continuously go through all nodes of the network. We call a network of this type an update network.

The operation of an update packet over time enters only a finite number of nodes and produces an infinite sequence v_1, v_2, \ldots , called a *run-time sequence*. Since the number of nodes is finite, some of the nodes, called *persistent nodes*, appear infinitely often in the run-time sequence. The success of a run-time sequence is determined by whether or not the sequence of nodes satisfies the liveness and fairness properties. That is, the set of persistent nodes coincides with the set of all nodes of the network. We can regard the run-time sequences as plays of a two-player game where one player, called *Survivor*, tries to ensure that persistent states span the network and the other player, called *Adversary*, does not.

One possible model for an update network is thus based on a finite directed graph as the underlying structure for games between Survivor and Adversary. We formalize this in the following definition, where for this paper we restrict our attention to the bipartite case. **Definition 1.** An update (bipartite) game is a bipartite graph G = (V, E) with partite sets S and A. There are two players, called Survivor and Adversary, which control vertices S and A, respectively. We stipulate that the out-degree of every vertex is nonzero.

The game rules allow moves along edges of G. Each *play* of an infinite duration game is a sequence of vertices $v_0, v_1, \ldots, v_i, \ldots$ such that the game rules are followed. We call a finite prefix sequence of a play a *history*. We say that a vertex v is *visited* if it occurs in the play. Note that either Survivor or Adversary may begin the play. Survivor wins a play if the persistent vertices of the play is the set V, otherwise Adversary wins. A *strategy* for Survivor is a function H from play histories v_0, \ldots, v_i , where $v_i \in S$, to the set A such that $(v_i, H(v_0, \ldots, v_i))$ is an edge. A strategy for Advisory is similarly defined.

A given strategy for a player may either win or lose the game when starting at an *initial vertex* v_0 , where $v_0 \in V$. A player's *winning strategy* for an initial vertex is one that wins no matter what the other player does.

Example 1. In Figure 1 we present a game \mathcal{G} , where $S = \{1, 3, 4\}$, and $A = \{0, 2, 5\}$. Here Adversary has a winning strategy from any starting node. If whenever node 0 is reached, Adversary moves to node 4. Thus, node 1 is only reached at most once during any play.

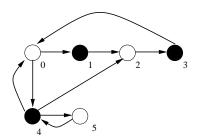


Fig. 1. Example of an update game.

We now formally define update networks.

Definition 2. An update game is an update network if Survivor has a winning strategy for every initial vertex.

Update networks can also be viewed as message passing networks where only Survivor's nodes actively participate in the updating process, while Adversary's nodes only need to passively forward a message.

Example 2. The graph displayed below in Figure 2 is an update network. Here, no matter what Adversary does node 0 is visited. Survivor then can alternate moving between nodes 2 and 3 to span the network.

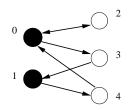


Fig. 2. A simple example of an update game which is an update network.

We end this section with a few related references. Previous work on twoplayer infinite duration games on finite bipartite graphs is presented in the paper by McNaughton [2] and extended by Nerode *et al.* [3]. More recently, update networks modeled by arbitrary digraphs were introduced in [1]. Also several earlier papers that deal with finite duration games on automata and graphs have appeared (e.g., see [4, 5]).

2 Constructing Update Networks

We now want to present several construction techniques for building update networks from smaller update networks. We introduce three primitive operations below for update networks $G_1 = (S_1, A_1, E_1)$ and $G_2 = (S_2, A_2, E_2)$.

- **Enrichment:** Take G_1 and add edges E' to E_1 where all new edges are directed from S_1 to A_1 . The resulting graph is called an *enrichment* of G_1 , denoted $G_1 \cup E'$.
- **Identification:** Take G_1 and G_2 and subsets T_1 of S_1 and T_2 of S_2 with the same cardinality. Let f be a bijection between T_1 and T_2 . Build the graph $(S_1 \cup S_2 \setminus T_2, A_1 \cup A_2, E_1 \cup E)$ where E is E_2 with all vertices in T_2 replaced with their image under f. This graph is called the *identification of* G_1 and G_2 under f, denoted by $G_1 \oplus_f G_2$.
- **Extension:** Take G_1 and G_2 then construct the graph $(S = S_1 \cup S_2, A = A_1 \cup A_2, E_1 \cup E_2 \cup E)$ where E is a new set of edges directed from S to A with at least one edge from S_1 to A_2 and at least one edge from S_2 to A_1 . This graph is called the *extension of* G_1 by G_2 via E, denoted $G_1 \cup G_2 \cup E$.

It is easy to observe the following.

Proposition 1. The enrichment, identification, extension operators applied to update networks yield update networks.

Example 3. The graphs displayed below in Figure 3 are obtained by applying the operations of enrichment, join and extension to the update networks B_1 and B_2 . The black and white vertices are Survivor and Adversary nodes, respectively.

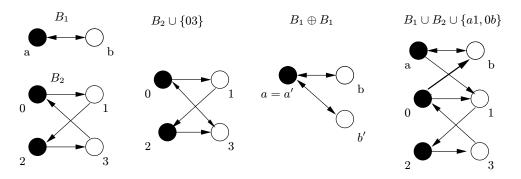


Fig. 3. Illustrating the operations of enrichment, join and extension.

3 Deciding Update Networks

In this section we provide an important operation, called the contraction operation, that applied to update games produces update networks if and only if the original update game is an update network. The operation reduces the size of the underlying bipartite graph, and hence produces a decision procedure for checking if a given update game is an update network.

We can easily characterize those bipartite update networks with only one Survivor vertex. These are the bipartite graphs where out-degree(s) = |A| for the single Survivor vertex s. We now mention several properties for all bipartite update networks [1].

Lemma 1. If $(V = A \cup S, E)$ is a bipartite update network then for every vertex $s \in S$ there exists at least one $a \in A$ such that $(a, s) \in E$ and out-degree(a) = 1

Proof. Let s be a vertex that does not satisfy the statement of the lemma. Let $A_s = \{a \mid (a, s) \in E\}$. Then out-degree(a) > 1 for all $a \in A_s$. Adversary has the following winning strategy. If the play history ends in a then since out-degree(a) > 1, Adversary moves to s', where $s' \neq s$ and $(a, s') \in E$. This contradicts the assumption of lemma that we have an update network. \Box

For any Survivor vertex s define $Forced(s) = \{a \mid out-degree(a) = 1 \text{ and } (a, s) \in E\}$, which denotes the set of Adversary vertices that are 'forced' to move to s.

Lemma 2. If B is a bipartite update network such that |S| > 1 then for every $s \in S$ there exists an $s' \neq s$ and an $a \in Forced(s)$, such that (s', a, s) is a directed path.

Proof. If B has more than one vertex in S then there must be a strategy for Survivor to create a play history to visit vertex s from some other vertex $s' \in S$. To do this we need a forced Adversary vertex a (of A) in the neighborhood of s'. There exists such a vertex a by Lemma 1.

Definition 3. Given a bipartite graph $(S \cup A, E)$ a forced cycle is a (simple) cycle $(a_k, s_k, \ldots, a_2, s_2, a_1, s_1)$ for $a_i \in Forced(s_i)$ and $s_i \in S$.

The following lemma is the main ingredient in characterizing bipartite update networks.

Lemma 3. If B is a bipartite update network such that |S| > 1 then there exists a forced cycle of length at least 4.

Proof. Take $s_1 \in S$. From Lemma 2 there exists a path (s_2, a_1, s_1) in B such that $s_2 \neq s_1$ and $a_1 \in \text{Forced}(s_1)$. Now for s_2 we apply the lemma again to get a path (s_3, a_2, s_2) in B such that $s_3 \neq s_2$ and $a_2 \in \text{Forced}(s_2)$. If $s_3 = s_1$ we are done. Otherwise repeat Lemma 2 for vertex s_3 . If $s_4 \in \{s_1, s_2\}$ we are done. Otherwise repeat the lemma for s_4 . Eventually $s_i \in \{s_1, s_2, \ldots, s_{i-2}\}$ since B is finite.

Note if B does not have a forced cycle of length at least 4 then either |S| = 1 or B is not a bipartite update network. That is, if |S| > 1 then Adversary has a strategy to not visit a vertex s_2 of S whenever the play begins at some different vertex s_1 of S.

We now define a contraction operator for reducing the size of an update game. Let $B = (S \cup A, E)$ be a bipartite update game with a forced cycle $C = (a_k, s_k, \ldots, a_2, s_2, a_1, s_1)$ of length at least 4. The contraction operator applied to B via C, denoted B/C, produces an update game $B' = (S' \cup A', E')$ with |S'| < |S|. We construct B' as follows. For new vertices a and s,

$$S' = (S \setminus \{s_1, s_2, \dots, s_k\}) \cup \{s\} \text{ and } A' = (A \setminus \{a_1, a_2, \dots, a_k\}) \cup \{a\}$$

and

$$E' = E(B \setminus \{s_1, a_1, \dots, s_k, a_k\}) \cup \{(s, a') \mid a' \in A' \text{ and } (s_i, a') \in E, \text{ for some } i \leq k\} \cup \{(a', s) \mid a' \in A' \text{ and } (a', s_i) \in E, \text{ for some } i \leq k\} \cup \{(s', a) \mid s' \in S' \text{ and } (s', a_i) \in E, \text{ for some } i \leq k\} \cup \{(a, s), (s, a)\}.$$

We now present a method that helps us decide if a bipartite game is a bipartite update network.

Lemma 4. If $B = (S \cup A, E)$ is a bipartite update game with a forced cycle C of length at least 4 then B is a bipartite update network if and only if B' = B/C is one.

Proof (sketch). We show that if B' is an update network then B is also an update network. We first define the natural mapping p from vertices of B onto vertices of B' by

$$p(v) = v \quad \text{if} \quad v \notin C$$

$$p(v) = a \quad \text{if} \quad v \in C \cap S$$

$$p(v) = s \quad \text{if} \quad v \in C \cap A.$$

Then any play history of B is mapped, via the function p(v) = v', onto a play history of B'. Consider a play history v_0, v_1, \ldots, v_n of B that starts at vertex v_0 . Let f' be a winning strategy for Survivor when the game begins at vertex v'_0 . We use the mapping p to construct Survivor's strategy f in game B by considering the two cases $v'_n = s$ and $v'_n \neq s$. It is not hard to construct a winning strategy f for Survivor in game B whenever f' is a winning strategy in B'.

By a similar case study one can show that if B is an update network then B' is also an update network (see [1] for details).

With respect to the above proof, Figure 4 shows how a forced cycle of B is reduced to a smaller forced cycle (of length 2) in B'.

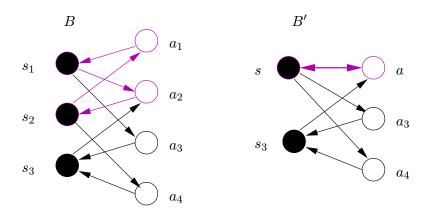


Fig. 4. Showing the bipartite update game reduction of Lemma 4.

Theorem 1. There exists an algorithm that decides whether a bipartite update game B is a bipartite update network in time $O(n \cdot m)$, where n and m are the order and size of the underlying graph.

Proof. We show that finding a cycle that is guaranteed to exist by Lemma 3 takes time at most O(m) and that producing B' from B in Lemma 4 takes time at most O(n + m). Since we need to recursively do this at most n times the overall running time is shown to be $O(n \cdot m)$.

The algorithm terminates whenever a forced cycle of length at least four is not found. It decides whether the current bipartite graph is a update network by simply checking that $S = \{s\}$ and out-degree(s) = |A|. That is, the singleton Survivor vertex is connected to all Adversary vertices.

Let us analyze the running time for finding a forced cycle C. Recall the algorithm begins at any vertex s_1 and finds an in-neighbor a_1 (of s_1) of outdegree 1 with $(s_2, a_1) \in E$ where $s_2 \neq s_1$. This takes time proportional to the number of edges incident to s_1 to find such a vertex a_1 . Repeating with s_2 we find an a_2 in time proportional to the number of edges into s_2 , etc. We keep a boolean array to indicate which s_i are in the partially constructed forced path (i.e., the look-up time will be constant time to detect a forced cycle of length at least 4). The total number of steps to find the cycle is at most a constant factor time the number of edges in the graph. Finally, we can observe that building B' from B and C of Lemma 4 runs in linear time by the definition of S', A' and E'. Note that if the data structure for graphs is taken to be adjacency lists then E' is constructed by copying the lists of E and replacing one or more vertices s_i 's or a_j 's with one s or a, respectively.

The above result indicates the structure of bipartite update networks. These are basically connected forced cycles, with possibly other legal moves for some of Survivor and Adversary vertices.

4 Level Structures of Update Networks

In this section we introduce a hierarchy of update networks which can be used to extract 'efficient' winning strategies for Survivor. The class H_0 contains all update networks with exactly one Survivor node. Assume that the class H_k has been defined. Then an update network G belongs to H_{k+1} if and only if Gcan be contracted (via some forced cycle) to a G' such that $G' \in H_k$. Thus, by Lemma 4, G is an update network if and only if there exists a sequence $G = G_0, G_1, \ldots, G_k$ of update networks and forced cycles $C_0, C_1, \ldots, C_{k-1}$ such $G_k \in H_0$ and G_i is obtained by the contraction operator applied to G_{i-1} via C_i .

Definition 4. The level number k of an update network G is the minimum number of contractions of forced cycles needed to produce an update network with one Survivor node, that is $G \in H_k$ and $G \notin H_{k-1}$.

Intuitively, the level number represents a complexity of update networks. Thus, given an update network it is quite natural to compute the level number. Computing the level number, in turn, is related to finding a large forced cycles in the network. One approach for estimating this level number can be based on a greedy algorithm that repeatedly finds a largest forced cycle and applies the contraction operator. However the following proposition shows that this approach is not efficient.

Proposition 2. To find the longest forced cycle in an update game is NPcomplete.

Proof. Let LONGFORCED be the decision problem to decide if an update game has a forced cycle of length at least k. We know that deciding, for input a digraph G and integer k, whether G has a cycle of length at least k, called LONGCYCLE, is **NP**-complete. We can polynomial-time many-one reduce the input (G, k) of LONGCYCLE to an input (G', 2k) for LONGFORCED as follows. Let G' be G with every edge subdivided, that is for every edge (u, v) in G is replaced with edges (u, a) and (a, v), where a is a new vertex. Clearly G' is a bipartite graph where the new vertices correspond to the Adversary set A. If G'has a forced cycle of length 2k then G has a cycle of length at least k. Obviously, the other direction holds. It turns out that, in general, the greedy algorithm does not compute the level number of update networks, as illustrated with the networks shown in Figure 5. Here, the graph has three forced cycles, one small one of length 4 and two large ones of length at least 5. The greedy algorithm would use 4 contractions, while the level number is actually 2.

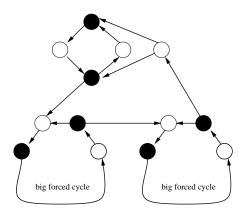


Fig. 5. Why the greedy algorithm does not find the smallest level number.

5 Simple Update Strategies

We now want to consider how to design Survivor winning strategies when we have an update network.

From the previous section we see that one way for constructing winning strategies is to explicitly apply the contraction operation repeatedly while remembering the sequence of forced cycles. From Lemma 4 a sequence G, G_1, \ldots, G_k of update networks is obtained from the original network G. Here the final network G_k with one survival node has a trivial winning strategy, where it moves repeatedly in a cyclic fashion to all of its neighbors. The proof of the lemma indicates how to compute a winning strategy for G_i from the winning strategy for G_{i+1} . (Recall, a forced cycle of length 2 in G_{i+1} is uncontracted to a forced cycle of length at least 4 in G_i .) We call strategies of this type contraction-based strategies.

Example 4. Consider the update network G' of Figure 4. If we apply the contraction algorithm again on the forced cycle (a_3, s_3, a, s) we get an update network with one Survivor node s' that has a winning strategy to alternate sending to the Adversary nodes a' and a_4 . Doing an uncontraction yields a winning strategy for G' where node s alternates sending to a_3 and a_4 and node s_3 always sends to a.

It turns out there is another simple local winning strategy (which may not be as efficient at updating all the nodes) that only requires each Survivor node to systematically update their neighbors. This strategy needs only to follow simple local rules with out any knowledge of the topology of the network. We now formally present this *cyclic neighbor strategy*. Each Survivor node *s* keeps an index i_s between 0 and d_s =out-degree(u)-1, initially set to 0. Survivor's next move from node *s* is to its i_s -th out-neighbor and i_s is incremented modulo d_s . We now show this is a winning strategy.

Theorem 2. The cyclic neighbor strategy is a winning strategy for any update network.

Proof. Consider any play consistent with the cyclic neighbor strategy for an update network G. There will be a set of persistent vertices called P. If $P = A \cup S$ then we are done. Now assume otherwise and consider the set $N = (A \cup S) \setminus P$. There is at least one Adversary node in N by Lemma 1 (an adversary vertex of out-degree 1 can not be in P if it's out-neighbor is in N). If we follow the cyclic neighbor strategy there must not be any edges from $P \cap S$ to $N \cap A$, since all vertices in P were reached infinitely often. Thus Adversary has a winning strategy to keep out of N forever (no matter what strategy Survivor uses). This contradicts the fact that G was an update network.

The above proof leads to the following more general winning strategy.

Corollary 1. Any Survivor strategy for an update network that from any $s \in S$ moves to each neighbor of s infinitely often is a winning strategy.

One can compare the cyclic neighbor strategies with the contraction-based strategies. In general the contraction-based strategies are more efficient at updating all nodes of a network more frequently. However, there is a quadratic cost of having to compute these. However, like the cyclic neighbor strategies, there is very little operational cost in using a contraction-based strategy. The only requirement is that a sequence of neighbors (with possible repetitions) has to be kept at each node instead of just a single counter.

We end this section with a couple comments about Adversary winning strategies for update games that are not update networks. The decision algorithm fails because there are no forced cycles of length at least 4 when there remains at least two Survivor nodes. In any case, Adversary knows which Survivor nodes to avoid. This strategy is a counter strategy to Survivor's contraction-based strategy mentioned above.

6 Conclusion

In this paper we have presented a game-theoretic model for studying update networks. We have shown that it is algorithmically feasible to recognize update networks. That is, we have provided an algorithm which solves the update game problem in $O(n \cdot m)$ time. Moreover, our algorithm for the case of bipartite update games is used to give a characterization of bipartite update networks. We have presented two simple routing strategies for updating these networks. We have also discussed the complexity of update networks in terms of the level number invariant. In general, update networks with the lower level numbers have simpler and more efficient updating strategies.

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