Reasoning about Synchronic Groups

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Abstract

Swarm is a computational model which extends the UNITY model in three important ways: (1) UNITY’s fixed set of variables is replaced by an unbounded set of tuples which are addressed by content rather than by name; (2) UNITY’s static set of statements is replaced by a dynamic set of transactions; and (3) UNITY’s static ||-composition is augmented by dynamic coupling of transactions into synchronic groups. This last feature, unique to Swarm, facilitates formal specification of the mode of execution (synchronous or asynchronous) associated with portions of a concurrent program and enables computations to restructure themselves so as to accommodate the nature of the data being processed and to respond to changes in processing objectives. This paper overviews the Swarm model, introduces the synchronic group concept, and illustrates its use in the expression of dynamically structured programs. A UNITY-style programming logic is given for Swarm, the first axiomatic proof system for a shared dataspaces language.
1 Introduction

Attempts to meet the challenges of concurrent programming have led to the emergence of a variety of models and languages. Chandy and Misra, however, argue that the fragmentation of programming approaches along the lines of architectural structure, application area, and programming language features obscures the basic unity of the programming task [3]. With the UNITY model, their goal is to unify seemingly disparate areas of programming with a simple theory consisting of a model of computation and an associated proof system.

Chandy and Misra build the UNITY computational model upon a traditional imperative foundation, a state-transition system with named variables to express the state and conditional multiple-assignment statements to express the state transitions. Above this foundation, however, UNITY follows a more radical design: all flow-of-control and communication constructs have been eliminated from the notation. A UNITY program begins execution in a valid initial state and continues infinitely; at each step an assignment is selected nondeterministically, but fairly, and executed atomically.

To accompany this simple but innovative model, Chandy and Misra have formulated an assertional programming logic which frees the program proof from the necessity of reasoning about execution sequences. Unlike most assertional proof systems, which rely on the annotation of the program text with predicates, the UNITY logic seeks to extricate the proof from the text by relying upon proof of program-wide properties such as invariants and progress conditions.

Swarm [14] is a model which extends UNITY by permitting content-based access to data, a dynamic set of statements, and the ability to prescribe and alter the execution mode (i.e., synchronous or asynchronous) for arbitrary collections of program statements. The Swarm model is our primary vehicle for study of the shared dataspace paradigm, a class of languages and models in which the primary means for communication among the concurrent components of a program is a common, content-addressable data structure called a shared dataspace. Elements of the dataspace may be examined, inserted, or deleted by programs. Linda [2], Associations [13], and production rule languages such as OP5 [1] all follow the shared dataspace approach.

In designing Swarm, we merged the philosophy of UNITY with the methods of Linda. Swarm has a UNITY-like program structure and computational model and Linda-like communication mechanisms. We partition the Swarm dataspace into three subsets: a tuple space (a finite set of data tuples), a transaction space (a finite set of transactions), and a synchrony relation (a symmetric relation on the set of all possible transactions). We replace UNITY’s fixed set of variables with a set of tuples and UNITY’s fixed set of assignment statements with a set of transactions.

A Swarm transaction denotes an atomic transformation of the dataspace. It is a set of concurrently executed query-action pairs. A query consists of a predicate over the dataspace; an action consists of a group of deletions and insertions of dataspace elements. Instances of transactions may be created dynamically by an executing program.

A Swarm program begins execution from a specified initial dataspace. On each execution step, a transaction is chosen nondeterministically from the transaction space and executed atomically. This selection is fair in the sense that every transaction in the transaction space at any point in the computation will eventually be chosen. An executing transaction examines the dataspace and then, depending upon the results of the examination, can delete tuples (but not transactions) from the dataspace and insert new tuples and transactions into the dataspace. Unless a transaction explicitly reinserts itself into the dataspace, it is deleted as a by-product of its execution. Program execution continues until there are no transactions in the dataspace.

The synchrony relation feature adds even more dynamism and expressive power to Swarm programs. It is a relation over the set of possible transaction instances. This relation may be examined and modified by programs in the same way as the tuple and transaction spaces are. To accommodate the synchrony relation, we extend the program execution model in the following way: whenever a transaction is chosen for execution, all transactions in the transaction space which are related to the chosen transaction by (the closure of) the synchrony relation are also
chosen; all of the transactions that make up this set, called a synchronic group, are executed as if they comprised a single transaction.

By enabling asynchronous program fragments to be coalesced dynamically into synchronous subcomputations, the synchrony relation provides an elegant mechanism for structuring concurrent computations. This unique feature facilitates a programming style in which the granularity of the computation can be changed dynamically to accommodate structural variations in the input. This feature also suggests mechanisms for the programming of a mixed-modality parallel computer, i.e., a computer which can simultaneously execute asynchronous and synchronous computations. Perhaps architectures of this type could enable both higher performance and greater flexibility in algorithm design.

In this paper we show how to add this powerful capability to Swarm without compromising our ability to formally verify the resulting programs. The presentation is organized as follows. Section 2 reviews the basic Swarm notation. Section 3 introduces the notation for the synchrony relation and discusses the concept of a synchronic group. Section 4 reviews a UNITY-style assertional programming logic for Swarm without the synchrony relation and then generalizes the logic to accommodate synchronic groups. Section 5 illustrates the use of synchronic groups by means of a proof of an array summation program. Section 6 discusses some of the rationale for our design decisions.

2 The Swarm Notation

By choosing the name Swarm for our shared dataspace programming model, we evoke the image of a large, rapidly moving aggregation of small, independent agents cooperating to perform a task. In this section we introduce a notation for programming such computations. We first present an algorithm expressed in a familiar imperative notation—a parallel dialect of Dijkstra's Guarded Commands [5] language. We then construct a Swarm program with similar semantics.

The program fragment given in Figure 1 (similar to the one given in [9]) sums an array of \( N \) integers. For simplicity of presentation, we assume that \( N \) is a power of 2. We also assume that elements 1 through \( N \) of the constant array \( A \) contain the values to be summed. The program introduces variables \( z \), an \( N \)-element array of partial sums, and \( j \), a control variable for the summing loop. The preamble of the loop initializes \( z \) to equal \( A \) and \( j \) to equal 1. Thus, the precondition \( Q \) of the program's loop is the assertion

\[
pow(2, N) \land j = 1 \land \forall i : 1 \leq i \leq N :: x(i) = A(i)
\]

where

\[
pow(2, k) \equiv (\exists p : p \geq 0 :: k = 2^p).
\]

At termination, \( z(N) \) is required to contain the sum of the \( N \) elements of \( A \). Thus the postcondition \( R \) of the program is the assertion

\[
z(N) = \text{sum}_A(0, N)
\]

where

\[
\text{sum}_A(l, u) = (\Sigma k : l < k \leq u :: A(k)).
\]

The loop computes the sum in a tree-like fashion as shown in the diagram: adjacent elements of the array are added in parallel, then the same is done for the resulting values, and so forth until a single value remains. The construct

\[
(\| k : \text{predicate} :: \text{assignment})
\]

is a parallel assignment command. The assignment is executed in parallel for each value of \( k \) which satisfies the predicate; the entire construct is performed as one atomic action. An invariant \( P \) for the program's loop is the assertion
\[ x(1) \quad x(2) \quad x(3) \quad x(4) \quad x(5) \quad x(6) \quad x(7) \quad x(8) \]

\[ j = 1 \]

\[ \Sigma_1^1 \quad \Sigma_1^2 \quad \Sigma_3^3 \quad \Sigma_1^4 \quad \Sigma_5^5 \quad \Sigma_5^6 \quad \Sigma_7^7 \quad \Sigma_7^8 \]

\[ j = 2 \]

\[ \Sigma_1^1 \quad \Sigma_1^2 \quad \Sigma_3^3 \quad \Sigma_1^4 \quad \Sigma_5^5 \quad \Sigma_5^6 \quad \Sigma_7^7 \quad \Sigma_7^8 \]

\[ j = 4 \]

\[ \Sigma_1^1 \quad \Sigma_1^2 \quad \Sigma_3^3 \quad \Sigma_1^4 \quad \Sigma_5^5 \quad \Sigma_5^6 \quad \Sigma_7^7 \quad \Sigma_7^8 \]

\[ j : \text{integer}; \]
\[ x(i : 1 \leq i \leq N) : \text{array of integer}; \]

\[ j := 1; \]
\[ \{ k : 1 \leq k \leq N :: x(k) := A(k) \}; \]
\[ \{ Q \} \]
\[ \text{do } j < N \rightarrow \{ P \} \]
\[ \quad \{ k : 1 \leq k \leq N \land k \mod (j + 2) = 0 :: x(k) := x(k) + x(k) \}; \]
\[ j := j + 2 \]
\[ \text{od} \]
\[ \{ R \} \]

Figure 1: A Parallel Array-Summation Algorithm using Guarded Commands
\[ \text{pow2}(N) \land \text{pow2}(j) \land 1 \leq j \leq N \land \forall i : \text{node}(i, j) \equiv x(i) = \text{sum}_A(i-j, i) \]

where

\[ \text{node}(k, l) \equiv (1 \leq k \leq N \land k \mod l = 0). \]

(Clearly \( Q \implies P \) and \( P \land j \geq N \implies R. \) The integer function \( \frac{N}{j} - 1 \) (call it \( \text{vf} \)) is an appropriate variant function to show the termination of the loop.

Swarm is a shared dataspace programming model. Instead of expressing a computation in terms of a group of named variables, Swarm uses a set of tuples stored in a dataspace. Each tuple is a pairing of a type name with a finite sequence of values; a program accesses a tuple by its content—type name and values—rather than by a specific name or address. Swarm programs execute by deleting existing tuples from and inserting new tuples into the dataspace. The transactions which specify these atomic dataspace transformations consist of a set of query-action pairs executed in parallel. Each query-action pair is similar to a production rule in a language like OPS5 [1].

How can we express the array-summation algorithm in Swarm? To represent the array \( x \), we introduce tuples of type \( x \) in which the first component is an integer in the range 1 through \( N \), the second a partial sum. We can express an instance of the array assignment in the do loop as a Swarm transaction in the following way:

\[ v_1, v_2 : x(k-j, v1), x(k, v2) \rightarrow x(k, v2), x(k, v1+v2) \]

The part to the left of the \( \rightarrow \) is the query; the part to the right is the action. The identifiers \( v_1 \) and \( v_2 \) designate variables that are local to the query-action pair. (For now, assume that \( j \) and \( k \) are constants.)

The execution of a Swarm query is similar to the evaluation of a rule in Prolog [16]. The above query causes a search of the dataspace for two tuples of type \( x \) whose component values have the specified relationship—the comma separating the two tuple predicates is interpreted as a conjunction. If one or more solutions are found, then one of the solutions is chosen nondeterministically and the matched values are bound to the local variables \( v_1 \) and \( v_2 \) and the action is performed with this binding. If no solution is found, then the transaction is said to fail and none of the specified actions are taken.

The action of the above transaction consists of the deletion of one tuple and the insertion of another. The \( \dagger \) operator indicates that the tuple \( x(k, v2) \), where \( v2 \) has the value bound by the query, is to be deleted from the dataspace. The unmarked tuple form \( x(k, v1+v2) \) indicates that the corresponding tuple is to be inserted. Although the execution of a transaction is atomic, the effect of an action is as if all deletions are performed first, then all insertions.

The parallel assignment command of the algorithm can be expressed similarly in Swarm:

\[
\begin{align*}
[[ & k : 1 \leq k \leq N \land k \mod (j \ast 2) = 0 :: \\
& v_1, v_2 : x(k-j, v1), x(k, v2) \rightarrow x(k, v2), x(k, v1+v2) ]
\end{align*}
\]

We call each individual query-action pair a \textit{subtransaction} and the overall parallel construct a \textit{transaction}. As with the parallel assignment, the entire transaction is executed atomically. The cumulative effect of executing a transaction is as if the subtransactions are executed synchronously: all queries are evaluated first, then the indicated tuples are deleted, and finally the indicated tuples are inserted.

In Swarm there is no concept of a process and there are no sequential programming constructs or recursive function calls. Only transactions are available. Like data tuples, transactions are represented as tuple-like entities in the dataspace. A transaction has a type name and a finite sequence of values called parameters. Transaction instances can be queried and inserted in the same way that data tuples are, but cannot be explicitly deleted. A Swarm dataspace thus has two components, the tuple space and the transaction space.

We model the execution of a Swarm program in the following way. The program begins execution with the specified initial dataspace. On each execution step, a transaction is chosen
program ArraySum($N,A : [3p:p \geq 0 :: N = 2^p],A(i:1 \leq i \leq N))$

tuple types
\[ [i,s:1 \leq i \leq N :: z(i,s)] \]

transaction types
\[ j : j > 0 :: \]
\[ \text{Sum}(j) \equiv \]
\[ \begin{array}{c}
\[ [\]
\[ || k : 1 \leq k \leq N \land k \mod (j \ast 2) = 0 :: \]
\[ v_1,v_2 : z(k-j,v_1),z(k,v_2) \rightarrow z(k,v_1+v_2) ] \]
\[ || j < N \rightarrow \text{Sum}(j \ast 2) \]
\end{array} \]

initialization
\[ \text{Sum}(1); [i : 1 \leq i \leq N :: z(i,A(i))] \]

end

Figure 2: A Parallel Array-Summation Program in Swarm

nondeterministically from the transaction space and executed atomically. This selection is fair
in the sense that every transaction present in the transaction space at any point in time must
eventually be executed (i.e., weak fairness [7]). Unless the transaction explicitly
reinserts itself into
the transaction space, it is deleted as a by-product of its own execution—regardless of the success
or failure of its component queries. Program execution continues until there are no transactions
remaining in the transaction space.

Two aspects of the array-summation program’s do command have not been expressed in
Swarm—the doubling of $j$ and the conditional repetition of the loop body. Both of these actions
can be incorporated into the transaction shown above. We define transactions of type $\text{Sum}$ having
one parameter as follows:

\[ \text{Sum}(j) \equiv \]
\[ \begin{array}{c}
\[ [\]
\[ || k : 1 \leq k \leq N \land k \mod (j \ast 2) = 0 :: \]
\[ v_1,v_2 : z(k-j,v_1),z(k,v_2) \rightarrow z(k,v_1+v_2) ] \]
\[ || j \ast 2 < N \rightarrow \text{Sum}(j \ast 2) \]
\end{array} \]

Note that we use parameter $j$ as a constant throughout the body of the transaction. A trans-
action instance $\text{Sum}(j)$—representing the $j$th iteration of the loop—updates the set of $z$
tuples to reflect the newly computed partial sum and inserts an appropriate transaction to continue the
computation.

For a correct computation of array $A$’s sum, the Swarm program must initialize the tuple
space to contain the $N$ elements of the array represented as $z$ tuples, i.e., to be the set

\[ \{ z(1,A(1)), z(2,A(2)), \ldots, z(N,A(N)) \}. \]

Similarly, the transaction space must consist of the single transaction $\text{Sum}(1)$.

Figure 2 shows a complete array-summation program. Since each $z$ tuple is only referenced
once during a computation, we modified the definition of the $\text{Sum}$ subtransactions to delete both
$z$ tuples that are referenced. Note that we have indicated the tuple deletion by marking tuple
forms in the query by $\dagger$ operators. If a tuple in a query is marked by a dagger, then, if the overall
query succeeds, the marked tuple is deleted as a part of the action.

At the beginning of this section we described a Swarm computation with the following metaphor:
"a large, rapidly moving aggregation of small, independent agents cooperating to perform a task."

5
So far we have taken a microscopic view of Swarm computations—focusing on the small, rapidly moving, independent agents (i.e., transactions and tuples). We should not, however, ignore the macroscopic view—losing sight of the large aggregation of agents cooperating to perform a task. Although the ArraySum program does not define a process in the sense of the CSP [10] model, the evolving “swarm” of transactions does embody a distinct, purposeful activity: computing the sum of an array. Using the assertional programming logics given in Section 4, we can specify and verify such “macroscopic” properties of Swarm computations. As we see in the next section, the synchrony relation enables us to organize simple transactions into complex groups which work on portions of the overall task.

3 Synchronic Groups

The discussion in the previous section ignored the third component of a Swarm program’s dataspace—the synchrony relation. The interaction of the synchrony relation with the execution mechanism provides a dynamic form of the \( \parallel \) operator. The synchrony relation is a symmetric, irreflexive relation on the set of valid transaction instances. The reflexive transitive closure of the synchrony relation is thus an equivalence relation. (The synchrony relation can be pictured as an undirected graph in which the transaction instances are represented as vertices and the synchrony relationships between transaction instances as edges between the corresponding vertices. The equivalence classes of the closure relation are the connected components of this graph.) When one of the transactions in an equivalence class is chosen for execution, then all members of the class which exist in the transaction space at that point in the computation are also chosen. This group of related transactions is called a synchronic group. The subtransactions making up the transactions of a synchronic group are executed as if they were part of the same transaction.

The synchrony relation can be examined and modified in much the same way as the tuple and transaction spaces can. The predicate

\[
\text{Sum}(i) \sim \text{Sum}(j)
\]

in the query of a subtransaction examines the synchrony relation for a transaction instance \( \text{Sum}(i) \) that is directly related to an instance \( \text{Sum}(j) \). Neither transaction instance is required to exist in the transaction space. The operator \( \approx \) can be used in a predicate to examine whether transaction instances are related by the closure of the synchrony relation.

Synchrony relationships between transaction instances can be inserted into and deleted from the relation. (The dynamic creation of a synchrony relationship between two transactions can be pictured as the insertion of an edge in the undirected graph noted above, and the deletion of a relationship as the removal of an edge.) The operation

\[
\text{Sum}(i) \sim \text{Sum}(j)
\]

in the action of a subtransaction creates a dynamic coupling between transaction instances \( \text{Sum}(i) \) and \( \text{Sum}(j) \) (where \( i \) and \( j \) have bound values). If two instances are related by the synchrony relation, then

\[
(\text{Sum}(i) \sim \text{Sum}(j))^\dagger
\]

deletes the relationship. Note that both the synchrony relation \( \sim \) and its closure \( \approx \) can be tested in a query, but that only the base synchrony relation \( \sim \) can be directly modified by an action. Initial synchrony relationships can be specified by putting appropriate insertion operations into the initialization section of the Swarm program.

Figure 3 shows a version of the array-summation program which uses synchronic groups. The subtransactions of \( \text{Sum}(j) \) have been separated into distinct transactions \( \text{Sum}(k, j) \) coupled by the synchrony relation. For each phase \( j \), all transactions associated with that phase are structured into a single synchronic group. The computation’s effect is the same as that of the earlier program.
program ArraySumSynch(N, A : [3 p : p ≥ 0 :: N = 2p], A(i : 1 ≤ i ≤ N))
  tuple types
  [i, s : 1 ≤ i ≤ N :: x(i, s)]
  transaction types
  [k, j : 1 ≤ k ≤ N, 1 ≤ j < N ::
   Sum(k, j) ≡
   v1, v2 : x(k − j, v1)↑, x(k, v2)↑ → x(k, v1 + v2)
   k ≠ N → (Sum(k, j) ∼ Sum(N, j))↑
   j < N, k mod 2 = 0 → Sum(k, j * 2), Sum(k, j * 2) ∼ Sum(N, j * 2)
  ]
  initialization
  [i : 1 ≤ i ≤ N :: x(i, A(i)));
  [k : 1 ≤ k ≤ N, k mod 2 = 0 :: Sum(k, 1); Sum(k, 1) ∼ Sum(N, 1)]
end

Figure 3: A Parallel Array Summation Program Using Synchronic Groups

4 Programming Logics

The Swarm computational model is similar to that of UNITY [3]; hence, a UNITY-style
assertional logic seems appropriate. However, we cannot use the UNITY logic directly because of
the differences between the UNITY and Swarm frameworks.

We follow the notational conventions for UNITY as used in [3]. Properties and inference
rules are written without explicit quantification; these are universally quantified over all the
values of the free variables occurring in them. We also introduce the notation "[t]" to denote
the predicate "transaction instance t is in the transaction space," TRS to denote the set of all
possible transactions (not a specific transaction space), and Initial to denote the initial state of
the program.

Below we summarize the proof rules for the subset of Swarm without the synchrony relation.
Reference [4] discusses these in more detail and [14] gives a formal operational model. The Swarm
programming logics have been defined so that the theorems proved for UNITY in [3] can also be
proved for Swarm.

The "Hoare triple"

\{p\} t \{q\}

means that, whenever the dataspace satisfies the precondition predicate p and transaction instance
t is in the transaction space, all dataspaces which can result from execution of transaction t satisfy
the postcondition predicate q.

We define Swarm’s unless relation with an inference rule similar to that given for UNITY’s
unless in [12]:

\[ \forall t : t \in \text{TRS} :: \{p \land \neg q\} t \{p \lor q\} \]

(p unless q)

The premise of this rule means that, if p is true at some point in the computation and q is not,
then, after the next step, p remains true or q becomes true.
Stable predicates and invariants are important for reasoning about Swarm programs. For a predicate \( p \) to be stable means that, if \( p \) becomes \textit{true} at some point in a computation, it remains \textit{true} thereafter. On the other hand, a predicate \( p \) is \textit{invariant} if \( p \) is \textit{true} at all points in the computation.

\[
\text{stable } p \equiv p \text{ unless false} \tag{3}
\]

\[
\text{invariant } p \equiv (\text{Initial } \Rightarrow p) \land (\text{stable } p). \tag{4}
\]

Following UNITY’s definition in [12], we define Swarm’s ensures relation with an inference rule:

\[
\text{p ensures q, } (\exists t : t \in \text{TRS } :: (p \land \neg q \Rightarrow [t]) \land \{p \land \neg q\} t \{q\} \Rightarrow p \text{ unless } q)
\]

\[
\text{p ensures q} \tag{5}
\]

The premise of this rule means that, if \( p \) is \textit{true} at some point in the computation, then (1) \( p \) will remain \textit{true} as long as \( q \) is \textit{false}, and (2) if \( q \) is \textit{false}, there is at least one transaction in the transaction space which can, when executed, establish \( q \) as \textit{true}. The \( p \land \neg q \Rightarrow [t] \) requirement generalizes the UNITY definition of ensures to accommodate Swarm’s dynamic transaction space.

The leads-to relation, written

\[
p \Rightarrow q \tag{6}
\]

means that, once \( p \) becomes \textit{true}, \( q \) will eventually become \textit{true}. (However, \( p \) is not guaranteed to remain \textit{true} until \( q \) becomes \textit{true}.) As in UNITY, the assertion \( p \Rightarrow q \) is \textit{true} if and only if it can be derived by a finite number of applications of the following inference rules:

- \( p \text{ ensures } q \\
  p \Rightarrow q \tag{basis}
- \( p \Rightarrow q, q \Rightarrow r \\
  p \Rightarrow r \tag{transitivity}
- \text{For any set } W, \\
  \{\forall m : m \in W :: p(m) \Rightarrow q\} \\
  \{\exists m : m \in W :: p(m)\} \Rightarrow q \tag{disjunction}

Unlike UNITY programs, Swarm programs \textit{terminate} when the transaction space is empty, that is

\[\text{Termination } \equiv (\forall t : t \in \text{TRS } :: \neg[t])\tag{7}\]

The logic we defined for Swarm programs may be generalized to accommodate synchronic groups. This involves the addition of a synchronic group rule and redefinition of the \textit{unless} and \textit{ensures} relations. The other elements of the logic are the same.

We define the “Hoare triple” for synchronic groups

\[
\{p\} S \{q\} \tag{8}
\]

such that, whenever the precondition \( p \) is \textit{true} and \( S \) is a synchronic group of the dataspace, all dataspaces which can result from execution of group \( S \) satisfy postcondition \( q \).

A key difference between this logic and the previous logic is the set over which the properties must be proved. For example, the previous logic required that, in proof of an \textit{unless} property, an assertion be proved for all possible transactions, i.e., over the set TRS. On the other hand, this generalized logic requires the proof of an assertion for all possible synchronic groups of the program, denoted by SG.

For the synchronic group logic, we define the logical relation \textit{unless} with the following rule:
\[
(\forall S : S \in \text{SG} :: \{ p \land \neg q \} \ S \ {p \lor q}\}) \\
\text{p unless q}
\]

If synchronous groups are restricted to single transactions, this definition is the same as the definition given for the earlier subset Swarm logic.

We define the ensures relation as follows:

\[
\begin{align*}
\exists t : t \in \text{TRS} :: (p \land \neg q \Rightarrow [t]) \land (\forall S : S \in \text{SG} \land t \in S :: \{ p \land \neg q \} \ S \ {q}\}) \\
\text{p ensures q}
\end{align*}
\]

This definition requires that, when \( p \land \neg q \) is true, there exists a transaction \( t \) in the transaction space such that all synchronous groups which can contain \( t \) will establish \( q \) when executed from a state in which \( p \land \neg q \) holds. Because of the fairness criterion, transaction \( t \) will eventually be chosen for execution, and hence one of the synchronous groups containing \( t \) will be executed. In the logic for the Swarm subset, the ensures rule requires that a single transaction be found which will establish the desired postcondition when executed. In the synchronous group logic, on the other hand, instead of requiring that a single synchronous group be found which will establish the desired state, the ensures rule requires that a set of synchronous groups be identified such that any of the groups will establish the desired state and that one of the groups will eventually be executed. If synchronous groups are restricted to single transactions, this definition is the same as the definition for the subset Swarm logic.

5 An Example Proof

In Sections 2 and 3 we derive a Swarm program for summing an array from a similar program expressed with the Guarded Commands (GC) notation. Figure 3 gives the Swarm program \textit{ArraySumSynch} which uses synchronous groups to compute the sum of an array. In this section we sketch a proof for this array sum program using the logic presented in Section 4.

The precondition for \textit{ArraySumSynch}, call it \texttt{Initial}, is similar to \texttt{Q}, the precondition of the GC program's loop given in Section 2. Of course, modifications are needed to account for the differences in data and program representation. (As in the previous section, we assume that all assertions are universally quantified over all the values of the free variables occurring in them.) Using the predicates \texttt{pow2} and \texttt{node} defined in Section 2, \texttt{Initial} can be stated as follows:

\[
\texttt{pow2}(N) \land (z(i,v) \equiv 1 \leq i \leq N \land v = A(i)) \land (\text{Sum}(i,j) \equiv \text{node}(i,2*j) \land j = 1) \land \\
(\text{Sum}(i,j) \sim \text{Sum}(k,l) \land i \leq k \equiv \text{node}(i,2*j) \land k = N \land j = l = 1)
\]

The second, third, and fourth conjuncts specify the structure of the tuple space, transaction space, and synchrony relation, respectively. Here the tuple and transaction forms, e.g., \( z(i,v) \) and \( \text{Sum}(i,j) \), represent predicates which are true when there is a matching entity in the dataspace and false otherwise. Likewise, predicates using the \( \sim \) and \( \approx \) connectives represent predicates over the synchrony relation.

The postcondition of \textit{ArraySumSynch}, call it \texttt{Post}, is similar to \texttt{R}, the postcondition of the GC program. Using the \texttt{sum\_A} expression defined in Section 2, it can be stated as

\[
z(i,v) \equiv (i = N \land v = \text{sum\_A}(0,N)).
\]

To verify that the program satisfies this specification we must prove that, when the program begins execution in a state satisfying \texttt{Initial}, it eventually reaches a state satisfying \texttt{Post} and, once such a state is reached, any further execution will not falsify \texttt{Post}. That is, we must prove a progress property and a safety property—the Sum Completion and Sum Stability properties, respectively.

Property 1 (Sum Completion) \texttt{Initial} \( \iff \) \texttt{Post}
Property 2 (Sum Stability) stable Post

To prove these properties, a property is needed which characterizes the invariant relationships among the elements of the datastore. We call this property the Structure Invariant. It serves a role in the Swarm proof similar to the role the loop invariant Q does in a proof of the GC program. In the statement of the Structure Invariant, we use the function \( W_x \), which is \( N \) divided by the number of \( x \)-tuples present in the datastore, i.e.,

\[
W_x = \frac{N}{(\# i, v :: x(i,v))}.
\]

\( W_x \) represents the "width" of the segment of array \( A \) whose sum is in each \( x \)-tuple—a role served by the variable \( j \) in the GC program.

Property 3 (Structure Invariant)

\[
\text{invariant} \quad \text{pow2}(N) \land \text{pow2}(W_x) \land (x(i,v) \equiv \text{node}(i,W_x) \land v = \text{sum}_A(i-W_x,i)) \land \\
(Sum(i,j) \equiv \text{node}(i,2*j) \land j = W_x) \land \\
(Sum(i,j) \sim Sum(k,l) \land i < k \equiv \text{node}(i,2*j) \land k = N \land j = l = W_x)
\]

The Structure Invariant is relatively complex. This complexity arises from the mutual dependencies among the data tuples and transactions of this highly synchronous program.

Proof of the Structure Invariant. Call this property \( I \). To prove the invariance of \( I \), we have to show that \( I \) holds initially and that it is stable. Since initially \( W_x = N/N = 1 \), \( Initial \Rightarrow I \). Hence, \( I \) holds initially.

To prove \( I \) is stable we must show that \( I \) is preserved by all possible synchronic groups \( G \), i.e., \{ \{ I \} \} \( G \) \{ \{ \} \} is true for arbitrary \( G \). For any synchronic group which does not satisfy \( I \), this predicate is trivially true. Thus, we only need to consider those synchronic groups which satisfy \( I \). Since the value of \( N \) is not altered by any transaction, the \( \text{pow2}(N) \) conjunct is preserved trivially. We will show that each of the remaining four conjuncts is preserved.

To see that the second and third conjuncts are preserved, note that each executing transaction deletes two \( x \)-tuples and inserts back a single \( x \)-tuple. The "indexes" (first components) of the deleted tuples are adjacent multiples of \( j \) (i.e., of \( W_x \)). The inserted tuple is positioned at the index of the rightmost deleted tuple—at a multiple of \( 2 \times j \). The value (the second component) of the inserted tuple is equal to the sum of the values of the two deleted tuples. Furthermore, note that the preconditon \( I \) guarantees that each of the transactions in the group operate upon different tuples.

Consider the fourth conjunct. Note that all transactions (allowed by \( I \)) have the same "phase" parameter \( j \). Further note that the values of the "index" parameter \( i \) for these transactions are multiples of \( 2 \times j \). Only half of the transactions, i.e., those whose index is a multiple of \( 4 \times j \), insert successor transactions. The phase for all the inserted transactions is \( 2 \times j \) (i.e., \( 2 \times W_x \)). As argued above, \( W_x \) is also doubled in value by the synchronic group's execution.

Also note that the only synchrony relationship for a transaction \( Sum(i,j) \), for \( i < N \), is with transaction \( Sum(N,j) \). Upon execution, a transaction deletes this relationship. For each new transaction inserted (into phase \( 2 \times j \)), a synchrony relationship is created with \( Sum(N,2*j) \). Thus the fifth conjunct is preserved.

Proof of Sum Stability. We must show that the predicate Post is preserved by all synchronic groups allowed by the Structure Invariant \( I \). Note that \( Post \land I \Rightarrow W_x = N \). But \( W_x = N \land I \Rightarrow (\forall i :: \neg \text{node}(i,2*W_x)) \). Thus, when \( Post \) is true, because of the fourth conjunct of \( I \), the transaction space must be empty. Therefore, \( Post \) is clearly stable.

Now we turn our attention to the Sum Completion progress property. This large-grained progress property can be proved by induction using a finer-grained progress property corresponding to the execution of a single synchronic group. The Sum Step property is stated as an ensures relation.
Property 4 (Sum Step)

\[ W_x = k < N \text{ ensures } W_x = 2 \times k. \]

\[ \exists \text{ Proof of the Sum Step property. The proof of an ensures property has two parts: an existential part and an unless part. } \]

The existential part requires us to prove that, whenever \( W_x = k < N \), there is a transaction in the transaction space such that any synchronic group containing that transaction will establish \( W_x = 2 \times k \). But, in accordance with the Structure Invariant \( I \), at most one synchronic group exists at a time. (Particularly, \( W_x = k < N \land I \Rightarrow \text{Sum}(N, W_x) \).) As argued in the proof of the Structure Invariant, this synchronic group will double \( W_x \), i.e., decrease the number of \( z \)-tuples by half.

The unless part requires us to prove \( W_x = k < N \) unless \( W_x = 2 \times k \). That is, we must show for all synchronic groups \( G \),

\[ \{ \ W_x = k < N \land I \ \} \ G \ \{ \ W_x = k \lor W_x = 2 \times k \ \} \]

is valid. As argued above, the only synchronic groups allowed by \( I \) will double \( W_x \). Therefore, the ensures property holds.

\[ \exists \text{ Proof of Sum Completion. To prove Sum Completion, we must show that } \neg \text{Post} \rightarrow \text{Post. Note that } \neg \text{Post} \land I \Rightarrow W_x < N \text{ and } W_x = N \land I \Rightarrow \text{Post. We choose the well-founded metric } \frac{N}{W_x} = 1. \text{ (}\frac{N}{W_x}\text{ is the count of the } z\text{-tuples present in the dataspace. The metric is similar to the variant function } v \text{ in the GC program proof.) Using this metric, the Leads-to Induction Principle [3] applied to the Sum Step property allows us to deduce that } W_x < N \rightarrow W_x = N. \text{ Therefore, we deduce that } \neg \text{Post} \rightarrow \text{Post} \text{ by the Leads-to Implication Theorem [3] and the Leads-to transitivity rule.} \]

We have thus proved the Sum Stability and Sum Completion properties of the Swarm program ArraySumSynch. Therefore, the program satisfies its specification. Also note that \( \text{Post} \land I \Rightarrow \text{Termination} \) (as we argued in the proof of Sum Stability). ArraySumSynch terminates immediately upon completing the computation of the desired sum.

6 Discussion

Content-based access to data, dynamic statements (i.e., transactions), and the synchrony relation are three key features which distinguish Swarm [4, 14] from UNITY [3]. In previous papers we justified the first two extensions on three grounds. First, even though the Swarm programming logic is more complex than the UNITY logic, the additional complexity is not evident in proofs unless the new features are actually used. The Swarm inference rules reduce to those employed in UNITY if we restrict the usage of Swarm to a UNITY-like subset. In such a subset, we can represent UNITY’s variables as tuples, UNITY’s assignments as transactions which delete and reinsert these tuples, and UNITY’s static set of statements as transactions which reproduce themselves without introducing additional transactions into the dataspace. Second, for problems whose precise structure and space requirements (in terms of number of variables and statements) cannot be determined a priori, Swarm allows one to tailor the dataspace appropriately. Scaling all the values of a sparse matrix in parallel, for instance, does not require the presence of tuples and statements for the zero entries. Third, to the best of our knowledge, Swarm is the first notational system to make assertional-style proofs of rule-based programs feasible [8]. Moreover, since tuples can easily simulate both variables and messages, Swarm makes it possible to write and prove programs that employ these three programming paradigms.

Why introduce the synchrony relation? Consider UNITY’s synchronous composition operator. The \( || \)-operator binds a group of assignments into a single statement; the assignments in the group are executed synchronously as a single atomic action. Syntactically, the \( || \) is placed between the assignments making up the UNITY statement. Swarm’s \( || \)-operator is similar both syntactically and semantically. However, the Swarm model led us to generalize the notion of synchronous
execution. While UNITY statements are static, anonymous entities, Swarm transactions are dynamically created entities which have unique identifiers (i.e., the type name and parameter values). This led us to a more dynamic notion of synchronous execution: the atomic execution of a group of related transactions where the relation between transactions is an entity subject to examination and modification by the program. As a result, Swarm separates the definition of statements (i.e., state transformations) from the specification of their processing mode (i.e., synchronous or asynchronous). The former appears in the transaction type definitions while the latter is captured by the synchrony relation present in the dataspace. Finally, the syntactic restriction regarding the use of the \[\|\text{-}\text{operator (i.e., no simultaneous assignments to the same variable)}\] had to be removed to allow for synchronous execution of any group of transactions. This was accomplished by giving precedence to dataspace insertions over deletion.

Among all these changes, the ability to alter the definition of the synchrony relation is clearly the most radical departure from UNITY. Thus, the asynchronous, partially synchronous, and fully synchronous modes of execution offered by UNITY are augmented in Swarm by dynamic partial synchrony. This processing mode is unique to Swarm and cannot be simulated easily in UNITY. Recently we have begun exploring the modeling and programming implications of dynamic partial synchrony \[15\]. There are strong indications that dynamic partial synchrony will be helpful in modeling reconfigurable or heterogeneous computer architectures and will lead to new kinds of solutions to a variety of programming tasks. A related concept is also proving useful in parallel program refinement. Liu and Singh \[11\] have subsequently applied a complementary concept, the synchrony relation, to the problem of refining a UNITY program toward architectures with different synchrony structures.

The capacity to model a variety of computer architectures makes Swarm attractive as a specification language for software which, because of performance considerations, is targeted to a specific and often heterogeneous hardware organization. In such cases, the software designer first derives a software specification expressed in the Swarm notation—verifying that it is correct with respect to the application’s requirements and is compatible with the chosen architecture. Subsequently, programmers use the specification as the basis for implementing the software modules allocated to the individual architectural components. To illustrate the kind of architectures we have in mind let us consider an application in which a three-dimensional geometric model is created, manipulated, and displayed. Furthermore, let us assume that the underlying architecture consists of a producer, a transformer, and a mapper. The producer, which could be a typical workstation, runs asynchronously, generating various objects in the 3D model. The transformer is a processor pipeline that can be dynamically reconfigured and activated by the producer. When active, the transformer takes objects from the producer’s memory, applies a series of transformations to them, and places the result in a very large buffer accessible to the mapper. The mapper, in turn, is an SIMD machine whose task is to copy sections of the buffer to the refresh memories of a number of display units. In Swarm, the producer activities would be specified as a set of transactions which are asynchronous with respect to each other and to the other components. The pipeline reconfiguration would involve the formation of a synchronic group consisting of transactions which manipulate, object by object, the geometric model. Finally, the SIMD machine might be captured by yet another synchronic group, one that is not subject to change. Although our solution sketch might be overly simplistic, this example illustrates that the synchrony relation is a convenient construct for modeling certain kinds of architectures.

Turning now to programming considerations, the first thing that must be noted is the use of the synchronic group to control the granularity of the computation. Because each synchronic group execution represents an atomic transformation of the dataspace, by adjusting the size of each group the programmer can switch between fine-grained and course-grained operations or can combine the two in a single program. Let us consider, for instance, the earlier 3D model and let us assume that it depicts a platform which holds several machines with moving parts. At each moment the position of a point in the model is affected by the combined movement of the platform, the machine to which the point belongs, and the moving part involved. Any parallel implementation
of the 3D model dynamics must maintain proper consistency in the model, i.e., the effects of each movement must be perceived as a series of small atomic changes in the positions of all points on some part, on some machine, or on the entire contents of the platform. An obvious Swarm solution associates with each point three transactions, each a member of a different synchonic group. The platform group includes one transaction for each point in the model and ensures the atomicity of the coordinate transformation resulting from the platform movement. Similar synchonic groups are formed for each machine and part.

The solution is elegant because it separates the implementation of the three independent movements and requires no synchronization code. Could we have the same solution in UNITY? If the contents of the platform does not change and if moving parts do not fall off, the answer is yes. Swarm, however, can accommodate without loss of elegance the materialization and dematerialization of machines or parts, and, in general, any arbitrary structural changes in the 3D model. This is accomplished by means of appropriate restructuring of the synchonic groups. The ability to mold the computation to the changing structure of the problem at hand, through the creation and restructuring of arbitrary atomic transformations, is yet another reason for introducing synchonic groups in Swarm.

In further pursuit of programming convenience, we have extended the semantics of synchonic groups to allow for a limited exchange of information among the constituent transactions. We added the special predicates OR, AND, NOR, and NAND, meaning any, all, none, and not-all, respectively [14]. Upon execution, each regular query (i.e., a query not containing special predicates) makes it’s success/failure status available to other queries in the synchonic group. When a query which contains a special predicate is executed, this set of boolean values is accessed to determine its outcome. The result is the convenient capability to detect certain kinds of “consensus” [6] among the subtransactions of a synchonic group. Since many common programming problems involve agreement among a set of participants, this capability can be very useful. For example, quiescence within a synchonic group involving only regular queries can be detected by introducing a NOR query into the group. The NOR succeeds only when all the regular queries have failed and, therefore, no further activity can originate within the group. Of course, an OR query is needed to recreate the synchonic group when quiescence is not yet reached. The definition below shows the basic code structure for transactions participating in some quiescence detection activity:

\[
\text{Worker}(n) \equiv \\
\text{any_work_for_me?} \rightarrow \text{do_the_work} \\
\text{|| OR} \rightarrow \text{Worker}(n) \\
\text{|| NOR} \rightarrow \text{follow-up_activities}
\]

In contrast to the Swarm solution, classical quiescence detection algorithms are quite complex.

The updating of local clocks in a distributed simulation is another example of the use of special queries. All local clocks are placed in the same synchonic group and execute the following logic:

\[
\text{Clock}(n, \text{local_time}) \equiv \\
\text{is_current_step_completed?} \rightarrow \text{skip} \\
\text{|| AND} \rightarrow \text{Clock}(n, \text{local_time} + 1) \\
\text{|| NAND} \rightarrow \text{Clock}(n, \text{local_time})
\]

Components that complete their simulation early can terminate by simply removing their clock from the synchonic group. The addition and removal of one clock is hidden from all others; the only interactions among the clocks are through the special predicate “consensus” mechanism.

Although our limited experience indicates that the synchonic group is a useful concept, a number of questions remain open. Among them, three are of immediate concern. First, does the programming convenience arising from the synchonic group feature compensate for the additional complexity of proofs of programs that employ them? Although proofs involving synchonic groups have shown, in general, to be more difficult than we had expected, drastic simplifications in
program logic brought about by the use of synchronic groups may ultimately ease the verification task. Second, will the study of dynamic partial synchrony lead to interesting new distributed algorithms for some classical problems? We believe so. Finally, how can Swarm’s apparent ability to model complex architectures be put to practical benefit?

7 CONCLUSIONS

The Swarm programming logic is the first axiomatic proof system for a shared dataspace "language." To our knowledge, no axiomatic-style proof systems have been published for Linda, production rule languages, or any other shared dataspace language. Taking advantage of the similarities between the Swarm and UNITY computational models, we have developed a programming logic for Swarm which is similar in style to that of UNITY. The Swarm logic uses the same logical relations as UNITY, but the definitions of the relations have been generalized to handle the dynamic nature of Swarm, i.e., dynamically created transactions and the synchrony relation. In this paper we have shown how one can extend the proof logic for Swarm to accommodate the dynamic formation of synchronic groups specified by the runtime redefinition of the synchrony relation.

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