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A Simulation of Demand-Driven Dataflow: **Translation from Lucid into MDC Language**

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Abstract

Message Driven Computation (MIX) **is** *a model of computation with which we have been experimenting at the Illinois Institute of Technology. It* **is** *our hire to prove the viability of MDC in practice for the expression of parallel algorithms and the implementation of ficnctional and datajlbw programming languages. In the following pages we* **&cuss** *our implementation of the Lucid programming language in ME. The discussion will present a subset of Lucid which illustrates the principles ofLuci4 Message Driven Computing, and the translation into and the interpretation of* dataflow graphs.

1.0 Message Driven Computing

Message Driven Computing **(MDC)** is a model of parallel and distributed computation developed at the **Illinois** Institute of Technology by Thomas Christopher **[Christopher 19891.** Central to **MDC** is the notion of a computational event. Computational events **are** executions of functions which map input messages into output messages. *All* message passing between locations in MDC is achieved unidirectionally and asynchronously. A computational event **occurs** at a location when a pattern of messages accumulates at the location. Locations **are** named by computable tuples of information. When two *or* more computational events OCCUT at a location, mutual exclusion between the computational events is guaranteed. **MDC** has been implemented on a variety of machines: the **Encore** Multimax, the BBN Butterfly, and the NCUBE.

2.0 Lucid

Lucid is a family of functional dataflow languages defined and designed by Wadge and Ashcroft **Wadge and Ashcroft 19851.** Inherent to the definition of any particular Lucid language **are** sequences, Lucid operators, pointwise infix and prefix numeric operators, user functions (filters), and list operators (optional).

2.1 Terminology of Lucid

A sequence in Lucid is defied to **be** an infinite series of values ordered (or tagged) by time. The sequence is the basic tenet of Lucid programming. Some examples of sequences **are** constants, definitions, and the results of function calls. A constant sequence is a sequence whose value at every time is the same. A **definition** of a sequence provides a programmer the facility to have variables which change over time but not to have variables whose history of updates is destroyed. A definition implies that Lucid is a single-assignment language (a tenet of pure dataflow and functional languages). A **function** maps one **or** more input sequences into an output sequence. In the literature, functions are often alluded to as filters. Sequences are operated on by Lucid operators, pointwise operators, and functions. A Lucid operator is a function which maps one or two input streams onto an output *stream* whose values are usually values in the history of the input streams.

2.2 Selected Lucid Operators Defmed

first

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The first operator is applied to a sequence **x** to produce a constant sequence whose value throughout

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is the first value of sequence x. Formally, the value of first **x** at time t is the value of x at time 0.

next

The next operator is applied to a sequence **x** to produce a sequence in which the value of next x at time t is the value of **x** at time t+l.

fbY

The fby operator is applied to sequences x and y to produce a sequence which is literally the first value of x followed by the sequence y.

23 Pointwise Operators

Pointwise operators **are** operators which **are** applied to every element of a sequence in much the same manner we **are** accustomed to in imperative programming languages. Arithmetic operators are all pointwise operators. The operation $x + y$, applied to sequences **x** and y, produces a sequence in which each element at a time t is the sum of the value of x at time t and the value of y at time t.

2.4 Language Characteristics

A Lucid program is an expression. Expressions include constants, variables, prefix expressions, infix expressions, list expressions, conditional expressions (if, *case,* and cond), function calls, and the where clause. A where clause is the mechanism provided for definitions of variables which are bound to expressions. If a variable in an expression cannot be resolved to one of the definitions in its where clause, the variable is a global variable whose history is defined at run time by the user.

3.0 Translation and Interpretation

The system we have developed is a translator for Lucid which produces MDC **as** target *code.* The translator accepts **as** input a program expressed in Lucid and produces an **MDC** initial behavior which constructs a run time dataflow graph whose vertices **are** MDC **locations** and edges **are messages. A** collation program links the MDC encoded initial behavior with a source file which **contains** a number of **MDC** behaviors to interpret the dataflow graph to produce an object **MDC** program. The MDC program is passed through the **MDC** translator to produce C code which is finally Compiled with

UNIX *cc* and linked with the MDC run time system to produce an executable, parallel program which simulates dataflow *on* a parallel machine.

3.1 Construction of Datailow Graphs

We now tum to a concrete example of how our Lucid system constructs dataflow graphs from a simple Lucid source program **and** the run time behaviors **requited** to completely interpret the dataflow graph. *As* an aside, we have implemented much more **than** just a handful of trivial *cases,* but we believe a simple example best illustrates the principles of translation into and the interpretation of dataflow graphs.

Below is a simple Lucid program which computes the value of x+y+z, given a sequence of natural numbers x, a constant sequence y, and a user-supplied sequence **z.**

```
x + y + zwhere 
     x = 1 f by x + 1;
    y = 1;
end
```
An initial behavior is produced by the Lucid translator which performs a number of send message operations. Below is **MDC** pseudocode for the initial behavior which is produced by the Lucid translator. Section A indicates the group of send operations which constructs the dataflow graph for the Lucid program in Figure 1, while section B indicates the group of **send** operations which establishes a demand pattern at the start vertex (of location) of the dataflow graph for the values of the expression x+y+z. *All* locations **are** generated by the translator **as** needed.

initial behavior

{ Section A } send a var (x_1) message to location L1. send a var(y-1) message to location **L2.** send a plus(L1,L2) message to location L3. send a **var(z-0)** message to location LA. send a plus(L3,L4) message to location L5. send a const(1) message to location **L6.** send a var (x_1) message to location L7. send a const(1) message to location **L8.** send a plus(L7,L8) message to location L9. send a fby(L6,L9) message to location x_1 . send a const (1) message to location y_1 .

{section B}

 $\ddot{}$

send a demand(globa1 environment, time, destination) message for the first n values of the Lucid program rooted at location **I5.** *end initial behaviar*

3.2 Interpretatkm of Dataflow Graphs

A run time system is quired to interpret the **MDC** dataflow graphs which were produced from the Lucid source program. The **run** time system is comprised of **MDC** behaviors. *As* mentioned earlier in the discussion of **MDC,** a behavior (computational event) is **specified** by a **pattern** of messages and a body of code. The MDC behaviors must be carefully designed to ensure that appropriate message patterns exist for every conceivable dataflow graph produced by the Lucid translator.

We now present a subset of the **run** time system behaviors. Though it would be interesting to the reader to examine the entire **run** time system, dozens of **pages** would be required (merely to present pseudocode). The subset of **run** time system behaviors which we unveil in the following **sections** is sufficient to completely interpret the dataflow graph which was constructed by the initial behavior above.

3.2.1 Interpretation of Infix Operators

behaviorlmessage pattern plus(e1, e2) and demand(env, t, dest) $actions$ Create a **new** location, **s,** to **perform** addition. Send a do plus(dest) message to location **s**. Let L be the left operand portal at location **s.** Send a demand(env,t,L) message to e1. Let R be the right operand portal at location **s.** Send a demand(env,t,R) message to e2. *end behavior*

behaviorlmessage pattern $do_{plus(dest)}$, lopnd(v1), and ropnd(v2) $actions$ Send value($v1 + v2$) message to dest. *end behavior*

3.2.2 Interpretation of Constants

behaviorlmessage pattern const(v) and demand(env, time, dest) $actions$

send value(v) message to dest. *end behavior*

3.23 Interpretation of Variables

behaviorlmessage pattern var(name) and demand(env, time, dest) *actwns* Send demand var(dest) message to location named <em, name, time>. *end behavior*

one time behaviorlmessage pattern demand var(dest) $actions$

Send a demand(env, time, here) message to a location whose name is extracted from the present location name. The name of the location to where the demand is sent **contains** message information pertinent to the definition of **the** variable name.

Leave the demand-var(dest) message at this location *(so* this behavior **cannot** be executed again). *end behavior*

behaviorlmessage pattern demand-var(dest) **and** value(v) *actwns*

Send a value(v) message to dest.

Leave the value(v) message at this location, **so** a subsequent demand will **not** result in recomputation a variable at the same time in the same environment. *end behavior*

3.2.4 Interpretation of a Lucid Operator

behaviorlmessage pattern send demand(env, time, dest) and fby(e1, e2) $actions$ If time is **zero,** send a demand(env, *0,* dest) message to location e1. Otherwise, **send** a demand(env, time-1, dest) message to location e2. *end behavior*

33 A Wave of Computation

As mentioned earlier, the above behaviors comprise only a fraction of the Lucid run time system implemented in MDC. In effed, these behaviors work together to implement the method of eduction proposed in **[Wadge and Ashcroft 1985].** The interpretation of the dataflow graphs by the MDC

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system can be viewed **as** a traveling wave of messages which commences at the initial behavior with demand messages. Demand messages which arrive at a location trigger behaviors which, also, send demand messages. Ultimately, demand messages lead **to** the computation of actual values which can be returned to the location from where the original demand was made.

4.0 Other Lucid Implementations

An eductive interpreter for the language pLucid was implemented under UNIX in the C programming language at Arizona State University **Faustini and Wadge 1987j.** This interpreter is perhaps the greatest of the success stories about Lucid implementation. It implements all of the features of the pLucid language described in *[wadge* **and Ashcroft 198Sl. The** eduction method implemented in the Arizona State pLucid system is exactly the one outlined in **[Wadge and Ashcmft 1985J.**

A translator is described by Pilgram [Pilgram 1983] [Wadge and Ashcroft 1985] which translates Lucid into message-passing actors. The method works for a large number of programs; however, it fails to work for a large number of programs. The **programs** which have failed to be interpreted are those which involve Lucid operators in function *calls.*

5.0 Sisnificance

We believe our research is significantly different in its focus from the **research** described above. The pattern matching capabilities of **MDC,** which **are** absent in actors languages, **are** much better suited **to** the interpretation of graphs. The notion of a location is particularly well-suited for the storage of variable histories.

Another significant **aspect** of **OUT** research is that our work is portable. The MDC system is written in the C programming language. Parallel versions of the system are available for many machines, while sequential vetsions of *the* system are available for practically every machine which has a C compiler.

6.0 Future Research with Lucid

We have a least two goals with respect to our Lucid research. Our foremost goal is to implement the entire *Arizona* pLucid language with some extensions to support arrays. We are exploring the

idea of an APL-like array package for MDC into which our Lucid extensions could be translated.

Our *dhet* goal is **to** finish an implementation of Lucid which **performs** well *on* all parallel computer architectures, especially distributed architectures. To do *so,* optimizations will have **to be** done by the compiler. Our current implementation does limited optimizations, because it is a prototype. We also believe that the addition of arrays to the language will result in improved performance on such architectures, because of studies we have done **on** the effeds of grain **size** *on* speed-up and efficiency [Christopher 1990].

7.0 References

- [1] E. Ashcroft, *Easyflow Architecture*, Technical Report, Computer Science Laboratory, SRI Intemational **1985.**
- **[2]** T. W. Christopher, *Early Experience with Object-Oriented Message Driven Computing,* Frontiers of Massively Parallel Computing **1990,** October **1990.**
- **[3]** T. W. Christopher, *Elqoloration of the Limits that Grain Size Imposes on Speed-up and Emieney* **of** *Two Transitive ClosureAlgorithms,* Fourth Annual Parallel Processing Symposium, Orange County IEEE, April 4-6, 1990.
- **[4]** A. Faustini and W. Wadge, *An Eductive Interpreter for the Language pLucid,* Proceedings of the SIGPLAN **'87** Symposium on Interpreters and Interpretive Techniques, 1987.
- *[5]* P. Pilgram, *Translating Lucid into Message Pusshg Actors,* **Ph.D. Dissertation,** University of Warwick, England.
- [6] **W. Wadge and E. Ashcroft,** *Lucid: the Dataflow Programming Language,* Academic Press, Orlando, **Florida, 1985.**

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