# Layering Techniques for Development of Parallel Systems 

An Algebraic Approach *<br>Mannes Poel \& Job Zwiers, University of Twente, Dept. of Computer Science P.O. Box 217, 7500 AE Enschede, The Netherlands E-mail:\{mpoel,zwiers\}@cs.utwente.n]


#### Abstract

A process language is presented which makes a clear distinction between temporal order and causal order. This allows for several algebraic laws that are particularly interesting for the design of concurrent systems. One of these is an algebraic formulation of the communication closed layers principle by [EF82]. These laws suffice to rewrite process terms that avoid specification of temporal ordering into a unique normal form. Other transformations allow for gradually imposing temporal ordering on an already functionally correct design. The combination of such laws enables a design strategy where architecture independent designs are transformed towards a form that matches a particular implementation architecture. We apply this style of design to various distributed algorithms, including an algorithm for the "point-in-polygon" problem transformed to a form suitable for pipelined execution on a tree network, and the Floyd-Warshall algorithm for the all-points shortest path transformed to a form suitable for execution on a SIMD architecture.


## 1 Introduction

In order to have a transformational algebraic approach that suits both specification and design of concurrent systems, a clear distinction should be made between temporal order and causal order between actions. In this paper we present a process language where a distinction is made between language constructs for specifying temporal order, such as the sequential composition operator, and causal order, using the layer composition operator, cf. [JPZ91]. Layer composition, denoted by $P \bullet Q$, gives rise to several important algebraic laws. For instance the communication closed layers principle from [EF82] can be formulated as the following algebraic law:

$$
(P \cdot Q) \|(R \cdot S)=(P \| R) \cdot(Q \| S)
$$

provided there is no "conflict" (communication) between $P$ and $S$, and between $Q$ and $R$. Another algebraic law, called the Left-Right Movers law states in its simplest form that

$$
(P \cdot Q) \cdot(R \cdot S)=(P \cdot R) \cdot(Q \cdot S)
$$

if there is no conflict between $Q$ and $R$. The above laws would not be valid if we replaced layer composition by sequential composition.
Layer composition is a valuable tool in the initial design stage of a system. Such operations allow for an architecture independent design strategy. In this initial phase of design, the above laws together

[^0]with an expansion theorem make it possible to rewrite a process term that avoids specification of temporal order into an unique normal form. This normal form is the maximal parallelization of that process term, and resembles the normal form for Mazurkiewicz traces. Other transformations and implementation relations, cf. section 2, then allow for a transformation of this normal form towards a process that can be implemented on a particular architecture.
This style of design is demonstrated on various distributed algorithms. First the "point-in-polygon" algorithm presented in the book of Akl, [Akl89] is considered. After removing all the irrelevant temporal order, but preserving the causal order induced by the temporal order (i.e. changing every sequential composition in a layer composition), this algorithm is rewritten into normal form. Afterwards this normal form is transformed to several distributed processes, all suitable for pipelining on a tree network.
Next the Floyd-Warshall algorithm for the "all-points shortest path" problem is discussed. Again after all the temporal order is deleted, but the induced causal order is preserved, this algorithm is transformed into normal form. Afterwards several optimal parallelizations are obtained, both for a CREW-PRAM and an EREW-PRAM.
Other applications to for instance protocols can be found in [JZ92b, JZ92a], and to parsing in [JPSZ91].
In section 2 we introduce the language together with informal discussion of the semantics. Also the algebraic laws and the theorem on normal forms are presented. In section 3 we apply the transformational algebraic approach to the point in polygon algorithm on a tree network. Section 4 gives several transformations of the Floyd-Warshall algorithm for the all-points shortest path problem. One transformation is suitable for a CREW-PRAM and another for an EREW-PRAM. Both are optimal parallelizations of the sequential Floyd-Warshall algorithm.

## 2 Language, Algebraic Laws, and Normal Forms

In this section we introduce a language which contains both layer composition and sequential composition. Assume that processes perform actions a that read and write shared variables $x, y, z, \ldots \in \mathcal{V} a r$, and perform (boolean valued) tests on shared variables. We employ the usual (simultaneous) assignment notation $\mathrm{x}:=\mathrm{f}$ where x and f are a list of variables and a list of expressions. Such assignments are guarded by means of a boolean expression $b$ which must evaluate to true before executing the assignment. For such actions, that we denote by $b \& x:=f$, the evaluation of the guard together with the assignment constitute a single atomic action. When the guard $b$ is identically true, we omit it and employ the usual simultaneous assignment notation $x:=f$. Similarly we regard boolean tests $b$ as degenerate cases where the assignment part has been left out. Such guards can be used to model more conventional constructs. For instance we use

$$
\text { if } b \text { then } P \text { else } Q \text { fi as an abbreviation for }(b \bullet P) \text { or }(\neg b \bullet Q)
$$

For an action $a$ of the form $b \& x:=f$ the set of variables $\{x\}$ is called the write-set $W(a)$ of $a$. Similarly, we define the read-set $R(a)$ as the set of variables occurring (free) in $b$ and the expression list $f$. Finally we define the base of $a$ as base $(a)=R(a) \cup W(a)$. Two actions $a_{0}$ and $a_{1}$ conflict if one of them writes a variable $x$ that is read-or written by the other one. Formally we define a conflict relation on actions, denoted by $a_{0}-a_{1}$ as follows:

$$
a_{0}-a_{1} \text { iff } W\left(a_{1}\right) \cap \text { base }\left(a_{0}\right) \neq 0 \text { or } W\left(a_{0}\right) \cap \text { base }\left(a_{1}\right) \neq 0
$$

Other models can be easily obtained by changing the conflict relation, for example by introducing also read-read conflicts, which is in correspondence with the EREW-PRAM.
The syntax for $D L$ is as follows:

```
\(P \in D L\),
\(P::=b \& x:=\mathrm{f}|P \| Q| P \cdot Q|P ; Q| P\) or \(Q\)
    \(\mid\) skip \(\mid\) empty \(|P \backslash x|\langle P\rangle \mid i o(P)\)
```

We will now provide the intuition for the language operations of $D L$. A process $P$ as a whole denotes the set $\llbracket P \rrbracket$ of all possible runs for that system. Execution of an action a results in a single
event. Therefore, actions are executed atomically. Our semantic domain is such that events, say $c_{0}$ and $e_{1}$, are in conflict, then they are ordered. Hence each run consists of a set of events, an a partial order $\rightarrow$ on events, such that events which are in conflict are ordered with respect to $\rightarrow$. Summarizing:
Each run $(E, \longrightarrow)$ of $P$ satisfies:

- $(E, \longrightarrow)$ is a pomset of events, i.e. a partially ordered multiset.
$-\longrightarrow$ is a partial order on $E$ such that if $e$ and $e^{\prime}$ are in conflict then $e$ and $e^{\prime}$ are ordered with respect to $\rightarrow$.

Consequently we regard two processes $P$ and $Q$ as equal, denoted by $P=Q$ iff their sets of possible runs are equal, i.e. iff $\llbracket P \rrbracket=\| Q \rrbracket$.
For parallel composition, the order that necessarily must exists between conflicting $P$ and $Q$ events is nondeterministically determined. The nondeterministic choices for different pairs of conflicting events are of course subject to the condition that the order must remain a partial order, so certain choices are excluded. Each choice corresponds to a (potential) different net effect of the whole run.

For layer composition $P \bullet Q$ the situation is the reverse: any $P$ event $e_{0}$ precedes any $Q$ event $e_{1}$ with which it conflicts. This resembles the sequential composition construct, but there is a substantial difference. For sequential composition $P ; Q$ all $P$ events precede any $Q$ event. This is not so for layer composition. In the latter case any $Q$ event $e$ must wait only for its causal predecessors, implying that it need not wait for all $P$ events.
Nondeterministic choice $P$ or $Q$ is a straightforward construct that either executes $P$ or $Q$.
The process skip performs no action at all, and the empty process, cannot perform any computation at all, not even the computation executed by skip which contains no events. Both processes aid in formulating some algebraic properties of $D L$, where they act as a unit element and as a zero element respectively.

Atomic execution. Atomic brackets $\langle P\rangle$ serve to indicate that $P$ should execute "atomically", i.e. without interference by other processes.
The hiding construct $P \backslash x$ hides the variable $x$ in each run of $P$,i.e. it is removed from the write-set and read-set of each event. (Moreover events with empty read-and write-set are removed from the run.) The complement of the hiding operator is the projection operator; let $S$ be a set of variables and let $A$ be all the variables of a process $P$ then

$$
P_{\mid S} \stackrel{\text { der }}{=} P \backslash(A-S)
$$

All the variables except those contained in $S$ are hidden in each run of $P$.
io $(P)$ denotes execution of a single action that captures the net effect of executing $P$ without admitting interference by other events. The io( $\cdot$ ) operation is also called the contraction operation, since it contracts complete $P$ runs into single events. Intuitively io $(P)$ represents the input-output behavior of a process $P$ if we execute that process in isolation, i.e. without interference from outside. This operation induces an interesting process equivalence, called IO-equivalence, and an associated implementation relation, denoted by $P$ sat $Q$.

$$
P \stackrel{I O}{=} Q \text { iff io }(P)=\mathrm{io}(Q), \text { and } P \stackrel{I O}{\text { sat }} Q \text { iff } i o(P) \subseteq i o(Q)
$$

Such equivalences play an important role in the book by K.R. Apt and E.-R. Olderog [AO]. Specification of what is often called the functional behavior of a process $P$ is really a specification of io $(P)$, i.e. of the 10 -equivalence class of $P$. The io(.) operation does (obviously) not distribute through parallel composition. For the case of layer composition though, we have the following laws:

$$
P \cdot Q \stackrel{I O}{=} \mathrm{io}(P) \cdot \mathrm{io}(Q) \text { and } P ; Q \stackrel{I O}{=} P \cdot Q
$$

The intuition here is that although execution of "layer" $P$ might overlap execution of "layer" $Q$ temporally, one can pretend that all of $P$, here represented as an atomic action io( $P$ ), precedes all of $Q$ as far as IO behavior is concerned.

### 2.1 Algebraic Laws

In this section we provide some algebraic laws for model informally introduced here and extensively studied in [JPZ91]. The well-known laws for sequential composition, such as associativity, are not stated.

Lemma 2.1
Commutativity and Associativity:

| $P \\| Q$ | $=Q \\| P$ | (COM1) |
| ---: | :--- | :--- |
| $P$ or $Q$ | $=Q$ or $P$ | (COM2) |
| $P \\|(Q \\| R)$ | $=(P \\| Q) \\| R$ | (ASSOC1) |
| $P \cdot(Q \cdot R)$ | $=(P \cdot Q) \cdot R$ | (ASSOC2) |
| $P$ or $(Q$ or $R)$ | $=(P$ or $Q)$ or $R$ | (ASSOC3) |

Distributivity:

$$
\begin{array}{rlll}
P \|(Q \text { or } R) & =(P \| Q) \text { or }(P \| R) & & \text { (DIST1) } \\
P \cdot(Q \text { or } R) & =(P \cdot Q) \text { or }(P \cdot R) & & \text { (DIST2) } \\
(P \text { or } Q) \cdot R & =(P \cdot R) \text { or }(Q \cdot R) & \text { (DIST3) }
\end{array}
$$

Idempotency:

$$
P \text { or } P=P
$$

Units and zeros:

| skip $\\| P$ | $=P \\|$ skip | $=P$ | (SKIP1) |
| :--- | :--- | :--- | :--- |
| skjp $P$ | $=P \bullet$ skip | $=P$ | (SKIP2) |
| empty or $P$ | $=P$ or empty | $=P$ | (EMPTY1) |
| empty $\\| P$ | $=P \\|$ empty | $=$ empty | (EMPTY2) |
| empty $P$ | $=P \bullet$ empty | $=$ empty | (EMPTY3) |

口
More interesting is the relationship between parallel composition and layer composition. We can formulate here a (simple form of) the principle of communication closed layers in the form of an algebraic law. The communication closed layers law (CCL) deviates somewhat from the usual style of algebraic laws in that there is a (syntactic) side condition that should be checked concerning conflicts between processes. Let act ( $P$ ) denote the (finite) set of actions that (syntactically) occur in a $D L$ process $P$. Let Act denote the set of actions, we extend the conflict relation on $\mathcal{A c t}$ to sets of Act elements as follows: for $X, Y \subseteq$ Act,
$X-Y$ iff there exist $a \in X, b \in Y$ such that $a-b$.
Conflicts between $D L$ processes are then defined thus: $P-Q$ iff act $(P)-\operatorname{act}(Q)$.
As usual, $P \not \subset Q$ denotes that $P-Q$ is not the case, i.e. $P$ actions do not conflict with $Q$ actions.
The CCL laws can now be formulated as follows.

## Lemma 2.2

Communication Closed Layers:
Provided that $P \nrightarrow S$, and $Q \not \subset$ :

$$
\begin{aligned}
(P \cdot Q) \|(R \cdot S) & =(P \| R) \cdot(Q \| S) & & \text { (CCL) } \\
(P \cdot Q) \| S & & =P \cdot(Q \| S) & \\
(P \cdot Q) \| R & =(P \| R) \cdot Q & & \text { (CLLL-R) } \\
Q \| R & =Q \cdot R & & \text { (Independence) }
\end{aligned}
$$

In order to state a generalized version of the CCL and related laws, we introduce the abbreviations:
for $i \leftarrow[n \ldots m]$ dopar $P(i)$ rof abbreviating: $P(n)\|\cdots\| P(m)$
for $i \leftarrow[n \ldots m]$ layer $P(i)$ rof abbreviating: $P(n) \bullet \ldots \bullet P(m)$
for $i \leftarrow[n \ldots m]$ choice $P(i)$ rof abbreviating: $P(n)$ or $\cdots$ or $P(m)$
for $i \leftarrow[n \ldots m]$ doseq $P(i)$ rof abbreviating: $P(n) ; \cdots ; P(m)$

## Lemma 2.3

- Generalized Communication Closed Layers Lawo.

Assume that if there is a conflict between $P_{i j}$ and $P_{k l}$ then either $i=k$ or $j=l$ is satisfied, for $1 \leq i, j, k, l \leq n$. Then
for $i+-[1, \ldots n]$ dopar
for $j \infty[1 \ldots m]$ layer $P_{i, j}$ rof
rof
$=$
for $j \leftarrow[1 \ldots m]$ layer
for $i \leftarrow[1 \ldots n]$ dopar $P_{i, j}$ rof
rof

- Left-Right Movers Law.

Assume that, for $1 \leq i<k \leq n$ and $1 \leq l<j \leq m, P_{i, j} \not \subset P_{k, l}$. Then for $i+[1 \ldots n]$ layer
for $j \leftarrow[1 \ldots m]$ layer $P_{i, j}$ rof rof
$=$
for $j \leftarrow[1 \ldots m]$ layer
for $i \leftarrow[1 \ldots n]$ layer $P_{i, j}$ rof rof

### 2.2 Normal Forms

Let $D L^{\prime}$ denote the language $D L$ with the sequential composition operator omitted. If $P$ is in $D L^{\prime}$ then for each $P$ run $(E, \rightarrow)$ the order $\rightarrow$ is generated by conflicts only. That is, if $e \in E$ and $e^{\prime}$ is a direct successor of $e$ with respect to $\longrightarrow$, then $e$ and $e^{\prime}$ are in conflict. (This is not the case if one adds the sequential composition to the language.)
Define layer $L(k)$ as

$$
L(k)=\{e \in E \mid \text { the longest chain below } e \text { has length } k\}
$$

Since runs are assumed to be finite, $L(k)=0$ for sufficient large $k$. Enumerate the elements in each layer $L(k)$ such that

$$
L(k)=\left\{e_{k, j} \mid 1 \leq j \leq l(k)\right\}
$$

where $l(k)$ is the number of elements in $L(k)$.
Observe that there can be no conflict between events in the same layer, and for all events $e$ in layer $L(k), k>1$, there exists an event $e^{\prime}$ in layer $L(k-1)$ such that $e-e^{\prime}$. The run $(E, \longrightarrow)$ is can be denoted syntactically by:

```
for }k\leftarrow[0\ldots.n] layer for j\leftarrow[1\ldots.l(k)] dopar e ck,j rof ro
```

where $n$ is such that

$$
(0 \leq k \leq n \Rightarrow L(k) \neq \emptyset) \wedge k>n \Rightarrow L(k)=0
$$

(Formally speaking $e_{k, j}$ is not an action but an event, and should replace $e_{k, j}$ by $\mu\left(e_{k, j}\right)$, where $\mu\left(e_{k, j}\right)$ is the action corresponding to the event $e_{k, j}$.) This can be done for every run of $P$, hence if we let $r$ denote the number of runs of $P$ then

```
P= for i\leftarrow[1_..r] choice
    for k\leftarrow[1\ldotsn(r)] layer for j\leftarrow[1\ldotsk(n(r))] dopar }\mp@subsup{e}{i,k,j}{}\mathrm{ rof rof
    rof
```

The above decomposition of $\boldsymbol{P}$ has the following properties:

- For each $j$ and $j^{\prime}$, with $j \neq j^{\prime}, e_{i, k, j} \nsucc e_{i, k, j^{\prime}}$.
- For each $k>1$ and each $e_{i, k, j}$ there exists an event $e_{i, k-1, j^{\prime}}$ such that $e_{i, k, j}-e_{i, k-1, j^{\prime}}$.

Hence we have written $P$ as a choice over layered maximal parallel processes. This above decomposition of $P$ leads to following definition
Definition 2.4 Nornal Form
A normal form is a process term of the form

```
for \(\boldsymbol{i} \leftarrow[1 \ldots r]\) choice
    for \(k \leftarrow[1 \ldots n(r)]\) layer
        for \(j \leftarrow[1 \ldots k(n(r))]\) dopar \(e_{i, k, j}\) rof
    rof
rof
```

where each $e_{i, k, j}$ is an elementary process term. Moreover the process term should satisfy

1. For each $j$ and $j^{\prime}$, with $j \neq j^{\prime}, e_{i, k, j}+e_{i, k, j^{\prime}}$.
2. For each $k>1$ and each $e_{i, k, j}$ there exists an event $e_{i, k-1, j^{\prime}}$ such that $e_{i, k, j}-e_{i, k-1, j^{\prime}}$.
3. All mutually distinct branches of the choice construct are syntactically different.

This normal form resembles the normal form for Mazurkiewicz traces.
It follows from the observations above that each process term in $D L^{\prime}$ has a normal form. But there is more to it:

## Theorem 2.5 Expansion theorem

Let $L_{P} \bullet P$ and $L_{Q} \bullet Q$ be process terms in $D L^{\prime}$ such that any two actions in $L_{P}$ and any two actions in $L_{Q}$ are non-conflicting. Furthermore assume that for every action $a_{P}^{\prime} \in P\left(a_{Q}^{\prime} \in Q\right)$ there exists an action $a_{P} \in L_{P}\left(a_{Q} \in L_{Q}\right)$ such that $a_{P}^{\prime} * a_{P}\left(a_{Q}^{\prime} * a_{Q}\right)$, where ${ }^{*}$ is the transitive closure of the conflict relation -. Then

```
    \(\left(L_{P}, P\right) \|\left(L_{Q} \bullet Q\right)=\)
        for \(a_{P} \in L_{P}\) choice \(a_{P} \bullet\left(\left(\left(L_{P}-\left\{a_{P}\right\}\right) \bullet P\right) \|\left(L_{Q} \bullet Q\right)\right)\) rof
    or
        for \(a_{Q} \in L_{Q}\) choice \(a_{Q} \bullet\left(\left(L_{P} \cdot P\right) \|\left(\left(L_{Q}-\left\{a_{Q}\right\}\right) \cdot Q\right)\right)\) rof
```

The expansion theorem, together with the Left-Right Movers law, is crucial for transforming every process term $P$ in $D L^{\prime}$ into normal form. The normal form is the maximal parallelization of $P$.

## Theorem 2.6

Each process term $P \in D L^{\prime}$ has an unique normal form (up to permutations of the indices). Moreover $P$ can be algebraically transformed into this normal form in an algorithmic way using the algebraic laws of section 2.1 and the expansion theorem, theorem 2.5.
A proof can be found in [PZ92].

## 3 The 'Point in polygon' algorithm: Pipelining on a tree network

The aim of this section is to show how to transform an initial, algorithmic design to a form that is suitable for implementation on a pipelined architecture. The (functional) correctness of the initial design is not our concern; we assume that it is the result from an initial design stage where it has been developed from a specification of the required functional behavior. We concentrate on the stage following the initial design phase, where not only functional correctness is of importance, but where also the architecture of the implementation must be taken into account.

The functional specification of the algorithm, that we do not formalize here, amounts to the following: Given a (fixed) polygon with edges $E_{1}, \cdots, E_{N}$ and a set $\left\{p_{0}, p_{1}, \cdots, p_{n}\right\}$ of points in a two dimensional Euclidean space, it is required to determine which of the points in $\left\{p_{0}, p_{2}, \ldots, p_{m}\right\}$ lay inside the polygon.
Our initial design has been essentially taken over from [Ak189]. It is based on the fact that a point $p$ lies inside the polygon if and only if the vertical line through $p$ intersects an odd number of edges above $p$. Thus the problem reduces to computing the number of intersections above $p$. The intuition for the algorithm in [Ak189] is as follows. Assuming that $N+1$ is a power of 2, put $s=\log (N+1)$. A tree-like arrangement of processes of depth $s$ is used, containing one process for each edge. The coordinates of the points are stored in an array $c$, with $c[a]$ the coordinates of point $p_{a r} a=0, \ldots, m$. The coordinates of the candidate points $p$ are read by the top process and are broadcast during so called descend phases. A separate descend phase is executed for each 'level' in the tree, starting with the top node and ending with the level consisting of all leaf nodes. Each process receives the coordinates of the candidate point, determines locally whether the vertical line through $p$ intersects the edge associated with that process, and broadcasts the results "downwards" to its children. After these descend phases, the total number of intersections is calculated during a number of ascend phases, where processes add together partial counts calculated by their children. If this count reaches the top process, then this top process assigns the appropriate boolean value to inside $[a]$, where inside is a boolean array, such that inside[a] holds if and only if the point $p_{a}$ lies in the polygon. The algorithm in [Ak189] is presented in the form of a sequential composition of "layers" $D(l)$ and $A(l)$, each consisting of independent, parallel executed, actions. Communication between processes is by means of shared variables, where variables $d_{i}$ and $p_{i}$ are used during the descend phases and variables $u_{i}$ during ascend phases. (Apart from these, there are local variables $s_{i}, t_{i}$, and $q_{i} . q_{i}$ is used to store a local copy of coordinates of the point under consideration) Each descend layer $D(l)$ is itself divided into a reading phase $D^{R}(l)$ and a writing phase $D^{W}(l)$, and similarly, $A(l)$ is split into $A^{R}(l)$ and $A^{W}(l)$. The layer corresponding to the "leaf" nodes is an exception; there is no writing for the descend phase, and no reading for the ascend phase. The algorithm given in Akl [Akl89], adapted to our notation, is:

```
Program 0
for \(a \leftarrow[0 \ldots m]\) doseq
    \(d_{1}:=0 ; p_{1}:=c[a] ;\)
    for \(l-[1 \ldots s-1]\) doseq \(D^{R}(l) ; D^{W}(l)\) rof ;
    \(D^{R}(s) ; A^{W}(s) ;\)
    for \(l \leftarrow[s-1 \ldots 1] \operatorname{doseq} A^{R}(l) ; A^{W}(l)\) rof ;
    inside \([a]:=\operatorname{odd}\left(u_{1}\right)\)
```

rof

This describes the "layered" structure of the algorithm. The layers $D^{R}(l), D^{W}(l), A^{R}(l)$ and $A^{W}(l)$ can each be described by means of parallel composition of independent actions. This implies that there is no interference among parallel processes, and consequently the (functional) correctness of the algorithm can be shown relying essentially on techniques for sequential programs, as explained for instance in [AO91]. The reason for this is that for independent processes $P$ and $Q$, parallel composition $P \| Q$ is (semantically) identical to layer composition $P \bullet Q$; the latter in turn is, though not identical, lO-equivalent to sequential composition $P ; Q$.
The layers for the algorithm presented in [Ak189] can be specified as follows
The descending read phase for level $l$ :

$$
\begin{aligned}
D^{n}(l)= & \text { for } j \leftarrow\left[2^{l-1} \ldots 2^{1}-1\right] \text { dopar } \\
& \text { if } \operatorname{Intersects}\left(e_{j}, p_{j}\right) \text { then } s_{j}:=d_{j}+1 \text { else } s_{j}:=d_{j} \text { fi } \| q_{j}:=p_{j} \\
& \text { rof }
\end{aligned}
$$

The descending write phase:

$$
\begin{aligned}
& D^{W}(l)= \\
& \text { for } j \leftarrow\left[2^{l-1} \ldots 2^{l}-1\right] \text { dopar } p_{2 j}:=q_{j}\left\|p_{2 j+1}:=q_{j}\right\| d_{2 j}:=s_{j} \| d_{2 j+1}:=0 \text { rof }
\end{aligned}
$$

The ascending read phase:

$$
A^{n}(l)=\text { for } j \leftarrow\left[2^{l-1} \ldots 2^{l}-1\right] \text { dopar } t_{j}:=u_{2 j}+u_{2 j+1} \text { rof }
$$

Finally the ascending write phase is given by:

$$
A^{W}(l)=\text { for } j \leftarrow\left[2^{i-1} \ldots 2^{\prime}-1\right] \text { dopar } u_{j}:=t_{j} \text { rof }
$$

except for $l=s$, in that case:

$$
A^{W}(s)=\text { for } j \leftarrow[(N+1) / 2 \ldots N] \text { dopar } u_{j}:=s_{j} \text { rof }
$$

The presentation in [Akl89] is in terms of sequential composition of layers. From the point of view of functional correctness this is unnecessary; replacing all sequential composition by layer composition results in a process that is 10 -equivalent. Moreover, this allows for algebraic manipulation that would be invalid for the version based on sequential composition. In particular, the layer composition version allows for overlapping execution of different layers, resulting in a pipelined execution where layers that have to be executed for different candidate points are executed in parallel. Thus, we will use, as starting point for a series of transformations, the following version:

## Program 1

for $a \leftarrow[0 \ldots m]$ layer

$$
G(a)
$$

$$
\text { for } l \leftarrow[1 \ldots s-1] \text { layer } D^{R}(l) \bullet D^{W}(l) \text { rof }
$$

$$
D^{R}(s) \cdot A^{W}(s)
$$

$$
\text { for } l \leftarrow[s-1 \ldots 1] \text { layer } A^{R}(l) \bullet A^{W}(l) \text { rof }
$$

$$
P(a)
$$

rof
where

$$
G(a)=d_{1}:=0 \| p_{1}:=c[a] \text { and } P(a)=\text { inside }[a]:=\operatorname{odd}\left(u_{1}\right)
$$

First we will rename the processes $D^{R}(l), D^{W}(l), A^{R}(l)$ and $A^{W}(l)$, according to read or write actions. Put $t=2 s$ and define

$$
\begin{aligned}
& R(l)= \begin{cases}D^{R}(l) & \text { if } 1 \leq l \leq s \\
A^{R}(t-l) & \text { if } s+1 \leq l \leq t-1\end{cases} \\
& W(l)= \begin{cases}D^{W}(l) & \text { if } 1 \leq l \leq s-1 \\
A^{W}(t-l) & \text { if } s \leq l \leq t-1\end{cases}
\end{aligned}
$$

Then Program 1 can be rewritten as

## Program 1'

```
for \(a \leftarrow[0 \ldots m]\) layer
    \(G(a)\).
    for \(l \leftarrow[1 \ldots t-1]\) layer \(R(l) \cdot W(l)\) rof .
        \(P(a)\)
rof
```

Conceptually we chanced the tree structure in a reflected tree structure, two identical trees with the leaves merged, such that all the data flows from top to bottom.

Observe that the above program is actually in pseudo normal form. It can be transformed into normal form, using the Left-Right Movers law and the Independence law, where one should check the side-conditions for these laws based on the following pattern of conflicts:

$$
R(j)-R(j), W(j)-W(j), R(j)-W(j) \text { and } R(j)-W(j-1)
$$

Indeed, using that skip is a unit for layer composition, the above program is (semantically) equal to:

$$
\text { for } a \leftarrow[0 \ldots m] \text { layer for } l \leftarrow[0 \ldots m+t] \text { layer } \tilde{R}(a, l) \cdot \tilde{W}(a, l) \text { rof rof }
$$

where

$$
\begin{aligned}
& \tilde{R}(a, b)= \begin{cases}P(a) & \text { if } l=a+t \\
R(l-a) & \text { if } a<l<a+t \\
\text { skip } & \text { otherwise }\end{cases} \\
& \tilde{W}(a, l)= \begin{cases}G(a) & \text { if } l=a \\
W(l-a) & \text { if } a<l<a+t \\
\text { skip } & \text { otherwise }\end{cases}
\end{aligned}
$$

The side-conditions for the Left-Right Movers Law are fulfilled, hence the program can be transformed into

$$
\text { for } l \leftarrow[0 \ldots m+t] \text { layer for } a \leftarrow[0 \ldots m] \text { layer } \tilde{R}(a, l) \cdot \tilde{W}(a, l) \text { rof rof }
$$

Again applying the Left-Right Movers and Independence Law to the inner layered loop gives that the program equals

## Program 2

$$
\begin{aligned}
& \text { for } l \leftarrow[0 \ldots m+t] \text { layer } \\
& \quad \text { for } a \leftarrow[0 \ldots m] \text { dopar } \tilde{R}(a, l) \text { rof } \\
& \text { for } a \leftarrow[0 \ldots m] \text { dopar } \tilde{W}(a, l) \text { rof } \\
& \text { rof }
\end{aligned}
$$

This last transformation results, for large enough $m$ and again using that skip is unit for layer composition, in a program consisting of a phase where the tree is gradually filled, a phase where all the nodes of the tree execute simultaneously, and a final phase where the tree is emptied. For instance the layered loop
for $a+[0 \ldots m]$ dopar $\tilde{R}(a, l)$ rof
equals, by definition of $\tilde{R}(a, b)$

$$
\text { if }(l-t) \geq 0 \text { then } P(l-t) \text { \& } \| \text { for } a \leftarrow[\max (0, l-t+1) \ldots \min (m, l-1)] \operatorname{dopar} R(l-a) \text { rof }
$$ which on it's turn equals, by substituting $i=l-a$

$$
\text { if }(l-t) \geq 0 \text { then } P(l-t) \text { fi } \| \text { for } i-[\max (1, l-m) \ldots \min (t-1, l)] \text { dopar } R(i) \text { rof }
$$

If we apply similar substitutions to the layered loop with $\widetilde{W}(a, l)$, the resulting program is

## Program 3

$$
\begin{aligned}
& \text { for } l \leftarrow[0 \ldots m+t] \text { layer } \\
& \quad \text { (if }(l-t) \geq 0 \text { then } P(l-t) \text { п } \| \\
& \text { for } i \leftarrow[\max (1, l-m) \ldots \min (t-1, l)] \text { dopar } R(i) \text { rof }) \cdot \\
& (\text { if } l \leq m \text { then } G(b) \text { f } \| \\
& \text { for } i \leftarrow[\max (1, l-m) \ldots \min (t-1, l)] \text { dopar } W(i) \text { rof }) \\
& \text { rof }
\end{aligned}
$$

The variable boundaries at the inner parallel loops are due to "filling" of the tree, for $0 \leq t \leq t-1$, and "emptying" of the tree, for $m+1 \leq l \leq m+t$. The "filling" of the tree consists of two phases
analogous to the descend and ascend phases in the original algorithm. First the information is broadcast "downwards", layer after layer, into the tree, until this downwards fill reaches the leaves. Then the results are broadcast "upwards", again layer after layer, towards the root of the tree, until it reaches layer 1. At that moment the tree is filled, this is the case for $l=t-1$. Thereafter all cells compute in parallel for $t \leq l \leq m$ and afterwards the tree is emptied in the reverse order for $m<l \leq m+t$.

Although this program admits pipelined execution, it has the disadvantage that at each moment of time there is a different set of active processes. This is well known for pipelining in general: there is a phase where the pipeline is gradually 'filled', a phase where all processes in the pipeline are simultaneously executing, and a final phase where the pipeline is 'emptied'. This picture of "filling" and "emptying" is adequate on the abstraction level of processes only. On a low level where allocation of processes to processors is considered, there is no corresponding "starting" and "halting" of processors. Rather, during the filling and emptying phase there will be processors executing the same algorithm as others, but on non-relevant data so to say. We model this by adding 'dummy' actions to Program 3, in order to obtain a regular pattern.

## Program 4

$$
\begin{aligned}
& \text { for } l-[0 \ldots m+t] \text { layer } \\
& \qquad \text { (if }(l-t) \geq 0 \text { then } P(l-t) \text { f } \| \\
& \text { for } j \leftarrow[1 \ldots t-1] \text { dopar } R(j) \text { rof }) \\
& \text { (if } l \leq m \text { then } G(l) \text { f } \| \\
& \text { for } j \leftarrow[1 \ldots t-1] \text { dopar } W(j) \text { rof ) }
\end{aligned}
$$

rof
Although this program doesn't semantically equal program 3, it has the property that it preserves functional correctness. More precisely program 4 projected onto the variables $c$ and inside is semantically equal to program 3 projected onto the variables $c$ and inside. Hence program 4 has the IO behavior with respect to $c$ and inside as the original program 1 . This can be seen by applying the inverse of the transformation steps above to program 4.
Next we take a step in the design that no longer preserves semantic equality, but only lO-equivalence, by imposing extra order by replacing the layer composition by a corresponding sequential composition in program 4. This cannot affect functional correctness, but it does allow for allocation of processes on (sequentially executing) processors.

## Program 6

$$
\begin{aligned}
& \text { for } l \leftarrow[0 \ldots m+t] \text { doseq } \\
& \qquad \text { if }(l-t) \geq 0 \text { then } P(l-t) \text { f } \| \text { for } j \leftarrow[1 \ldots t-1] \text { dopar } R(j) \text { rof }) \text {; } \\
& (\text { if } l \leq m \text { then } G(l) \text { п } \| \text { for } j \leftarrow[1 \ldots t-1] \text { dopar } W(j) \text { rof })
\end{aligned}
$$

Note that for each sequential phase, there are at most $(3 N+1) / 2$ and at least $(3 N-1) / 2$ processes active, which suggest a rather obvious allocation onto $(3 N+1) / 2$ processors. In essence there are $(3 N+1) / 2$ processes executing in parallel their contribution to some read phase and afterwards there are $(3 N+1) / 2$ processes executing part of some ascend phase. Note that the structure of the program matches the class of SIMD machines, where a number of processors execute in lockstep the same (parameterized) program.
Applying the Independence law and commutativity of parallel composition to the program fragment

$$
\begin{aligned}
& \quad \text { (if }(l-t) \geq 0 \text { then } P(l-t) \text { i } \| \text { for } j \leftarrow[1 \ldots t-1] \text { dopar } R(j) \text { rof ) } \\
& \quad \text { (if } l \leq m \text { then } G(l) \text { fi } \| \text { for } j \leftarrow[1 \ldots t-1] \text { dopar } W(j) \text { rof }) \\
& \text { yields the following program fragment }
\end{aligned}
$$

```
for \(j \leftarrow[1 \ldots s]\) dopar \(R(j)\) rof
(for \(j \leftarrow[1 \ldots s-1]\) dopar \(W(j)\) rof \(\|\) if \(l \leq m\) then \(G(l)\) i).
(for \(j \leftarrow[s+1 \ldots t-1]\) dopar \(R(j)\) rof \(\|\) if \((l-t) \geq 0\) then \(P(l-t)\) a).
for \(j \leftarrow[s \ldots t-1]\) dopar \(W(j)\) rof
```

Invoking the definition of $R(j)$ and $W(j)$ and replacing layer composition by sequential composition results in $N+1$ processes that alternate between four phases, corresponding to the read/write and the descend/ascend phase. As before, the algorithm matches an SIMD architecture. The final results is

## Program 7

```
for \(l \leftarrow[0 \ldots m+t]\) doseq
        for \(j+[1 \ldots s]\) dopar \(D^{R}(j)\) rof ;
    ( for \(j \leftarrow[1 \ldots s-1]\) dopar \(D^{W}(j)\) rof \(\|\) if \(l \leq m\) then \(G(l)\) fi);
    (for \(j \leftarrow[1 \ldots s-1]\) dopar \(A^{R}(j)\) rof \(\|\) if \((l-t) \geq 0\) then \(\left.P(l-t) f i\right)\);
    for \(j \leftarrow[1 \ldots s]\) dopar \(A^{W}(j)\) rof
    rof
```

In this program each cell, Cell( $i), 1 \leq i \leq N$, executes the following program:

```
\(\operatorname{Cell}(i)=\)
    for \(l \leftarrow[0 \ldots m+t]\) doseq
        \(q_{i}:=p_{i} ;\)
        if Intersects \(\left(e_{i}, q_{i}\right)\) then \(s_{i}:=d_{i}+1\) else \(s_{i}:=d_{i} \mathbf{f}\);
        if \(i<2^{n-1}\) then \(p_{2 i}:=q_{i} ; p_{2 i+1}:=q_{i} ; d_{2 i}:=s_{i} ; d_{2 i+1}:=0\) fi;
        if \(i<2^{n-1}\) then \(t_{i}:=u_{2 i}+u_{2 i+1}\) fi;
        if \(i<2^{s-1}\) then \(u_{i}:=t_{i}\) else \(u_{i}:=s_{i}\) f
    rof
```

It's clear that in this program for Cell( $i$ ) we can take the local variables $s_{i}$ and $t_{i}$ equal.

## 4 The All-Points Shortest Path Problem

We have given some weighted directed graph of $n$ nodes. The weights of the edge from node $i$ to node $j$ is given as $w_{i j}$, where $w_{i j} \geq 0$. If there is no edge from $i$ to $j$ in the graph, then we add one with weight $w_{i j}=\infty$.
The problem is to find the shortest distance $d_{i j}$ from node $i$ to $j$ for all $i$ and $j$. A path from node $i$ to node $j$ is a sequence of nodes ( $i_{0} i_{1} \ldots i_{k}$ ) where $i_{j} \in 1 . . n$ for $j \in 0 . . k$ and where $i_{0}=i$ and $i_{k}=j$. The length of a path $\left(i_{0} i_{1} \ldots i_{k}\right)$ is the number of edges $k$, whereas the distance along it (its weight) is $w_{i_{0} i_{1}}+w_{i_{1} i_{2}}+\cdots+w_{i_{n-1} i_{k}}$. Note that since all weights are non-negative the shortest path from $i$ to $j$ is well defined, it will never cross itself, and so has length smaller than $n$. We denote the collection of paths from $i$ to $j$ with length smaller than $m$ by path(i,j,m). Then the distance $d_{i j}$ along the shortest path from $i$ to $j$ is defined as:

$$
i=j \rightarrow d_{i j}=0, \text { and } i \neq j \rightarrow d_{i j}=\min \left\{w_{i_{0} i_{1}}+w_{i_{1} i_{2}}+\cdots+w_{i_{n-1} i_{k}}\right\}
$$

where the minimum is over all sequences in $\operatorname{path}(i, j, n)$. A simple fact that follows directly from the definition: If node $p$ is on the shortest path from $i$ to $j$, then the subpaths from $i$ to $p$ and from $p$ to $j$ are the shortest paths from $i$ to $p$ and from $p$ to $j$. Moreover, $d_{i j}=d_{i p}+d_{p j}$. (For instance, if the subpath from $i$ to $p$ would not be the shortest one, then we could improve the $i-j$ path, contrary to the assumption that it was the shortest $i-j$ path.)

A well-known sequential algorithm for computing the shortest path, based on the above observations, is the Floyd-Warshall, cf. [Akl89], Chapter 10.

```
Program Floyd-Warshall
    for \(i \leftarrow[1 \ldots n]\) doseq
        for \(j \leftarrow[1 \ldots n]\) doseq \(m[0, i, j]:=w[i, j]\) rof
    rof ;
    for \(k \leftarrow[1 \ldots n]\) doseq
        for \(i \leftarrow[1 \ldots n]\) doseq
        for \(j \leftarrow[1 \ldots n]\) doseq \(m[k, i, j]:=\min \{m[k-1, i, j], m[k-1, i, k]+m[k-1, k, j]\}\) rof
        rof
    rof
```

An invariant of the above program is; $m[k, i, j]$ is the shortest path from node $i$ to node $j$ with all the intermediate nodes taken from $\{1, \ldots, k\}$. From which the correctness easily follows. Put

$$
e_{i, k, j}=m[k, i, j]:=\min \{m[k-1, i, j] ; m[k-1, i, k]+m[k-1, k, j]\}
$$

The Floyd-Warshall is a sequential algorithm where all actions are ordered in time. Again from point of functional correctness this is unnecessary; we can replace each sequential composition by a layer composition to get a process which is 10 -equivalent to the initial process:

```
for \(i \leftarrow[1 \ldots n]\) layer
            for \(j \leftarrow[1 \ldots n]\) layer \(m[0, i, j]:=w[i, j]\) ro \([\)
rof.
for \(k \leftarrow[1 \ldots n]\) layer
            for \(i \leftarrow[1 \ldots n]\) layer
                for \(j \leftarrow[1 \ldots n]\) layer \(\left(e_{i, k, j}\right)\) rof
            rof
rof
```

This process without sequential composition allows for algebraic transformations towards a specific architecture. Take for instance a CREW-PRAM, then there is only a conflict between $e_{i, k, j}$ and $e_{i, k^{\prime}, j}$. This yields, invoking the Left-Right Movers law, in that case the following normal form:

## Program Floyd-Warshall Normal Form I

for $\boldsymbol{i} \leftarrow[1 \ldots n]$ dopar
for $j \leftrightarrow[1 \ldots \pi]$ dopar $m[0, i, j]:=w[i, j]$ rof
rof.
for $k \leftarrow[1 \ldots n]$ layer
for $i \leftarrow[1 \ldots n]$ dopar for $j \leftarrow[1 \ldots n]$ dopar $\left(e_{i, k, j}\right)$ rof
rof
rof
This leads to an implementation on a CREW-PRAM by changing every layer composition in a sequential composition, with complexity $O(n)$ time, on $O\left(n^{2}\right)$ processors. This is an optimal parallelization of the sequential Floyd-Warshall algorithm.
If we consider a EREW-PRAM then for fixed $k e_{i, k, j}$ is in conflict with $e_{i^{\prime}, k, j}$, if and only if $i=i^{\boldsymbol{t}}$ or $j=j^{\prime}$, because in that case they want to read both the same shared variable. The conflict order imposed by the sequential Floyd-Warshall algorithm is, again for fixed $k$

$$
e_{i, k, j} \longrightarrow e_{i^{\prime}, k, j} \text { iff } i<i^{\prime} \text { and } e_{i, k, j} \rightarrow e_{i, k, j^{\prime}} \text { iff } j<j^{\prime}
$$

cf. figure 1 , where one has to take the transitive closure in each row and column.


Figure 1: The causal ordering and layering for fixed $k$.
If we write this in pseudo normal form the computation becomes
Program Floyd-Warshall Nornal Form II
for $i \leftarrow[1 \ldots n]$ dopar
for $j \leftarrow[1 \ldots n]$ dopar $m[0, i, j]:=w[i, j]$ rof
rof -
for $k \leftarrow[1 \ldots n]$ layer
for $l-[1 \ldots 2 * n-1]$ layer $L(k, l)$ rof
rof
where for $l \leq n$

$$
L(k, l)=\text { for } j \leftarrow[1 \ldots l] \text { dopar }\left\langle e_{j, k, l+1-j}\right\rangle \text { rof }
$$

and for $n<l \leq 2 * n-1$

$$
L(k, l)=\text { for } j \mapsto[1 \ldots 2 * n-l] \text { dopar }\left\langle e_{l-n+j, k, n+1-j}\right\rangle \text { rof }
$$

which leads to an algorithm with time complexity $O\left(n^{2}\right)$ and processor complexity $O(n)$ on an EREW-PRAM, which is again optimal.
Observe that the causal ordering on the $e_{i, k, j}$ for fixed $k$ is only induced by Read/Read conflicts, which means that the choice of the ordering doesn't influence the total result of the computation. Hence we can take another minimal conflict closed ordering, for instance the one given in figure 2.
Writing this computation in a pseudo normal form gives:
Program Floyd-Warshall Nornal Forn III

```
for \(i \leftarrow[1 \ldots n]\) dopar
            for \(j \leftarrow[1 \ldots n]\) dopar \(m[0, i, j]:=w[i, j]\) rof
    rof -
    for \(k \leftarrow[1 \ldots n]\) layer
            for \(l t-[1 \ldots n]\) layer \(L^{\prime}(k, l)\) rof
    rof
```

where

$$
L^{\prime}(k, l)=\text { for } i \leftarrow[1 \ldots n] \text { dopar }\left\langle e_{i, k, i \oplus 1}\right\rangle \text { rof }
$$

with

$$
i \oplus l=(n+l-i) \bmod n+1
$$



Figure 2: Another causal ordering for fixed $k$ and the corresponding layering.

## 5 Conclusion

A language is presented which makes clear distinction between temporal order and causal order. This allows for a transformational approach, in an algebraic way, for the specification and verification of concurrent systems. Each process term without sequential composition can be transformed, in an algorithmic way, using the algebraic laws, into a normal form. This normal form is the maximal parallelization of the process under consideration. The techniques involved are exemplified by some examples from pipelining on a tree network, and the Floyd-Warshall algorithm for the all-points shortest path.

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