

Synchronous Modeling of Data-Intensive Applications

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This talk :

- ▶ Detailed presentation of ARRAY-OL/GASPARD
- ▶ First results of study



Plan

Introduction

- GASPARD methodology
- General scheme

Data-intensive processing

- ARRAY-OL language
- Existing works

Simple synchronous modeling of GASPARD models

- Parallel model
- Serialized model

Validation issues

Conclusions



Introduction

Context : data-intensive applications (**DIA**) in embedded systems

- ▶ regular multidimensional data processing
- ▶ parallel processing in System-on-Chip (**SoC**)

Motivations : adequate techniques for

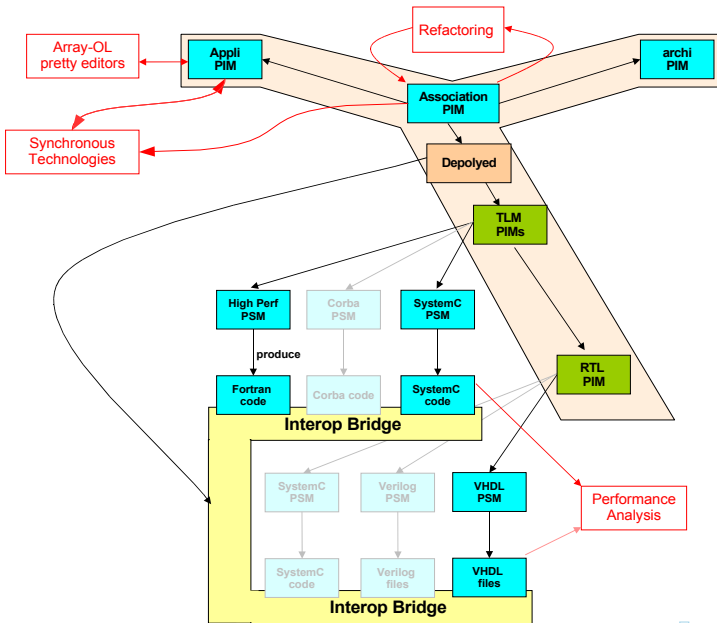
- ▶ efficient data manipulation
- ▶ analysis of implementation properties

Approach : combination of

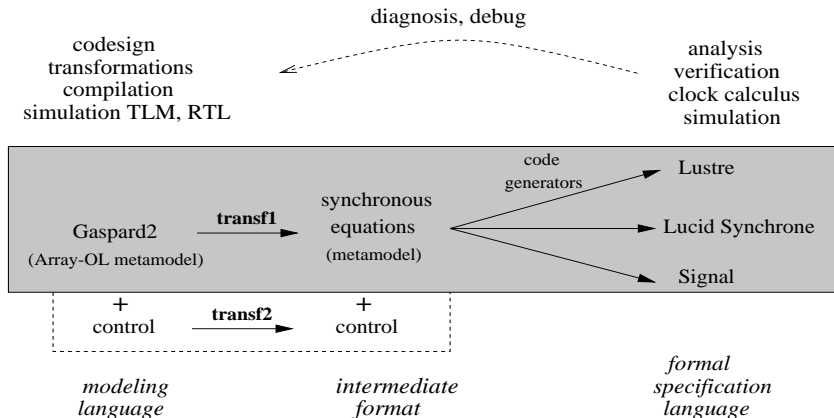
- ▶ a formalism dedicated to DIA (**ARRAY-OL**)
- ▶ data-flow synchronous equation models



GASPARD methodology



General scheme



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ARRAY-OL (Array-Oriented Language) : initially proposed by Thomson Marconi Sonar [DD98]

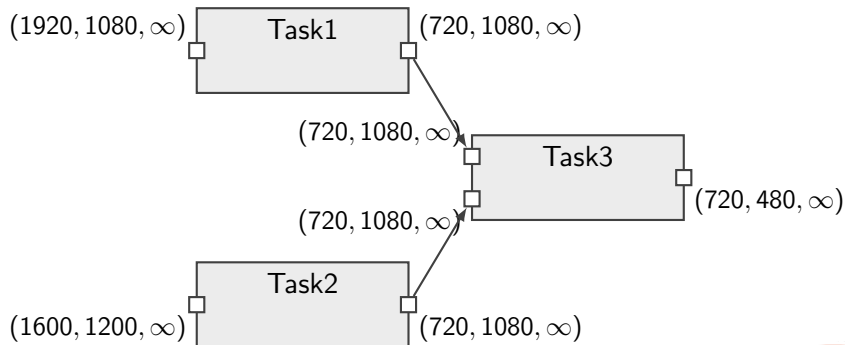
- ▶ Specification language for full parallelism
- ▶ Data manipulation through arrays
- ▶ Deadlock free and deterministic by construction

ARRAY-OL (Array-Oriented Language) : initially proposed by Thomson Marconi Sonar [DD98]

- ▶ Specification language for full parallelism
- ▶ Data manipulation through arrays
- ▶ Deadlock free and deterministic by construction
- ▶ Descriptions independent from implementation platforms
- ▶ Two types of parallelism in application specifications :
Task parallelism and Data parallelism

ARRAY-OL (cont'd)

Task parallelism and data dependencies :



Different task models :

- ▶ **Elementary task** : atomic computation block
(instantaneous function)

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Different task models :

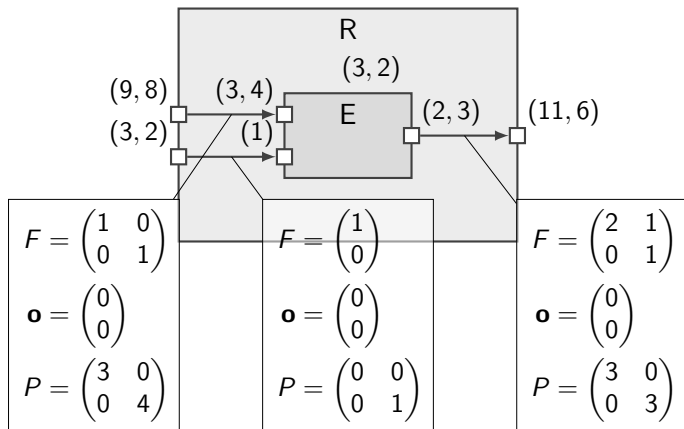
- ▶ **Elementary task** : atomic computation block (instantaneous function)
- ▶ **Hierarchical task** : task represented by hierarchical acyclic graphs in which
 - ▶ each node consists of a task, and
 - ▶ edges are labeled by the arrays
- ▶ **Repetition task** : expression of data parallelism

Data parallelism

- ▶ **Repetition element** : the subtask to be repeated
- ▶ **Repetition space** : limitation of repetition number and link between inputs and outputs
- ▶ **Interface** : input and output arrays
- ▶ **Tiler** : defines how to obtain sub-arrays from a input array and how to store sub-arrays in a output array

ARRAY-OL (cont'd)

Example of a repetition task



ARRAY-OL (cont'd)

Tiler specification :

- ▶ o : original point of the array or reference pattern
- ▶ P : paving matrix (how the array is tiled by patterns)
- ▶ F : fitting matrix (how patterns are filled by array elements)

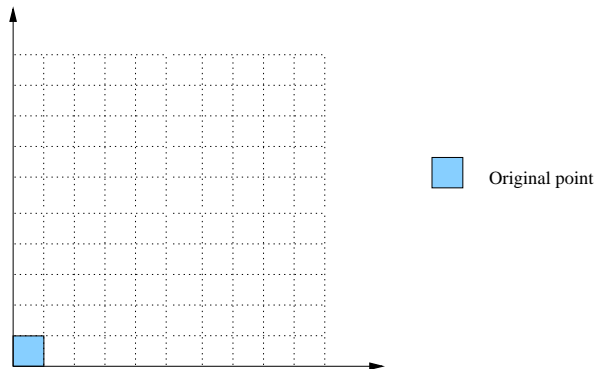


ARRAY-OL (cont'd)

Paving : how the array is tiled by patterns.

These patterns are calculated in **any order**.

Paving example : $o = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$, $P = \begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix}$



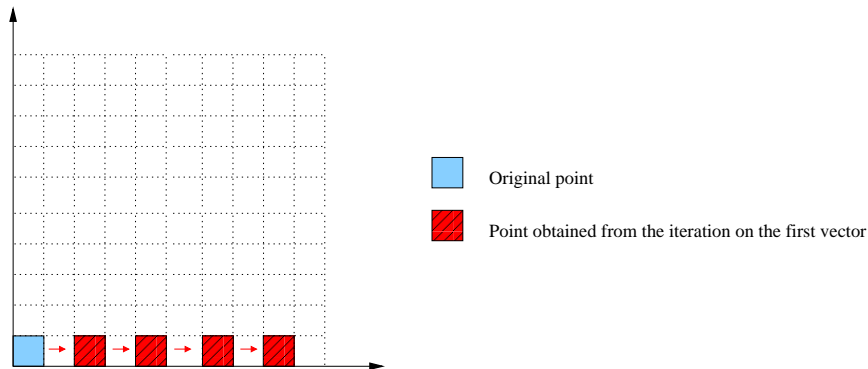
Repetition space : $[5,4]$, limitation of pattern repetitions.

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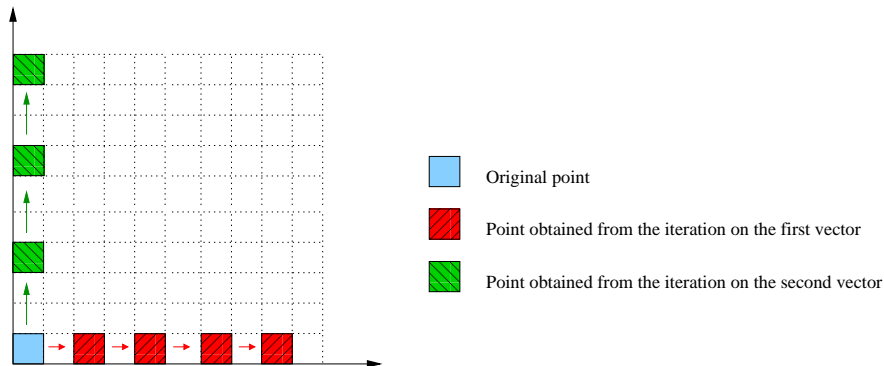
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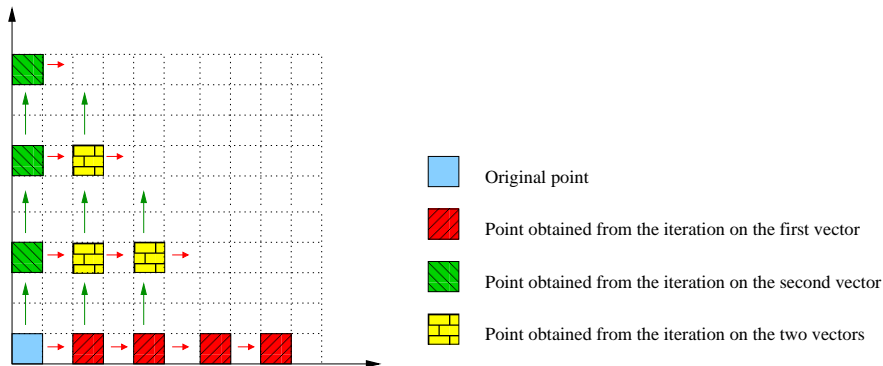
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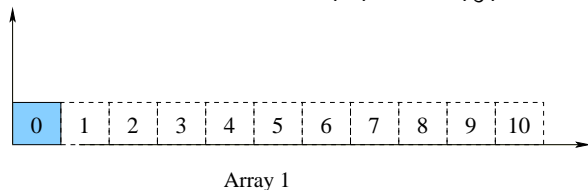


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ARRAY-OL (cont'd)

Fitting : how each pattern is filled by array elements.

Fitting example 1 : $o = (0)$, $F = (\frac{1}{3})$



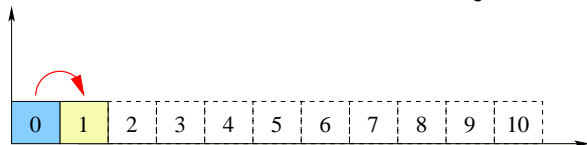
0		

Pattern

ARRAY-OL (cont'd)

Fitting : how each pattern is filled by array elements.

Fitting example 1 : $o = (0)$, $F = (\frac{1}{3})$



Array 1

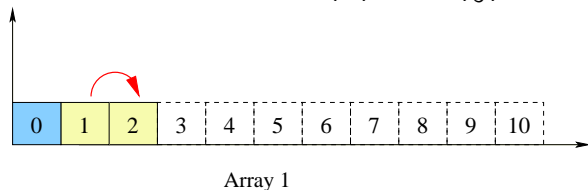
0	1	

Pattern

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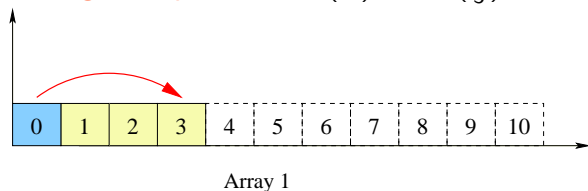
0	1	2

Pattern

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Fitting example 1 : $o = (0)$, $F = (\frac{1}{3})$



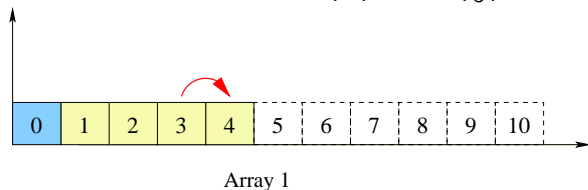
3		
0	1	2

Pattern

ARRAY-OL (cont'd)

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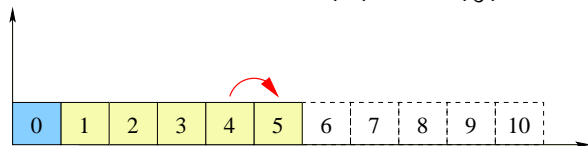
3	4	
0	1	2

Pattern

ARRAY-OL (cont'd)

Fitting : how each pattern is filled by array elements.

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Array 1

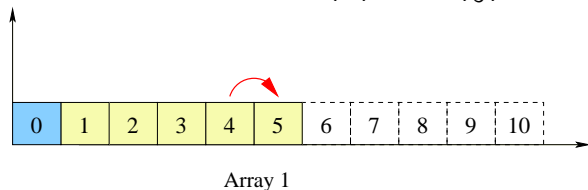
3	4	5
0	1	2

Pattern

ARRAY-OL (cont'd)

Fitting : how each pattern is filled by array elements.

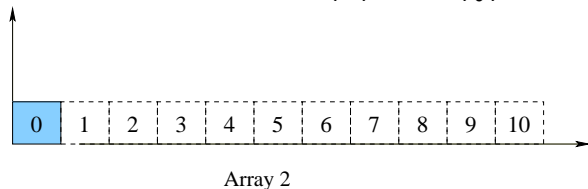
Fitting example 1 : $o = (0)$, $F = (\frac{1}{3})$



3	4	5
0	1	2

Pattern

Fitting example 2 : $o = (0)$, $F = (\frac{2}{6})$



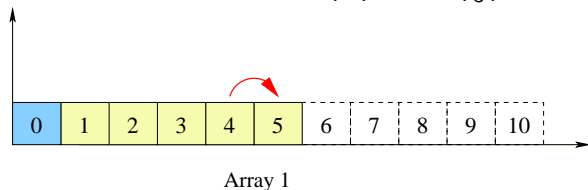
0		

Pattern

ARRAY-OL (cont'd)

Fitting : how each pattern is filled by array elements.

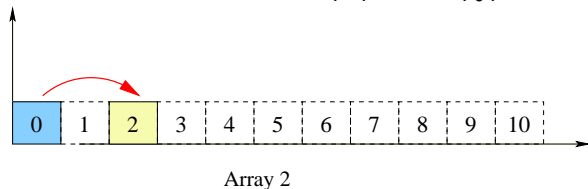
Fitting example 1 : $o = (0)$, $F = (\frac{1}{3})$



3	4	5
0	1	2

Pattern

Fitting example 2 : $o = (0)$, $F = (\frac{2}{6})$



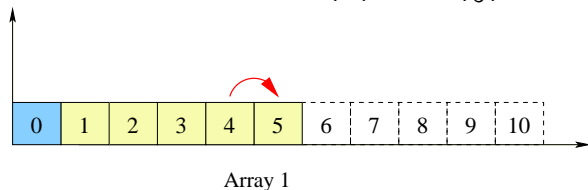
0	2	

Pattern

ARRAY-OL (cont'd)

Fitting : how each pattern is filled by array elements.

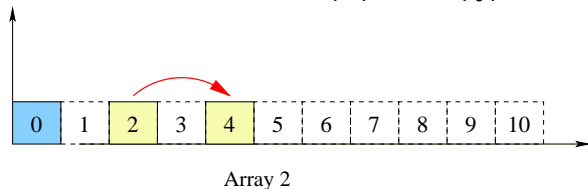
Fitting example 1 : $o = (0)$, $F = (\frac{1}{3})$



3	4	5
0	1	2

Pattern

Fitting example 2 : $o = (0)$, $F = (\frac{2}{6})$



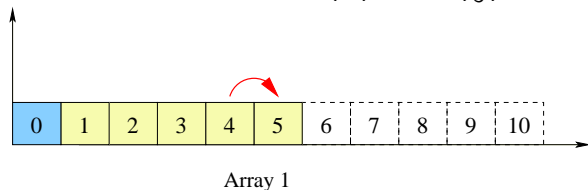
0	2	4

Pattern

ARRAY-OL (cont'd)

Fitting : how each pattern is filled by array elements.

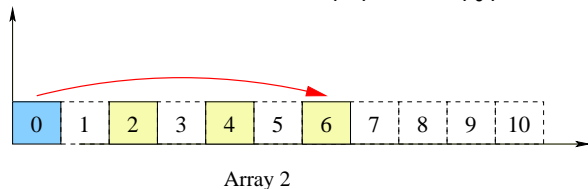
Fitting example 1 : $o = (0)$, $F = (\frac{1}{3})$



3	4	5
0	1	2

Pattern

Fitting example 2 : $o = (0)$, $F = (\frac{2}{6})$



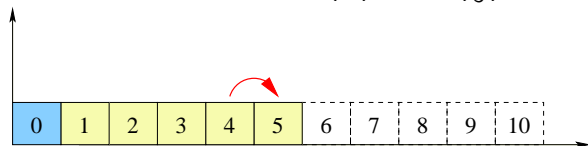
6		
0	2	4

Pattern

ARRAY-OL (cont'd)

Fitting : how each pattern is filled by array elements.

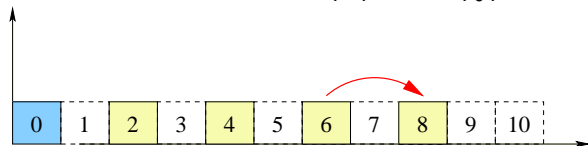
Fitting example 1 : $o = (0)$, $F = (\frac{1}{3})$



3	4	5
0	1	2

Pattern

Fitting example 2 : $o = (0)$, $F = (\frac{2}{6})$



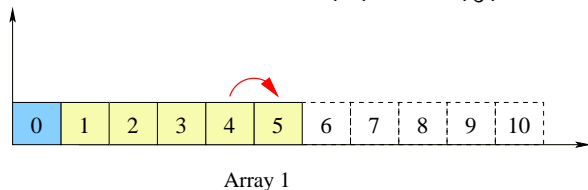
6	8	
0	2	4

Pattern

ARRAY-OL (cont'd)

Fitting : how each pattern is filled by array elements.

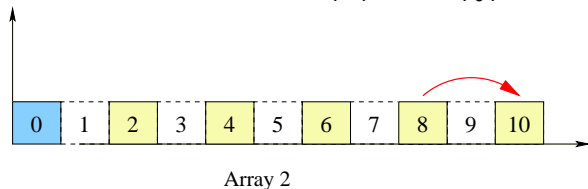
Fitting example 1 : $o = (0)$, $F = (\frac{1}{3})$



3	4	5
0	1	2

Pattern

Fitting example 2 : $o = (0)$, $F = (\frac{2}{6})$

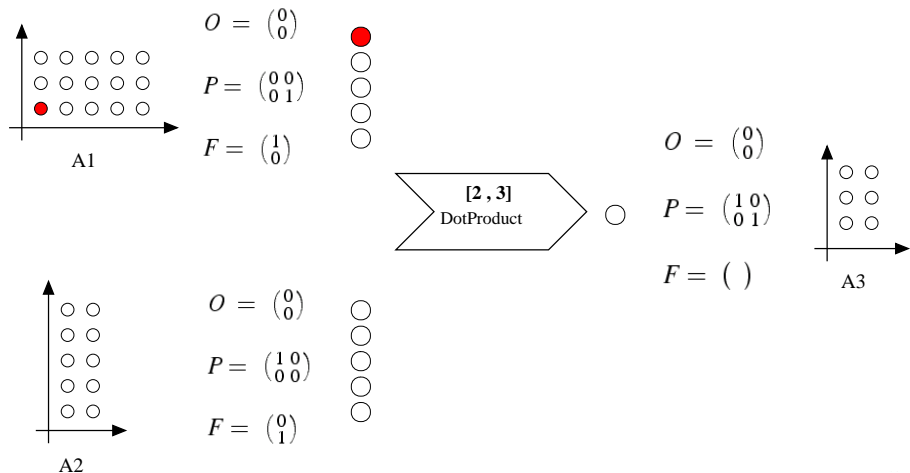


6	8	10
0	2	4

Pattern

ARRAY-OL (cont'd)

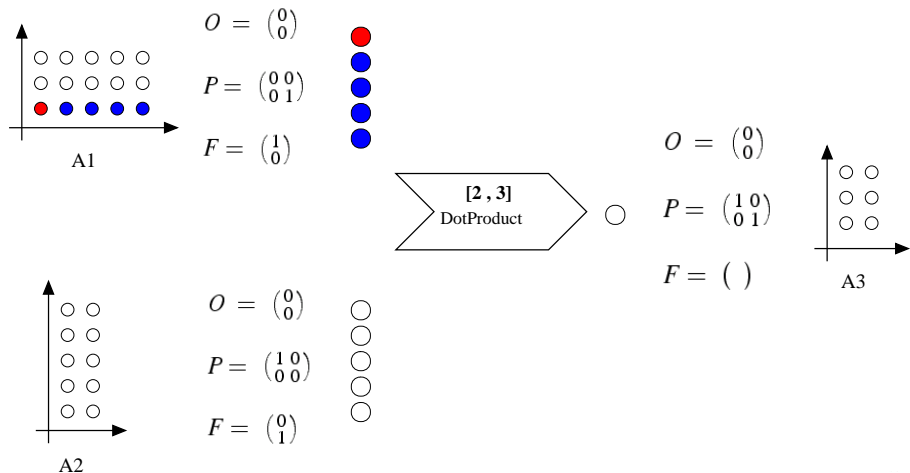
An example of repetition task : array product



Number of instances : $2*3=6$; Repetition point : $[0,0]$

ARRAY-OL (cont'd)

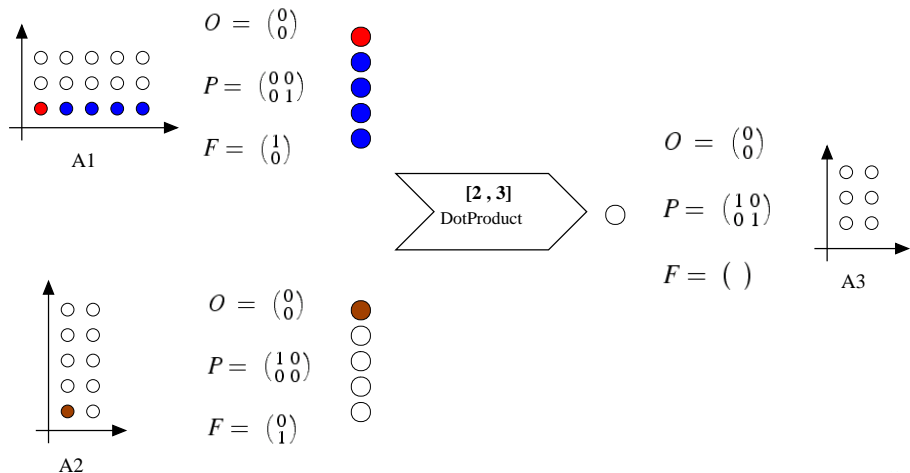
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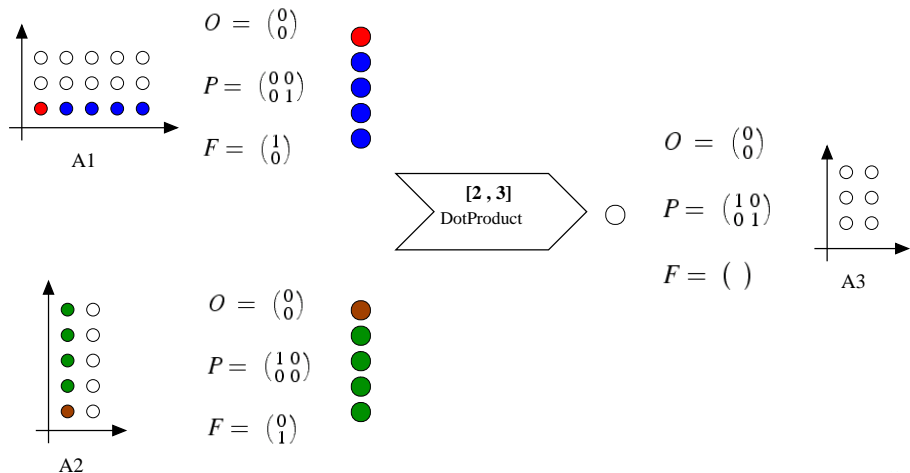
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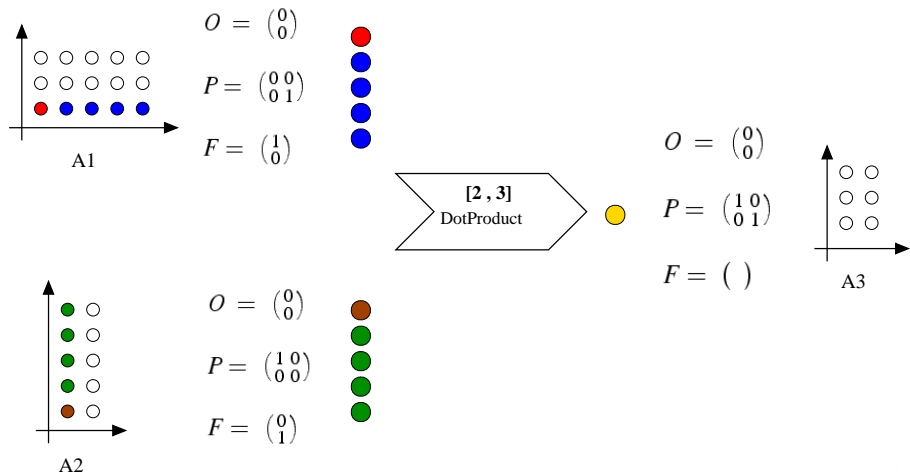
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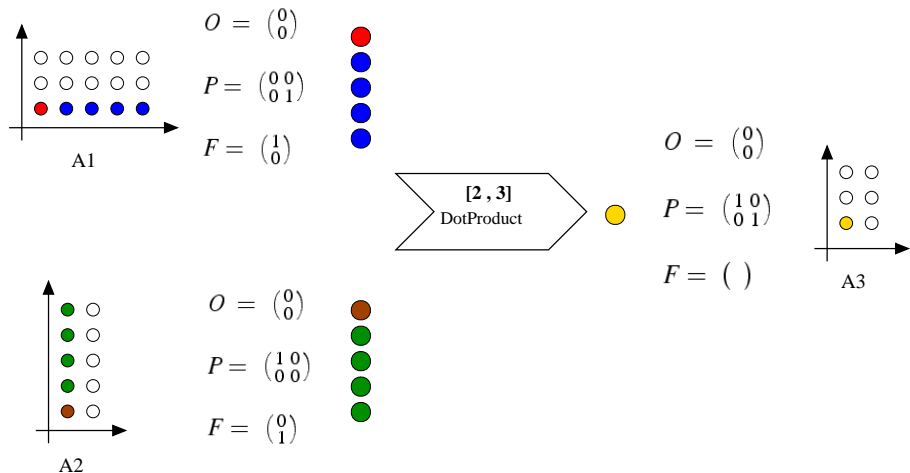
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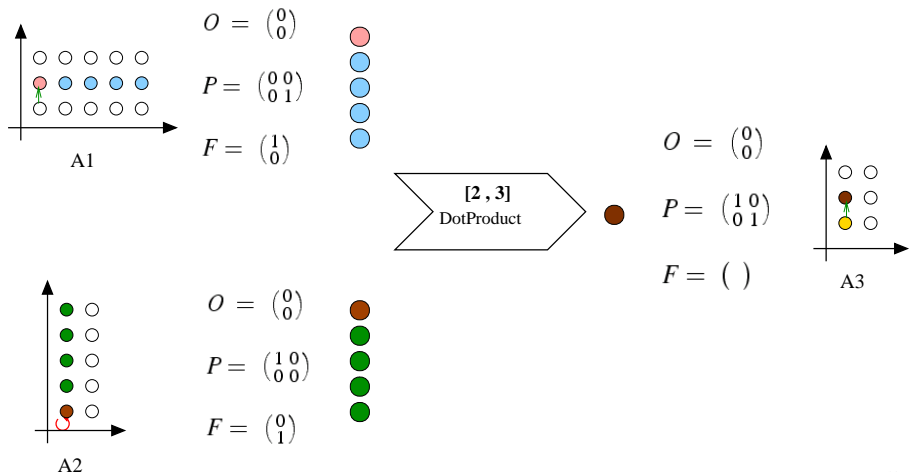
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Number of instances : $2*3=6$; Repetition point : $[0,0]$

ARRAY-OL (cont'd)

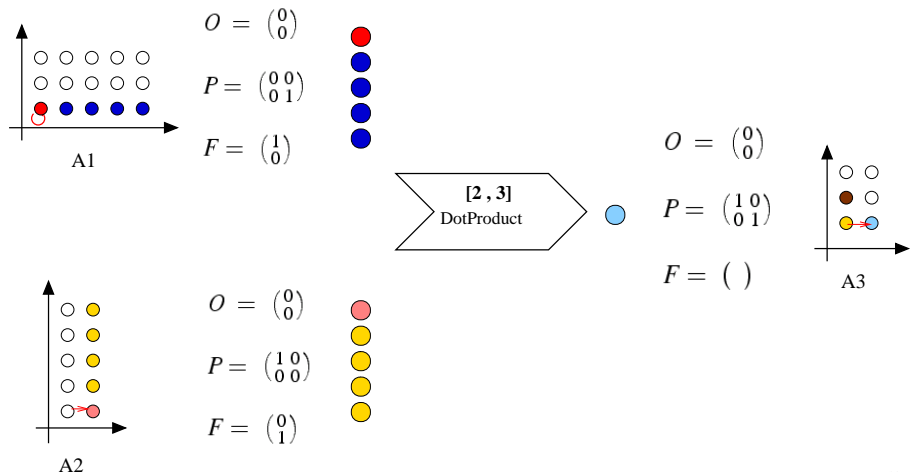
An example of repetition task : array product



Number of instances : $2*3=6$; Repetition point : $[0,1]$

ARRAY-OL (cont'd)

An example of repetition task : array product



Number of instances : $2*3=6$; Repetition point : $[1,0]$

Existing works

ALPHA language [Mauras, 1989] (vs. **ARRAY-OL**)

- ▶ multidimensional data structures for data-intensive applications
- ▶ union of convex polyhedra (vs. arrays)
- ▶ data access through indices calculated by affine functions (vs. hierarchical and modular pattern)
- ▶ absence of modulo (vs. presence of modulo)

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Simple synchronous modeling of GASPARD models

- Parallel model
- Serialized model

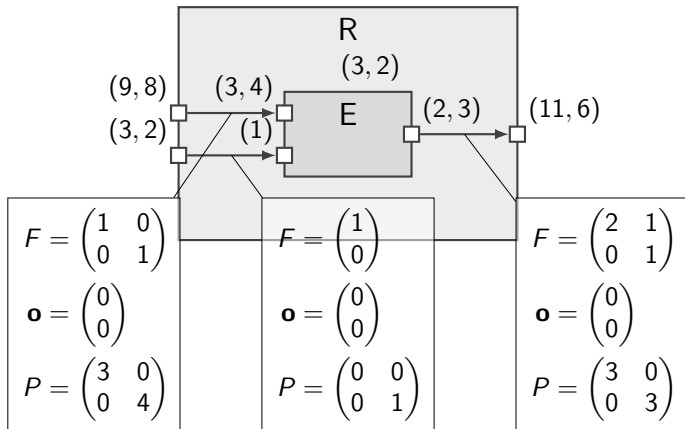
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Synchronous modeling of Gaspard models

Illustration of the modeling by the following example :



Modeling of repetition task

$$\forall j \in r, A_3[\langle ind_3^j \rangle] := E(A_1[\langle ind_1^j \rangle], A_2[\langle ind_2^j \rangle])$$

- ▶ j : a point in the repetition space r
- ▶ $\langle ind_i^j \rangle$: the set of index associated with pattern j
- ▶ $A_i[\langle ind_i^j \rangle]$: the pattern j associated with array A_i

Parallel model (cont'd)

Decomposition of a repetition

Input tilers : $p_1^j := A_1[\langle ind_1^j \rangle]$ $p_2^j := A_2[\langle ind_2^j \rangle]$

Task : $p_3^j := E(p_1^j, p_2^j)$

Output tiler : $A_3[\langle ind_3^j \rangle] := p_3^j$

Introduction of local variables : p_1^j, p_2^j, p_3^j

Parallel model (cont'd)

Decomposition of a repetition

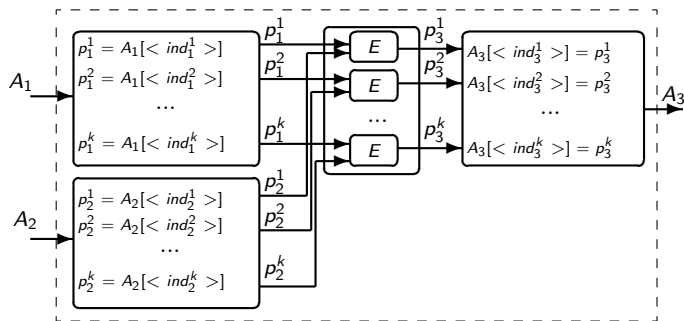
A complete system of equations :

$$\begin{array}{l} (| p_1^1 := A_1[< ind_1^1 >] | p_2^1 := A_2[< ind_2^1 >] \\ | p_3^1 := E(p_1^1, p_2^1) | A_3[< ind_3^1 >] := p_3^1 \\ | \dots \\ | p_1^k := A_1[< ind_1^k >] | p_2^k := A_2[< ind_2^k >] \\ | p_3^k := E(p_1^k, p_2^k) | A_3[< ind_3^k >] := p_3^k \\) \end{array} \quad (1)$$

where $p_1^1, p_2^1, p_3^1, \dots, p_1^k, p_2^k, p_3^k$; end ;

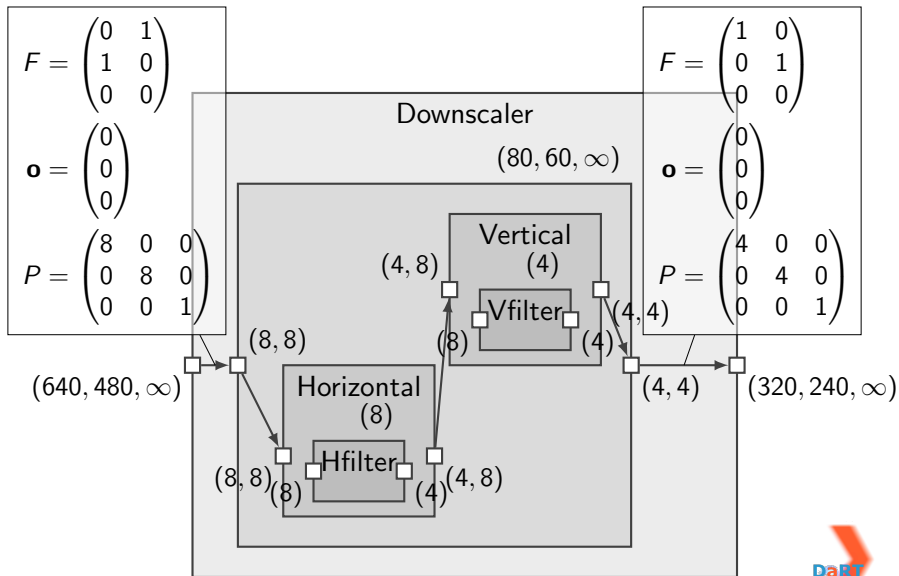
Parallel model (cont'd)

Restructuring and finalization of the model



- ▶ **Commutativity** and **associativity** of composition operator

Case study : video downscaling



Generated SIGNAL code

```
module Downscaler_module =
  process DOWNSCALER =
    (?type_array_i A_i;
     !type_array_o A_o;)
    (|(P_i1,...,P_iN:=
       HV_TILER_i(A_i)
      |(P_o1,...,P_oN):=
         R_HV_FILTER(P_i1,...
        |A_o:=HV_TILER_o(P_o1,... |)
  where
    type_pattern_i P_i1,...
    type_pattern_o P_o1,...
    process HV_TILER_i =
      (?type_array_i A_i;
       !type_pattern_i P_i1,...,
      (|P_i1:=HV_PATTERN_i1(A_i)
       |...|)
    where
      process HV_PATTERN_i1 =
```

```
    ...
    end%HV_TILER_i% ;
  process R_HV_FILTER =
    (?type_pattern_i P_i1,...,
     !type_pattern_o P_o1,...,
    (|P_o1:=HV_FILTER(P_i1)
     |...|)
  where
    process HV_FILTER =
      (? type_pattern_i P_i;
       ! type_pattern_o P_o;)
      (| p:= H_FILTER (P_i)
       | P_o := V_FILTER(p)
       |)
    where
      type_pattern_l p;
      process H_FILTER = ...
      process V_FILTER = ...
    end%HV_FILTER%;
```

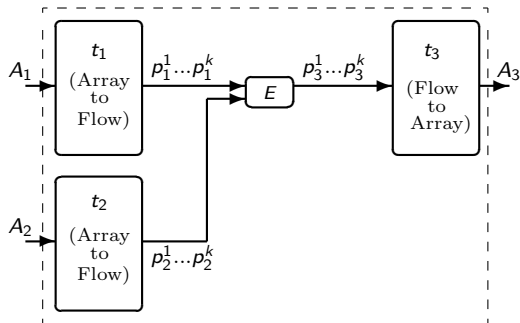


Serialized model

- ▶ Simple parallel model
 - ▶ Semantically equivalent
 - ▶ naively enumeration
- ▶ Association of application with architecture :
from repetition to iteration, introduction of flows
- ▶ Sequentialization at different granularity degrees [Labrani 2006]

Serialized model

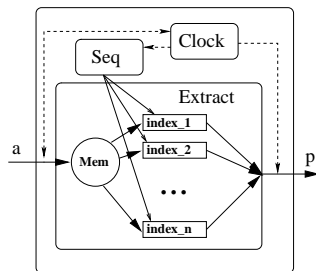
From repetition to iteration : introduction of flows



- ▶ **Array to flow** : produces pattern flows from arrays
- ▶ **Flow to array** : produces arrays from pattern flows

Serialized model

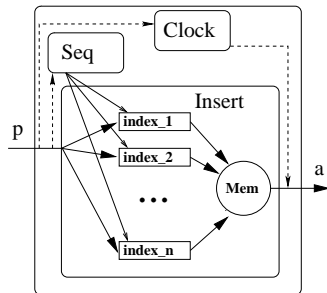
Array to flow



Main components : clock oversampling, sequencer and Extraction

Serialized model

Flow to array



Main components : clock undersampling, sequencer and insertion

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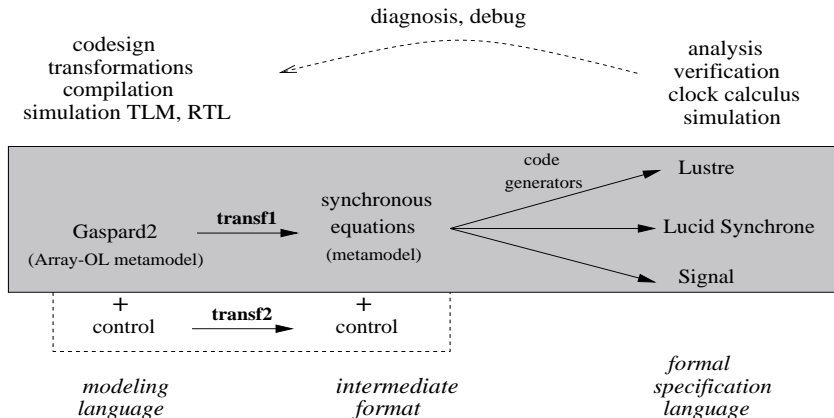
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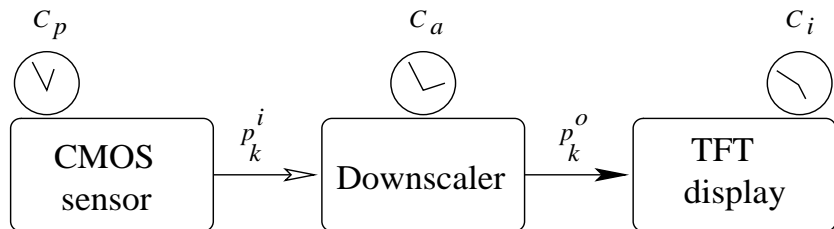
- ▶ We have a synchronous model with parallel and serialized version that can be combined (mixed model)
- ▶ We want to use synchronous analysis tools to address design correctness issues
ex. N-synchronous Kahn network [Cohen et al. 2006], clock calculus, model-checking
- ▶ Example : a simple application with affine clocks synchronizability analysis [Smarandache et al. 1999]

General scheme



Synchronizability analysis

Camera functionality in a cell phone



Synchronizability analysis

Clock constraints :



1. c_a is an affine undersampling of c_p : $c_p \xrightarrow{(1, \phi_1, d_1)} c_a$;
2. c_i is an affine undersampling of c_a : $c_a \xrightarrow{(1, \phi_2, d_2)} c_i$;

Synchronizability analysis

Clock constraints :



1. c_a is an affine undersampling of c_p : $c_p \xrightarrow{(1, \phi_1, d_1)} c_a$;
2. c_i is an affine undersampling of c_a : $c_a \xrightarrow{(1, \phi_2, d_2)} c_i$;

Now, let us consider a given external constraint, which imposes a particular image production rate c'_i , from c_p such that :

$c_p \xrightarrow{(1, \phi_3, d_3)} c'_i$. What about the **synchronizability** of c'_i and c_i ?

Synchronizability analysis

Clock constraints :



1. c_a is an affine undersampling of c_p : $c_p \xrightarrow{(1, \phi_1, d_1)} c_a$;
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$$c'_i \text{ and } c_i \text{ are synchronizable} \Leftrightarrow \begin{cases} \phi_1 + d_1 \phi_2 = \phi_3 \\ d_1 d_2 = d_3 \end{cases} \quad (2)$$

Conclusions and perspectives

Current results :

- ▶ Synchronous modeling of Gaspard specifications
- ▶ Analysis of GASPARD applications with the help of synchronous techniques
- ▶ Implementation of modeling approach following MDE

In the future :

- ▶ Complete implementation and validation of current results
- ▶ Extension with control features : mode-automata
- ▶ Using mixed models (parallel/serialized) combined with task fusion technique for placements in time and space

