

# Scheduling Parametric Data Flow Graphs

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SYNCHRON 2012

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## Goal:

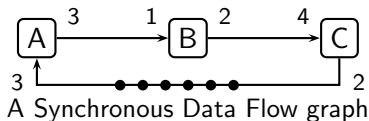
### **Scheduling parametric data flow applications on many core platforms**

- **Background:** Data flow models, from synchronous to parametric
- **Many-Core platform:** Platform 2012
- **Scheduling:** Scheduling framework for parametric data flow
- **Perspective:** Possible future work and exploration

- 1 Data Flow Models
  - Synchronous Data Flow
  - Parametric Data Flow
- 2 Platform 2012
- 3 Scheduling
- 4 Future Work

# Synchronous Data Flow

- Synchronous Data Flow <sup>1</sup> (SDF): Each port has a fixed rate known at compile time.



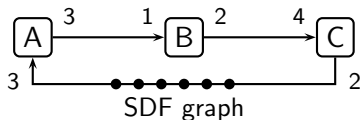
- **Actors:** Function units
- **Edges:** Communication links (FIFO)
- **Port rate:** Number of tokens transferred through a port
- **Graph State:** Number of tokens on the graph's edges

$$S_i = [0 \ 0 \ 6]$$

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<sup>1</sup>Synchronous Data Flow, E.A.Lee et al. 1987

# Synchronous Data Flow



$$\#A \cdot 3 = \#B$$

- **Balance equations:**  $\#B \cdot 2 = \#C \cdot 4$

$$\#C \cdot 2 = \#A \cdot 3$$

- **Repetition vector:**  $[2 \ 6 \ 3]$
- **Iteration:** Sequence of firings that return the graph to the initial state
- **Schedule:** Execution of a complete iteration  
e.g. Single Appearance Schedule:  $A^2, B^6, C^3$
- **Liveness:** enough initial tokens to fire actors on a directed cycle

# Advantages & Disadvantages of SDF

## Advantages

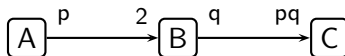
- + Natural expression of DSP applications
- + Finite memory - Boundedness guarantee
- + Deadlock-free operation - Liveness guarantee
- + Static scheduling - Timing guarantee

## Disadvantages

- Not expressive enough  
(e.g. for video codec applications)

# Parametric Data Flow

- Parametric Data Flow (PDF) uses parameterized instead of fixed port rates.



A parametric data flow graph

- Simplified version of SPDF<sup>2</sup> and PSDF<sup>3</sup> models
- No parameter changes within iterations
- No hierarchical structures
- Symbolic analysis of the graph

Repetition vector:  $\begin{bmatrix} 2 & p & 1 \end{bmatrix}$

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<sup>2</sup>Schedulable Parametric Data Flow, P.Fradet et al. 2012

<sup>3</sup>Parametric Dataflow Modelling for DSP Systems, B.Bhattacharya et al. 2001

# Advantages & Disadvantages of PDF

## Advantages

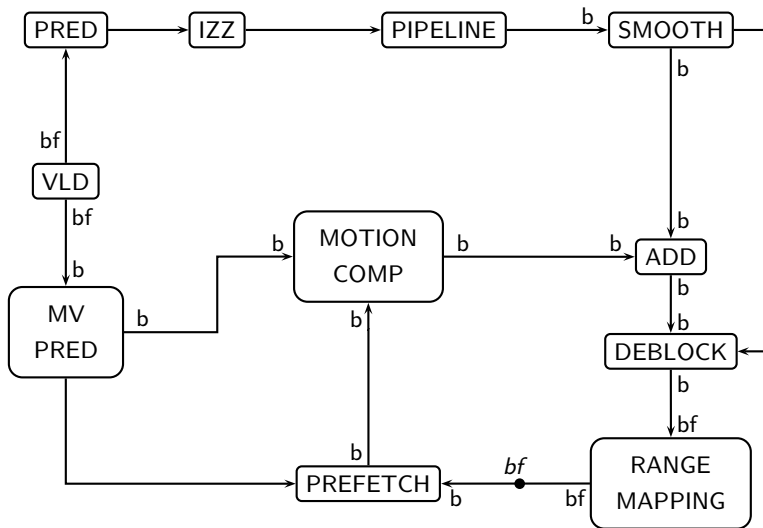
- + Finite memory - Boundedness guarantee
- + Deadlock-free operation - Liveness guarantee
- + Expression of video applications

## Disadvantages

- Static scheduling possible but too restrictive  
(e.g. As soon as possible schedule cannot always be expressed)



# Example: VC-1 decoder capture in PDF



# Outline

- 1 Data Flow Models
- 2 Platform 2012**
- 3 Scheduling
- 4 Future Work

## Platform Features

- Many - core platform designed by [STMicroelectronics](#)
- 1-32 clusters with 1-16 cores:
  - Software cores: General Purpose Processors (GPP)
  - Hardware cores: Hardware processing elements (HWPE)
- Native programming model

## Mapping assumptions

- Application fits in a **single cluster**
- Each actor is executed by a **GPP** or implemented as a **HWPE**
- The schedule is executed by a **GPP**

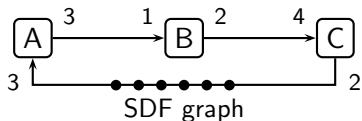
# Native programming model

- Predicated Execution Data Flow (PEDF) model
- Simplifies the parallel implementation of applications
- Data flow implementation of applications:
  - **Filter**: Basic functional block
  - **Controller**: Schedules the firing of filters and controls the configuration parameters
- Uses a slot notion for scheduling like in blocked scheduling <sup>4</sup>
  - + All filters synchronize after each cycle of iterations
  - + Reduces computational complexity of parallel scheduling
  - + Compatible with other many-core platforms (CUDA, OpenGL)
  - Introduces slack

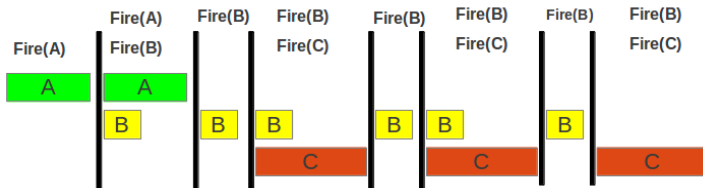
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<sup>4</sup>Compile-time Scheduling and Assignment of Data-flow Program Graphs with Data-Dependent Iteration, S.Ha et al. 1991

# Scheduling example



Repetition vector:  $[2 \ 6 \ 3]$



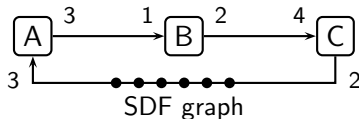
- 1 Data Flow Models
- 2 Platform 2012
- 3 **Scheduling**
  - Scheduling SDF graphs
  - Scheduling PDF graphs
  - Scheduling framework
- 4 Future Work

# Scheduling SDF graphs

## Scheduling goal

Schedule all firings to complete an iteration

- Repetition vector:  $[2 \ 6 \ 3]$



- Scheduling examples:

- **Sequential schedules:**

- $A^2, B^6, C^3$

- Single appearance schedule

- $A, B^2, C, B, A, B, C, B^2, C$

- Minimum buffer size schedule

- **Parallel schedules:**

- $A, (A \parallel B), [B, (B \parallel C)]^2, B, C$

- As Soon As Possible schedule (ASAP)

# Scheduling PDF graphs



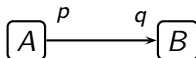
PDF graph

- Repetition vector:  $[2 \ p \ 1]$
- Single appearance schedule:  $A^2, B^p, C$
- Quasi-static schedule uses parameters to express a schedule at compile time
- Unable to produce ASAP schedule (*next slide*)
- No scheduling flexibility



# ASAP scheduling of PDF graphs

- Considering the generic PDF edge:



- Minimum firings of actor  $A$  for ASAP firing of actor  $B$ :

①  $n_1 = \left\lceil \frac{q}{p} \right\rceil$  and  $r_1 = n_1 \cdot p - q$

②  $n_2 = \left\lceil \frac{q-r_1}{p} \right\rceil$  and  $r_2 = n_2 \cdot p - q$

③  $n_3 = \left\lceil \frac{q-r_2}{p} \right\rceil$  and  $r_3 = n_3 \cdot p - q$

④  $\dots$

- As the number of remaining tokens change constantly a **quasi-static** schedule is not possible.

## Features

- **As Soon As Possible** schedule
  - Minimizes schedule span (no resource sharing)
  - Maximizes parallelism
- **Flexible** to be reused
  - Different platforms
  - Optimization criteria
  - Scheduling strategies
- **Main idea:** Produce different schedules while keeping the same algorithm
- Usage of scheduling constraints

# Scheduling framework overview

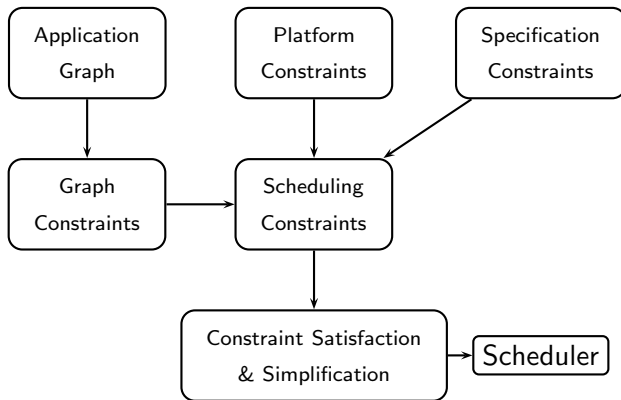


Figure: Scheduling framework

# Scheduling constraints

A constraint is a relationship between the firings of two actors  $(X, Y)$ :

$$X_i > Y_{f(i)}$$

## Interpretation

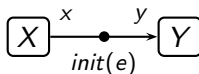
- The  $i_{th}$  firing of  $X$  waits for the  $f(i)_{th}$  firing of  $Y$   
 $\Rightarrow X_i$  must be scheduled at a later slot than  $Y_{f(i)}$

## Constraint types

- **Graph constraints:** Data dependencies
- **Platform constraints:** Constraints due to platforms specificities
- **User constraints:** Constraints to optimize some criterion

## Graph Constraints

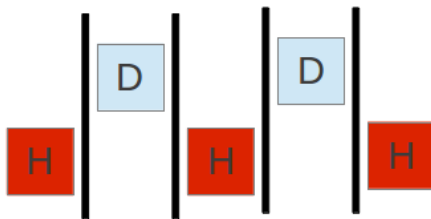
- Data dependencies derive from the graph.
- For each edge of the graph we get:



- $Y_i > X_{f(i)}$  with  $f(i) = \left\lceil \frac{y \cdot i - init(e)}{x} \right\rceil$

## Platform Constraints

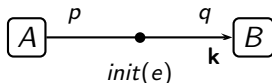
- Platform constraints are given and derive from platform specificities
- Applied to all application of the platform



- Example:  
**Prevention of power intensive filter  $H$  from firing twice in a row**  
 $\Rightarrow$  Introduction of dummy filter  $D$
- $H_i > D_{i-1}$  and  $D_i > H_i$

## User Constraints

- User constraints are used to achieve specific schedule behaviour
  - To express an optimization
  - To set a scheduling strategy



- Example:

**Buffer capacity restriction to k tokens**

$$\Rightarrow A_i > B_{g(i)} \quad \text{with} \quad g(i) = \left\lceil \frac{p \cdot i + \text{init}(e) - k}{q} \right\rceil$$

- **May introduce deadlocks!!!**  $\Rightarrow$  Constraint satisfaction algorithm to ensure compatibility

## Deadlock

A set of constraints **deadlocks** when there is a constraint that propagates from an actor's firing to the **same or future** firing of the same actor.

$$\exists A, i, j, (A_i > A_j) \wedge (i \leq j)$$

- Thus for each cycle:

$$A_i > B_{f_1(i)} > \dots > C_{f_{n-1}(i)} > A_{f_n(i)}$$

$$\Rightarrow A_i > A_{f_1(\dots(f_{n-1}(f_n(i))))}$$

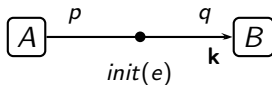
- Check if

$$i > f_1(\dots(f_{n-1}(f_n(i))))$$



# Deadlock detection example

- Considering the constraints between actors A and B:

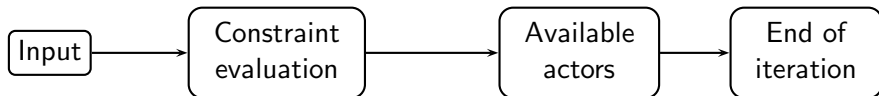


- Graph constraint:  $B_i > A_{f(i)}$  with  $f(i) = \left\lceil \frac{q \cdot i - \text{init}(e)}{p} \right\rceil$
- User constraint:  $A_i > B_{g(i)}$  with  $g(i) = \left\lceil \frac{p \cdot i + \text{init}(e) - k}{q} \right\rceil$
- Circular constraint:  $A_i > A_{f(g(i))}$
- Deadlock check:  $i > f(g(i)) \Rightarrow i > \left\lceil \frac{q \cdot \left\lceil \frac{p \cdot i + \text{init}(e) - k}{q} \right\rceil - \text{init}(e)}{p} \right\rceil$

# Run-time scheduler

## Input

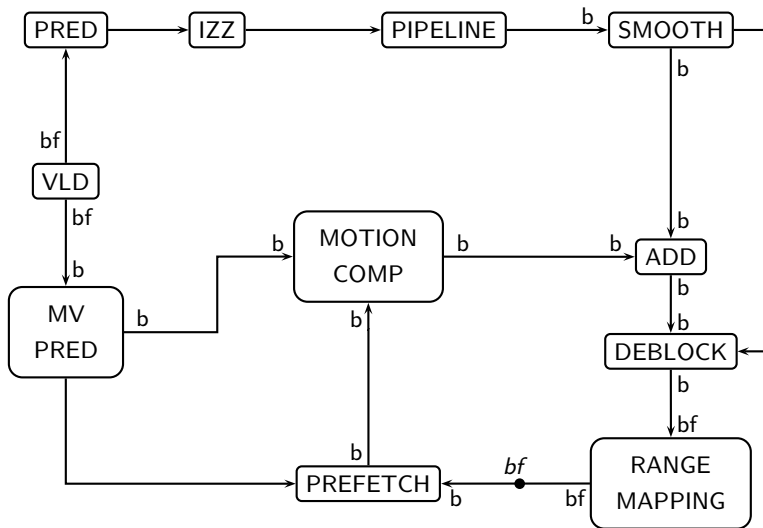
- Set of actors
- Repetition vector
- List of constraints



## Overhead

- Overall small overhead:
  - Concurrent execution with graph actors
  - Small amount of constraints
  - Optimization of static parts of the graph

# Example use case: VC-1 decoder



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## Conclusions

- **Flexible constraint framework** for PDF graphs
- **Modular way to change the schedule** without changing the scheduling algorithm
- **Ability to express** platform specificities and scheduling strategies
- **Compile time guarantees** of schedule liveness

## Future work

- **Scheduler optimization**: Solve all static and quasi-static actors at compile-time, run-time scheduling only when necessary
- **Extension of the PDF model** to include boolean parameters
- **Extension of the constraints** to express more interesting scheduling strategies
- **Implementation and integration** of the framework within **ST**'s tool-chain for the platform