Scheduling Parametric Data Flow Graphs

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Introduction



Scheduling parametric data flow applications on many core platforms

- Data flow
 - Data flow models
 - Data flow scheduling
- Scheduling framework
 - Scheduling model STHORM platform (ex. P2012)
 - Scheduling framework

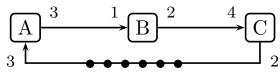
Outline



- Data Flow Models
 - Synchronous Data Flow
 - Parametric Data Flow
- Scheduling

SDF - Synchronous Data Flow





A Synchronous Data Flow graph

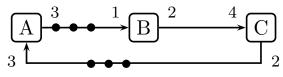
- Actors (A, B, C) & edges (AB, BC, CA)
 Function units & Communication links (FIFOs)
- Port rates:
 Number of tokens transferred through a port
- Graph State: $S_i = \begin{bmatrix} 0 & 0 & 6 \end{bmatrix}$ Number of tokens on the graph's edges

SDF1: Port rates are fixed and known at compile time



SDF - Synchronous Data Flow





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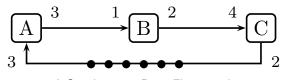
SDF1: Port rates are fixed and known at compile time



¹Lee and Messerschmitt 1987

SDF - Analysis



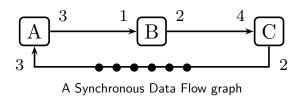


A Synchronous Data Flow graph

- Actor solutions (#A, #B, #C) & repetition vector (r): #A = 2, #B = 6, #C = 3 \Rightarrow r = $\begin{bmatrix} 2 & 6 & 3 \end{bmatrix}$
- Iteration: Sequence of firings that return the graph to S_i
- Schedule: Execution of a complete iteration e.g. A^2 ; B^6 ; C^3
- Liveness: Enough initial tokens on each directed cycle

SDF - Scheduling





- Repetition vector: [2 6 3]
- Sequential schedules:
 - A^2 ; B^6 ; C^3
 - $A; B^2; C; B; A; B; C; B^2; C$

- Single appearance schedule
- Minimum buffer size schedule

- Parallel schedules:
 - $A; (A||B); [B; (B||C)]^2; B; C$
- As Soon As Possible schedule (ASAP)

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SDF - Advantages & Disadvantages



Advantages

- + Modular and reusable design, suitable for DSP
- + Parallelism Exposure
- Boundedness and liveness guaranteed at compile time
- + Static scheduling Timing guaranteed

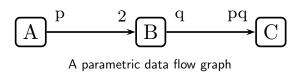
Disadvantages

 Too restrictive to express more advanced applications (e.g. video codec applications)



PDF - Parametric Data Flow





- Simplified version of SPDF² and PSDF³ models
- Parameters change between iterations
- Symbolic analysis of the graph Repetition vector: [2 p 1]

PDF: Parametric port rates that change between iterations

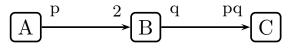


²P.Fradet et al. 2012

³B.Bhattacharya and S.Bhattacharyya 2001

PDF - Scheduling





A parametric data flow graph

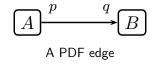
- Repetition vector: [2 p 1]
- Sequential schedules:
 - ► A²; B^p; C

- Single appearance schedule

- Parallel schedules:
 - Difficult to express ASAP schedule

PDF - ASAP scheduling





- Case $p \ge q$:
 - A; $(A|B)^{q-1}$; B^{p-q+1}
- Case q > p:
 - ▶ Subcase q = kp:
 - A; $(A^{k-1}; (A|B))^{p-1}; A^{k-1}; B$
 - ▶ Subcase q = kp + r, 0 < r < p:
 - Needs iterative comparison of the values of p and q



PDF - Advantages & Disadvantages



Advantages

- + Modular and reusable design, suitable for streaming applications
- + Parallelism Exposure
- + Boundedness and liveness guaranteed at compile time
- + Increased expressiveness

Disadvantages

 (Quasi -) static parallel scheduling possible is more involved (e.g. ASAP scheduling is a challenge)



Outline



- Data Flow Models
- Scheduling
 - STHORM platform
 - Scheduling framework
 - Future Work

STHORM platform



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)

Mapping assumptions

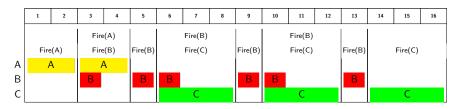
- Application fits in a single cluster
- Each actor is executed by a GPP or implemented as a HWPE
- The schedule is executed by a GPP

Slotted scheduling model



- Compatible with the scheduling model of STHORM
- Uses a slot notion like in blocked scheduling ⁴
 - + Actors synchronize after each execution
 - + Reduces complexity of parallel scheduling
 - + Compatible with other parallel programming models (CUDA, OpenGL)
 - May introduce slack

Repetition vector: [2 6 3]



⁴S.Ha et al. 1991



Scheduling framework features



The framework should

- Automatically produce ASAP schedules
- Be expressive and flexible for different
 - Platforms
 - Optimization criteria
 - Scheduling strategies

Main idea: Production of different schedules with the same (ASAP) algorithm



Scheduling framework overview



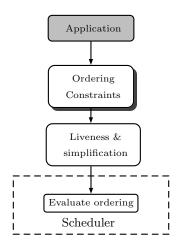


Figure : Scheduling framework



Scheduling framework overview



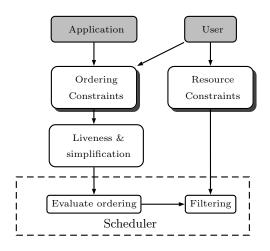


Figure: Scheduling framework



Scheduling constraints



Ordering Constraints: Express the partial ordering of the firings

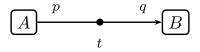
$$X_i > Y_{f(i)}$$

• Resource Constraints: Control the parallel execution

replace
$$S_A$$
 by S_B if condition
where $S_B \subseteq S_A$ and $S_B \neq \emptyset$

Constraint Examples





Graph Constraint: Data dependency

$$B_i > A_{f(i)}$$
 with $f(i) = \left\lceil \frac{q \cdot i - t}{p} \right\rceil$

User Contraint: Buffer capacity restriction to k

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

Resource Constraint: Mutual exclusion of A and B

replace
$$\{A, B\}$$
 by $\{A\}$



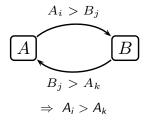
Constraint deadlock Detection



Deadlock

A set of ordering constraints deadlocks when it implies (by transitivity) a constraint of the form:

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



$$\forall$$
 cycle $A_i > A_k$ check if $i > k$

Deadlock detection example



Constraints:

$$B_i > A_{f(i)}$$

$$A_i > B_{g(i)}$$

Cycle:

$$A_i > A_{f(g(i))}$$

Deadlock free condition:

Solution:

$$i > f(g(i)) \Leftrightarrow i > \left| \frac{q \cdot \lceil \frac{p \cdot i - k}{q} \rceil}{p} \right|$$

$$\Leftrightarrow i > \frac{q \cdot (\frac{p \cdot i - k}{q} + 1)}{p} + 1$$

$$\Leftrightarrow i > i + \frac{q - k}{p} + 1$$

$$\Leftrightarrow k > p + q$$

$$\Leftrightarrow k > p_{max} + q_{max}$$

Constraint simplification





Constraints:

$$A_i > B_{\left\lceil \frac{i}{2} \right\rceil}$$
 $C_i > B_{\left\lceil \frac{2pi}{2} \right\rceil}$
 $C_i > A_{pi}$

Firing Function:

$$B_i = i$$
 , $i \in [1 \cdots p]$

$$A_i = max(B_{\left\lceil rac{i}{2} \right\rceil}, A_{i-1}) + 1$$
 , $i \in [1 \cdots 2p]$

$$\Rightarrow A_i = 2i + 1$$

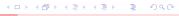
$$\begin{aligned} &C_i = max \big(B_{pi}, A_{pi}, C_{i-1}\big) + 1 \quad, \quad i = 1 \\ \Rightarrow &C_i = max \big(A_{pi}, C_{i-1}\big) + 1 \\ \Rightarrow &C_i = 2p + 1 \end{aligned}$$

Schedules:

$$B: \mathcal{F}^{p}$$

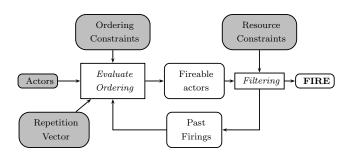
$$A: \mathcal{E}; (\mathcal{E}; \mathcal{F})^{2p}$$

$$C: \mathcal{E}^{2p}: \mathcal{F}$$



Run-time scheduler





Overall small overhead:

- Concurrent execution with actors
- Coarse grain graph
- Optimization of static parts of the graph



Conclusions



We presented a **scheduling framework for PDF** applications that:

- Flexible constraint framework for PDF graphs:
- Modular way to adjust the schedule
- Expressive power to optimize the schedule
- Automatically generates of ASAP schedules
- Statically guarantees boundness and liveness of the schedule

On-going work

- Implementation and integration within ST's SDK
- Evaluation of the scheduler with real world applications

Future work

- Introduction of timing information
- Flexible slotted scheduling model already used on the platform

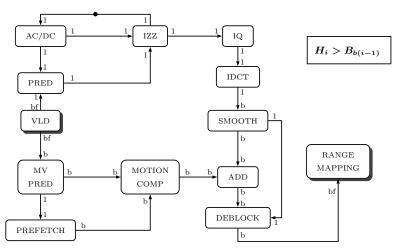
Questions



Thank you for your attention!

Use case example: VC1 decoder





VC-1 capture in PDF



Resource constraint examples



Mutual Exclusion (A and B: specific actors)

replace
$$A$$
, B by A

Bounded Parallelism (x,y,z: variables - can be any actor)

replace
$$x, y, z$$
 by x, y

Timing optimization

replace
$$x$$
, y **by** x **if** $short(x) \land long(y)$

Power optimization

replace
$$x$$
, y **by** x **if** $high(x) \land high(y)$

replace
$$x, y, z$$
 by x, y if $high(x) \land low(y) \land low(z)$