Parametric Dataflow Programming: Model of Computation and Many-Core Scheduling

Alain Girault (alain.girault@inria.fr)



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Joint work with Vagelis Bebelis & Pascal Fradet (INRIA)

Presentation Overview



- Data Flow Models of Computation
- Scheduling Framework
- 3 Experiments
- 4 Conclusions

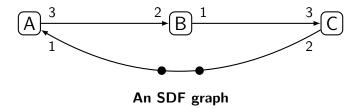
Outline



- Data Flow Models of Computation
 - Synchronous Data Flow
 - Boolean Parametric Data Flow
- Scheduling Framework
- 3 Experiments
- 4 Conclusions

SDF - Synchronous Data Flow ¹

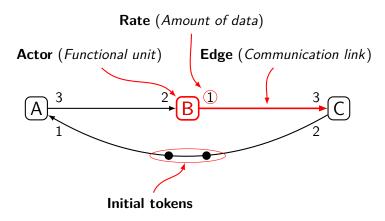




¹E.A.Lee and D.G.Messerschmitt, Proc. of the IEEE, 1987

SDF - Synchronous Data Flow ¹





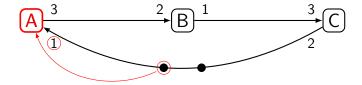
In SDF all rates are fixed and known at compile time

¹E.A.Lee and D.G.Messerschmitt, Proc. of the IEEE, 1987

SDF - Firing



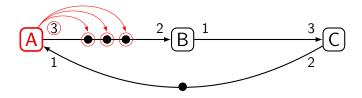
Firing of actor A: Consumes 1 token



SDF - Firing

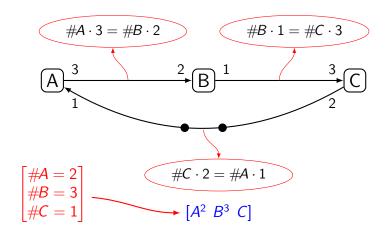


Firing of actor A: Produces 3 tokens



SDF - Balance Equations

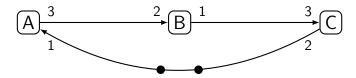




SDF - Scheduling



Schedule: Series of firings that complete one iteration Here: $[A^2 B^3 C]$



Sequential Single Appearance Sequential Minimum Buffer Size Parallel ASAP

$$A^2$$
; B^3 ; C
 A ; B ; A ; B^2 ; C
 A ; $(A||B)$; B^2 ; C

Parametric data flow models

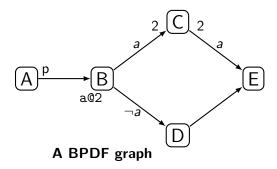


Need for more expressive data flow models

- SDF is not expressive enough for complex applications.
- More expressive models are employed that use
 - Parametric rates
 - Dynamic graph topology
- Both features make scheduling more difficult.
- Such a model is Boolean Parametric Data Flow (BPDF)¹.

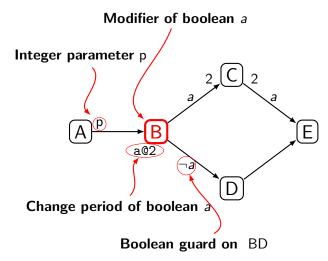
Boolean Parametric Data Flow - BPDF





Boolean Parametric Data Flow - BPDF

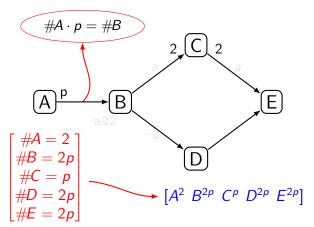




BPDF - Balance Equations



BPDF analysis: Balance Equations

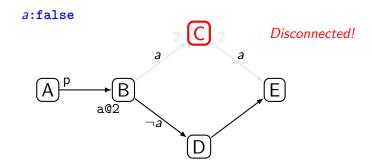


Parametric solution of balance equations

BPDF - Balance Equations



BPDF analysis: Balance Equations

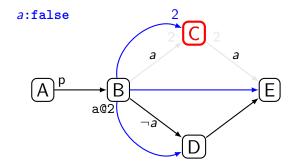


Actor C fires despite being disconnected

BPDF - Balance Equations

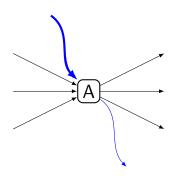


BPDF analysis: Balance Equations



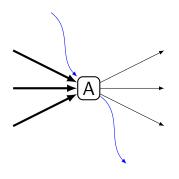
There are implicit boolean propagation links





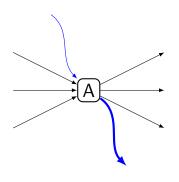
- (1) Read boolean parameters
- (2) Read data from connected inputs
- (3) Set boolean parameters
- **(4)** ... Compute ...
- (5) Write data to connected outputs





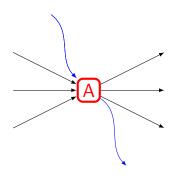
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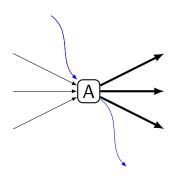
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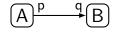




- (1) Read boolean parameters
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BPDF Scheduling - Integer parameters





Sequential Single Appearance Sequential Minimum Buffer Size Parallel ASAP A^q; B^p
Difficult to express

Difficult to express

Needs iterative comparison of the values of p and q

BPDF Scheduling - Boolean parameters



Parallel ASAP schedule when g = true A^p ; BParallel ASAP schedule when g = false A; (A||B); A^{p-2}

Also needs constant checking of the boolean values

Outline



- Data Flow Models of Computation
- Scheduling Framework
 - STHORM platform
 - Scheduling framework
- 3 Experiments
- 4 Conclusions

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)

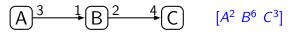
Mapping assumptions

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

Slotted scheduling model



- Compatible with the scheduling model of STHORM.
- Uses a slot notion like in blocked scheduling ²
 - + Actors synchronize after each execution
 - + Reduces complexity of parallel scheduling
 - + Compatible with other parallel programming models (CUDA, OpenGL)
 - May introduce slack



	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
			Fire(A)				Fire(B)				Fire(B)					
	Fire	e(A)	Fire	e(B)	Fire(B)		Fire(C)		Fire(B)		Fire(C)		Fire(B)		Fire(C)	
Α	A	4	F	4												
В			В		В	В			В	В			В			
C							С				С				С	

²S.Ha et al., IEEE Trans. On Computers, 1991

Scheduling framework features



The framework should

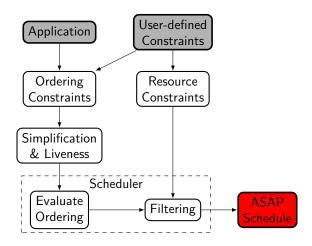
- Automatically produce ASAP schedules
 - Best strategy when timing is unknown ³
- Be expressive and flexible for different
 - Platforms
 - Optimization criteria
 - Scheduling strategies

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

 $^{^3}$ S.Sriram and S.S. Btathacharyya. Embedded Multiprocessors: Scheduling and Synchronization. 2000

Scheduling framework overview





Scheduling constraints



Ordering Constraints: Express the partial ordering of the firings

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

where X_i is the *i*th firing of actor X

• Resource Constraints: Control the parallel execution

replace S_A by S_B if condition

where S_A and S_B are subsets of actors such that $S_A \supseteq S_B \neq \emptyset$ (used for filtering the preliminary schedule)

Application Constraints





Graph constraint: Data dependency

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

because
$$t + f(i) \cdot p \ge q \cdot i \Longleftrightarrow f(i) \ge \frac{q \cdot i - t}{p}$$

Modifier to user constraint: Boolean dependency

$$U_i > M_{f(i)}$$
 with $f(i) = \pi_w \cdot \left\lfloor \frac{i-1}{\pi_r} \right\rfloor + 1$

where π_r (resp. π_w) is the reading (resp. writing) period of U (resp. M)

User Constraint Examples



$$A \xrightarrow{p} \xrightarrow{q} B$$

Buffer Contraint: Buffer capacity restriction to k

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

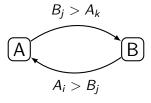
Resource Constraint: Mutual exclusion of A and B

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

Constraint liveness condition



User ordering constraints may introduce deadlocks



$$\Rightarrow A_i > A_k$$

Liveness condition:

$$\forall$$
 cycle $A_i > A_k$
we need $i > k$

Deadlock detection example



Constraints:

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

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

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

Liveness condition:

Solution:

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

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

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

$$\Leftrightarrow k > p + q$$

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

Constraint simplification



$$[A B^{2p} C^{3p}]$$

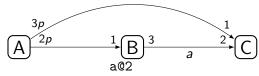
Constraints

$$B_{i} > A_{\left\lceil \frac{i}{2p} \right\rceil}$$

$$C_{i} > B_{\left\lceil \frac{2i}{3} \right\rceil}$$

$$C_{i} > A_{\left\lceil \frac{i}{3p} \right\rceil}$$

$$C_{i} > B_{2\left\lceil \frac{i}{3} \right\rceil - 1}$$



$$\begin{aligned} & \frac{A_1 = 1}{B_i = \max(A_1, B_{i-1}) + 1} & \text{ for } i \in [1..2p] \\ & B_i = i + 1 \\ & \overline{C_i = \max(A_1, B_{\left\lceil \frac{2i}{3} \right\rceil}, B_{2\left\lceil \frac{i}{3} \right\rceil - 1}, C_{i-1}) + 1} & \text{ for } i \in [1..3p] \\ & C_i = i + 2 \end{aligned}$$

ASAP Schedule: $A; B; (B||C)^{2p-1}; C^{p+1}$

Run-time scheduler



If simplification is not possible then a run-time scheduler is employed

Small overhead:

- Concurrent execution with actors
- Coarse grain graphs (small number of actors)
- Simplification of constraints at compile time
- Optimization of static parts of the graph

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Scheduler Overhead



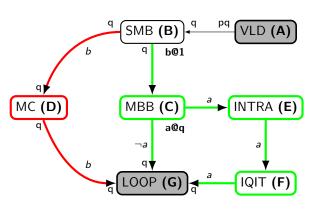
- Implementation of TNR (Temporal Noise Reduction) application on STHORM platform
- Overhead comparison between
 - Dynamic scheduler
 - Simplified scheduler
 - Manually optimized schedule

	Best Actor Performance	Dynam. sched.	Simpl. sched.	Manual Sched.
Cycles / frame	2.140.000	1.100.000	360.000	340.000

Schedule overhead for different schedules of TNR

Use Case: VC-1 decoder



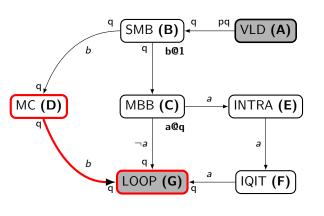


- Inter pipeline
- Intra pipeline

Repetition vector: $[A B^p C^{pq} D^p E^{pq} F^{pq} G^p]$

Buffer restriction





Token accumulation on edge DG

- Maximum buffer estimation: $\sim pq-p$
- Buffer restriction to: q

$$D_i > G_{\left\lceil \frac{q \cdot i - q}{q} \right\rceil}$$

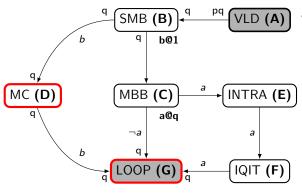
Increase of 2% in total schedule duration

Repetition vector:

$$[A B^p C^{pq} D^p E^{pq} F^{pq} G^p]$$

Timing Optimization





When timing is known, the goal is to minimize slack

 Clustering of actors D and G in the same slot with:

 $\mathbf{replace}\ D,\ E\ \mathbf{by}\ E\ \mathbf{if}\ \neg \mathtt{fireable}(G)$

 Improvement of 15% in the schedule duration.

Repetition vector:

 $[A B^p C^{pq} D^p E^{pq} F^{pq} G^p]$

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Conclusions



We proposed a scheduling framework for BPDF applications that

- is flexible and modular through constraints
- is expressive to optimize the schedule
- automatically generates ASAP parallel schedules
- Statically guarantees the boundness and liveness of the produced schedule

Ongoing and future work



- Use the framework with a non-slotted scheduling model
- Formalization of the constraint simplification procedure
- Use the framework to optimize bi-criteria scheduling, specifically power consumption vs. throughput
- Design a high-level language to express scheduling policies that can be automatically compiled into constraints
- Implementation of BPDF within Ptolemy II
- Formal comparison between BPDF and SADF ⁴

⁴Stuiik et al. IC-SAMOS'11

Thank you for your attention!



Questions?

Run-time scheduler



