

Symmetry Breaking for Multi-Criteria Mapping and Scheduling on Multicores

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August 2013

Context

- Typical in parallel programming: spawn multiple **identical tasks**
 - data parallelism
 - obtain hyperperiod of a multi-periodic system
 - duplicate tasks for fault-tolerance



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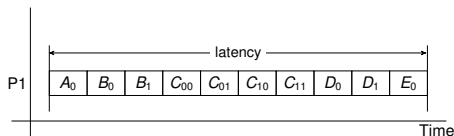
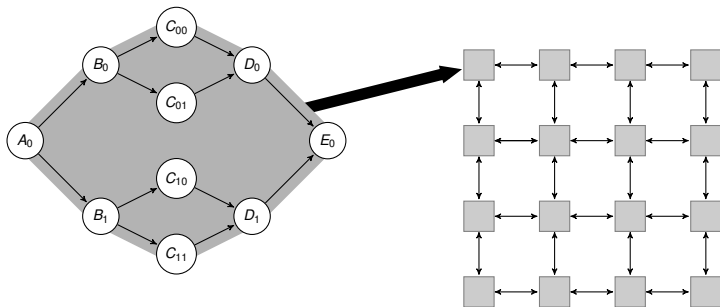
Context

- Typical in parallel programming: spawn multiple **identical tasks**
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 - obtain hyperperiod of a multi-periodic system
 - duplicate tasks for fault-tolerance
- Often the platform have multiple **identical processors**.
- Hence, **symmetry** in the solution space.



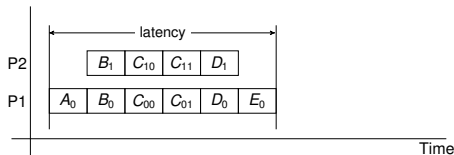
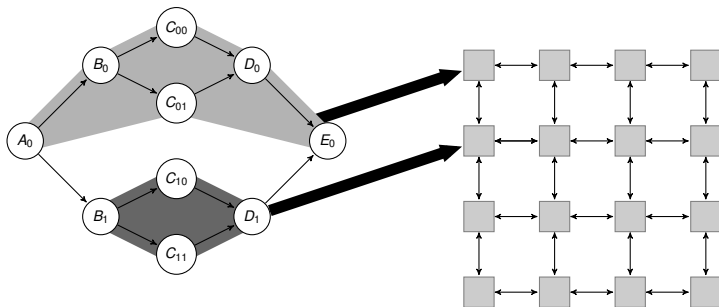
Multi-criteria Optimization

minimize **latency** using minimal number of **processors**



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Contribution

context:

static mapping and scheduling for programs with data-parallelism

multi-criteria optimization using **SMT solvers**



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symmetry breaking in solution space for identical tasks and processors

goal: increase the tractable problem size of SMT solvers

experiments : problem size increase from 20 to 50 tasks



Outline

- 1 Motivation
- 2 Application Model
- 3 Problem Formulation - SMT
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Model of Computation

synchronous dataflow graphs (SDF)

by E. Lee and D. Messerschmitt in 1987

task graph + symbolic representation of data parallelism

signal-processing, video-coding applications

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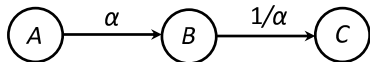
we introduce **split-join graphs** : restriction of SDF

still covering perhaps 90% of use cases



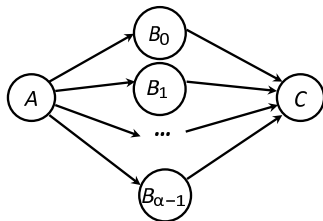
Split-Join Graphs

a simple split-join graph example:



α : spawn and split

$1/\alpha$: wait and join



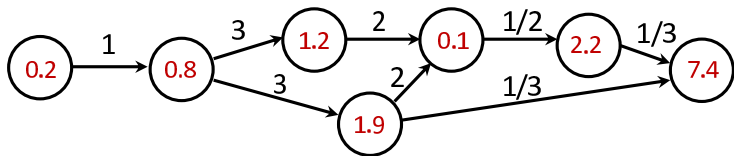
Split-Join Graphs

Definition (Split-Join Graph)

$S = (V, E, d, \alpha)$, $(V, E) : \text{DAG}$, $V : \text{actors}$, $E : \text{channels}$

$d : V \rightarrow \mathbb{R}_+$: actor execution time,

$\alpha : E \rightarrow \mathbb{Q}$: channel counter: split (> 1), join (< 1) or neutral ($= 1$)



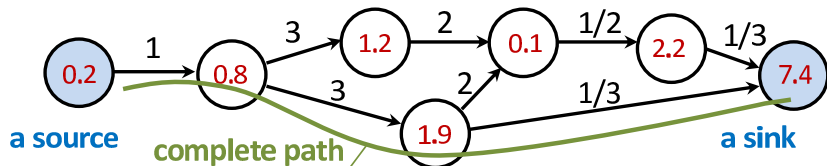
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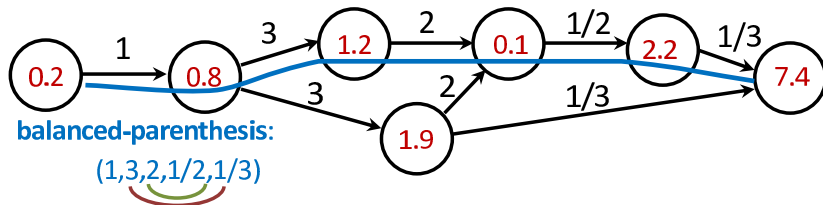
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Well-behaved Graphs

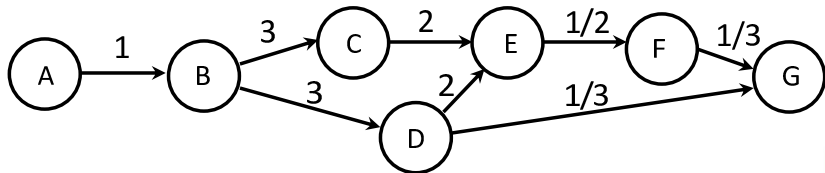
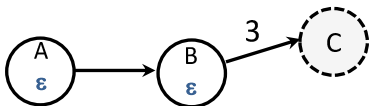
Definition (Well-behaved)

$S = (V, E, d, \alpha)$ is **well-behaved** if any complete path has balanced-parenthesis signature

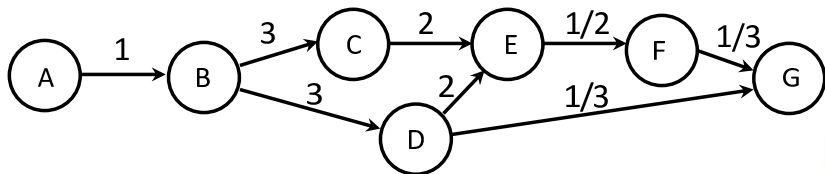
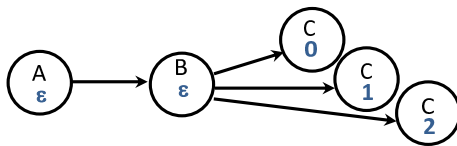
Such a graph can be unfolded to a task graph.



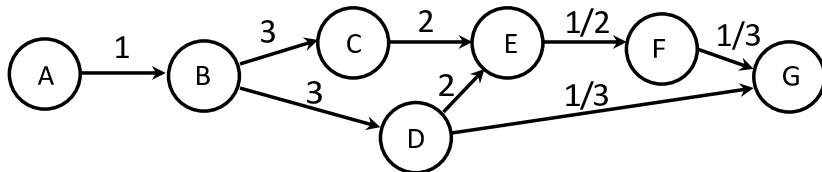
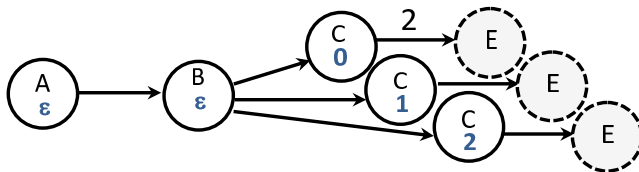
Unfolding to Task Graph



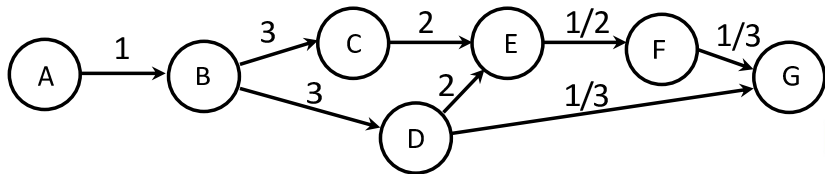
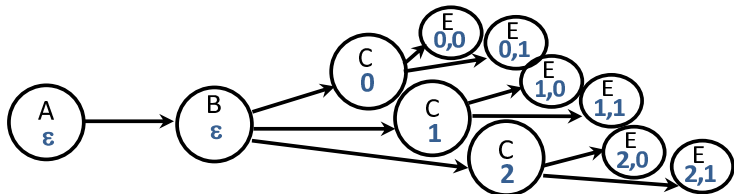
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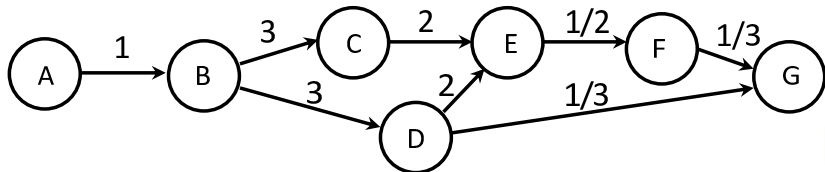
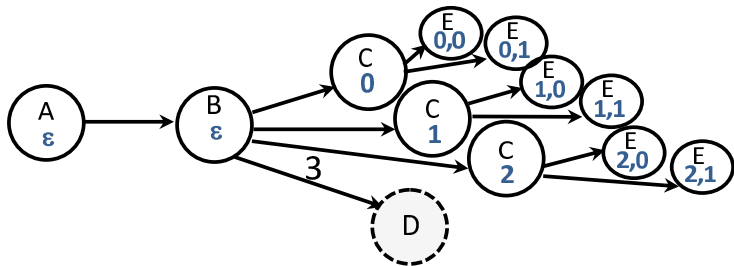
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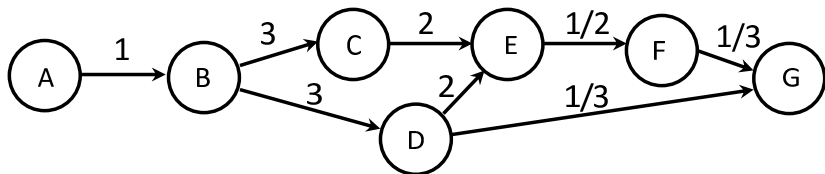
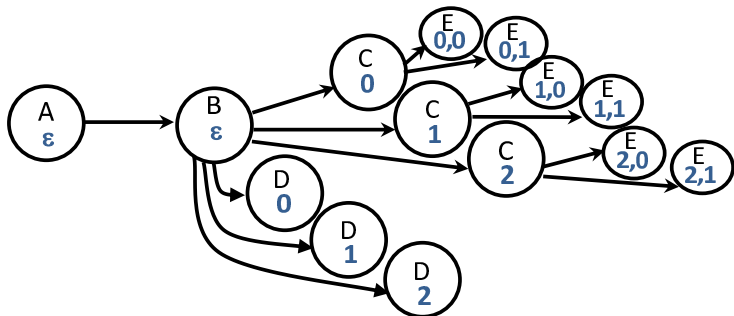
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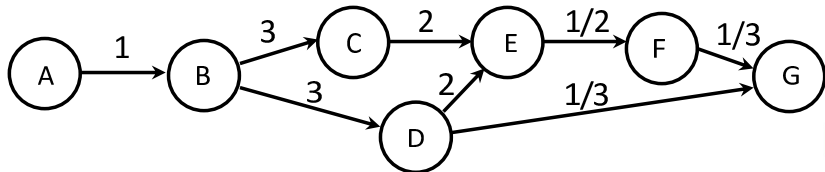
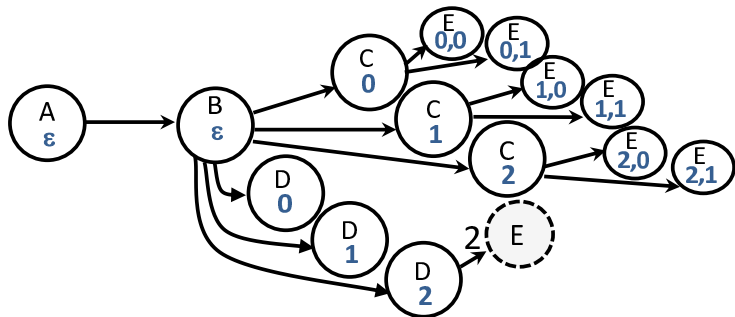
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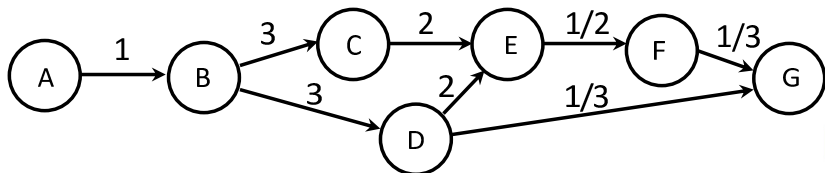
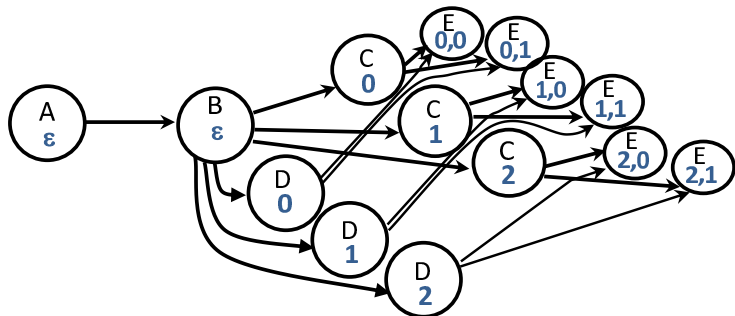
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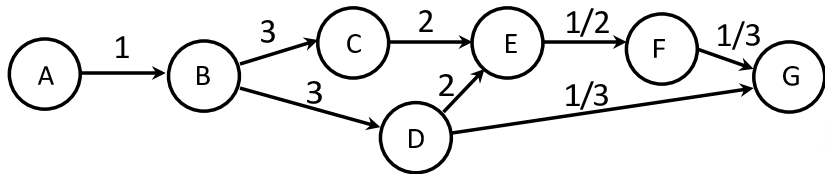
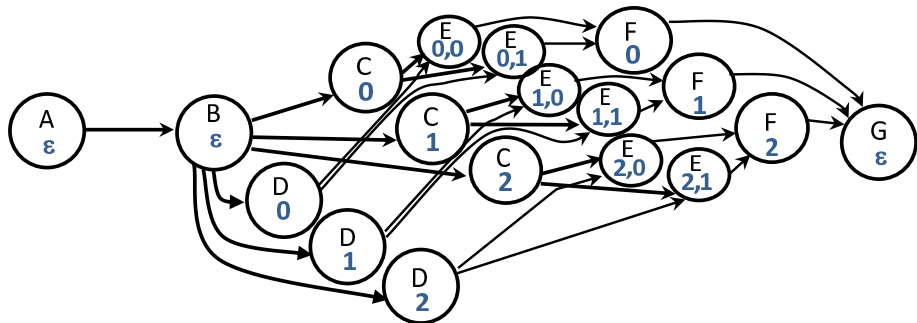
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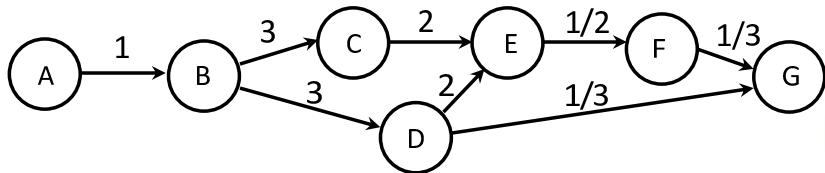
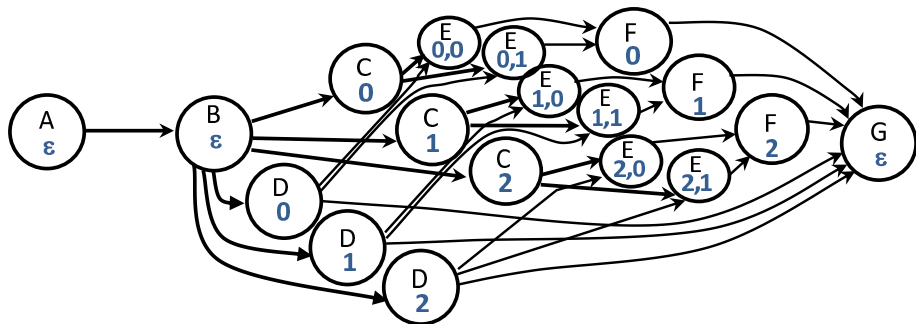
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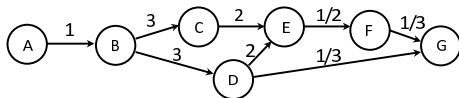


Unfolding to Task Graph



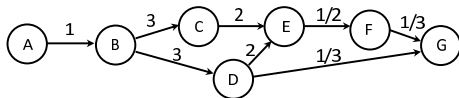
Actors, Tasks, Lexicographic Order

split-join graph: **actors** e.g., *A, B, C*



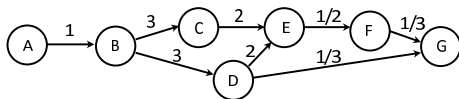
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notation for actors: $v, v \in V$

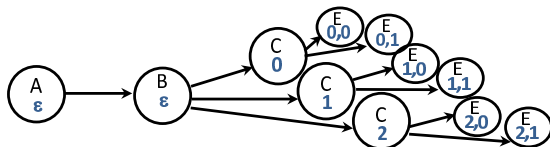


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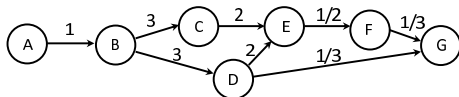


unfolded task graph: **tasks** e.g., $E_{0,1}$, B , C_2

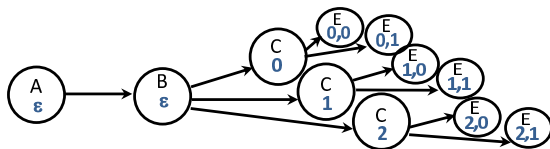


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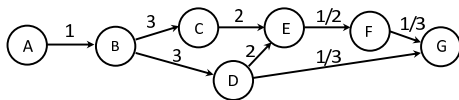


notation for tasks: $u \in U$



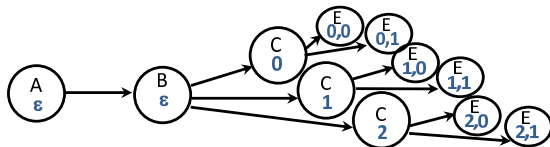
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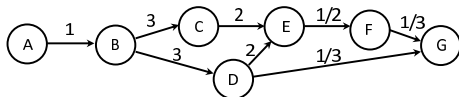
notation for tasks: $u \in U$

$u = v_h, v \in V$ and h - hier. index, e.g., $v_h = E_{0,1}$



Actors, Tasks, Lexicographic Order

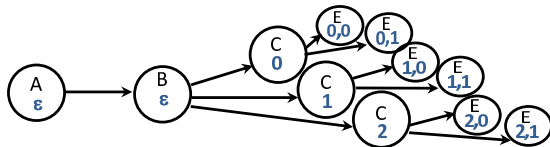
notation for actors: $v, v \in V$



notation for tasks: $u \in U$

$U_v = \{v_h\}$: **lexicographically ordered** (\ll) set of instances of v

$U_E : E_{0,0} \ll E_{0,1} \ll E_{1,0} \ll E_{1,1} \ll E_{2,0} \ll E_{2,1}$



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Decision variables:

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- $s(u)$ - the **schedule**: start time of u



Constraints

Predicate $\varphi(u, u')$:

task u' starts after the completion of task u

$$\varphi(u, u') : s(u') \geq s(u) + \delta(u)$$



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Mutual exclusion:

$$\bigwedge_{u \neq u' \in U} (\mu(u) = \mu(u')) \Rightarrow \varphi(u, u') \vee \varphi(u', u)$$

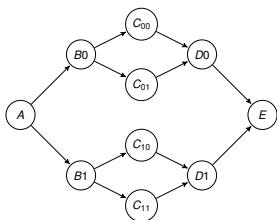


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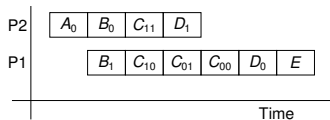
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Task Symmetry



task graph

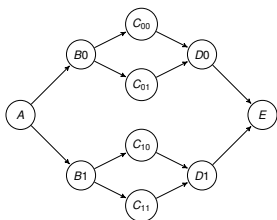


a schedule

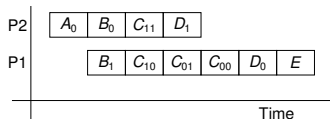
- all instances of given actor v are similar (symmetric)



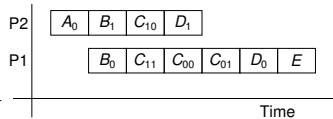
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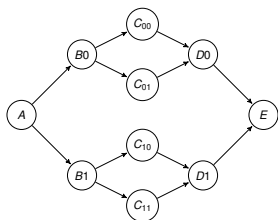


a permuted schedule

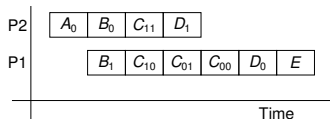
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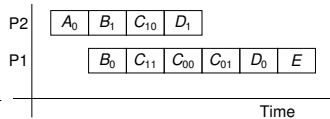
Task Symmetry



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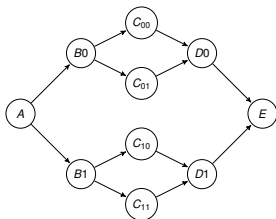


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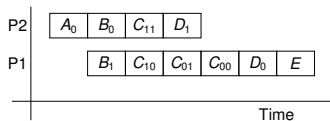
- all instances of given actor v are similar (symmetric)
- permutation of symmetric tasks does not change the latency,
- ... but extends the solution space exponentially



Task Symmetry



task graph

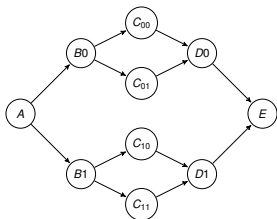


schedule

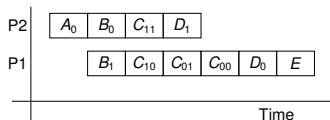
- enforce the schedule to be **compatible** with lexicographic order:
 $s(C_{00}) \leq s(C_{01}) \leq s(C_{10}) \leq s(C_{11})$



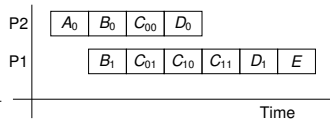
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task graph



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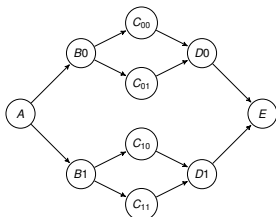


compatible schedule

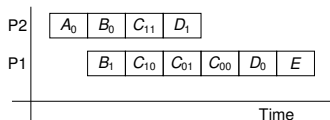
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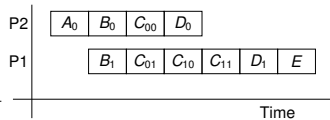
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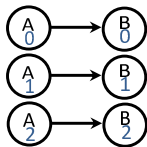


compatible schedule

- enforce the schedule to be **compatible** with lexicographic order:
 $s(C_{00}) \leq s(C_{01}) \leq s(C_{10}) \leq s(C_{11})$
- Theorem:** adding constraints $s(u) \leq s(u')$ for $u \ll u'$ does not eliminate optimality



Proof Sketch

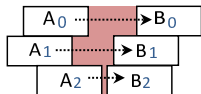


modify a feasible schedule such that:

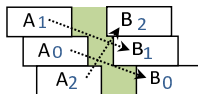
$$s(v_0) \leq s(v_1) \leq s(v_2) \leq \dots$$

prove that precedence constraints are satisfied

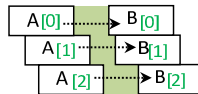
here: for **neutral** channels ($\alpha = 1$), unfolded to (v_h, v'_h)



↓
lexicographic
order



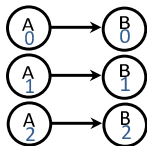
↓
start-time
compatible



↓
new hier. index;
new precedence relation



Proof Sketch

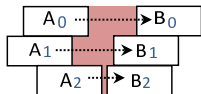


modify a feasible schedule such that:

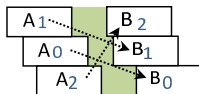
$$s(v_0) \leq s(v_1) \leq s(v_2) \leq \dots$$

prove that precedence constraints are satisfied

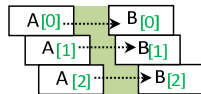
here: for **neutral** channels ($\alpha = 1$), unfolded to (v_h, v'_h)



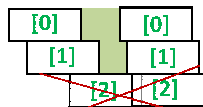
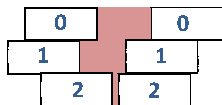
↓
lexicographic
order



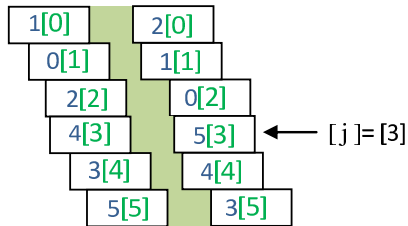
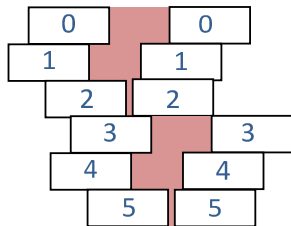
↓
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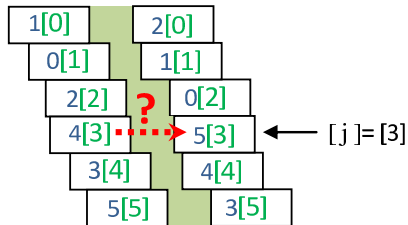
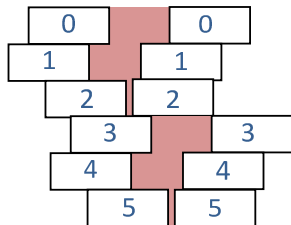
Proof Sketch



take successor $[j]$



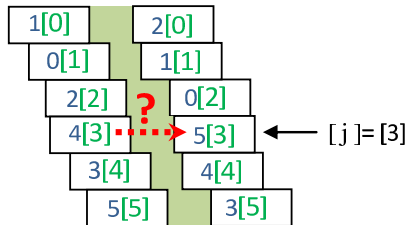
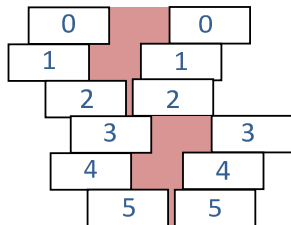
Proof Sketch



take successor $[j]$



Proof Sketch

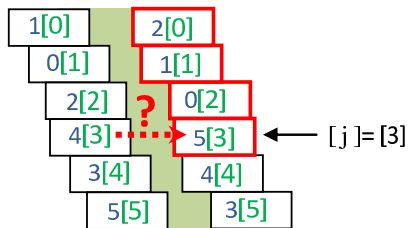
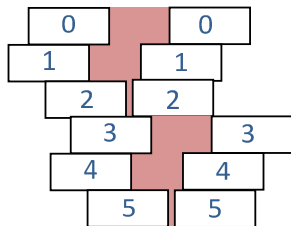


take successor $[j]$

by definition there exist $j + 1$ same or earlier successors



Proof Sketch

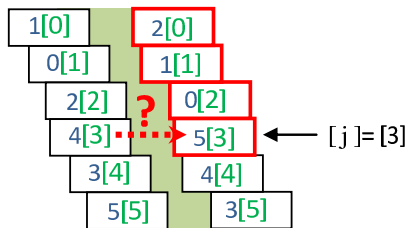
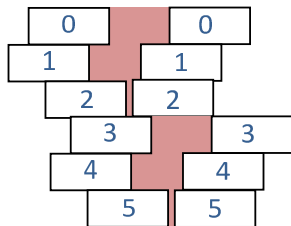


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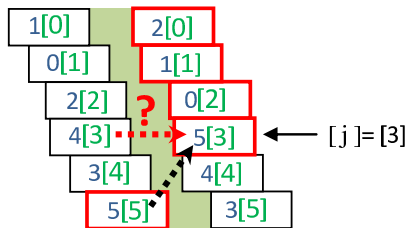
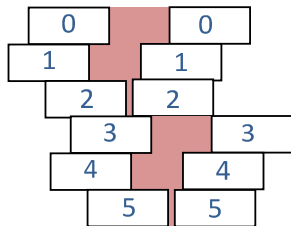


take successor $[j]$

by definition there exist $j + 1$ same or earlier successors
 their original predecessors finish before successor $[j]$:



Proof Sketch

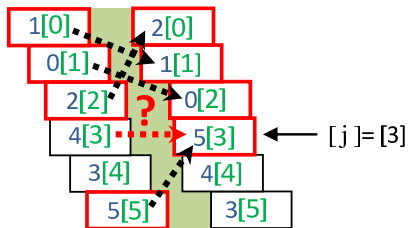
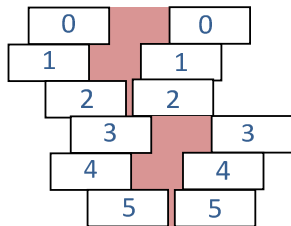


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Proof Sketch

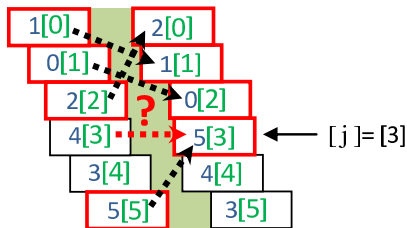
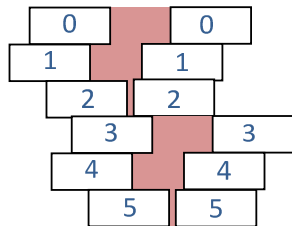


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Proof Sketch



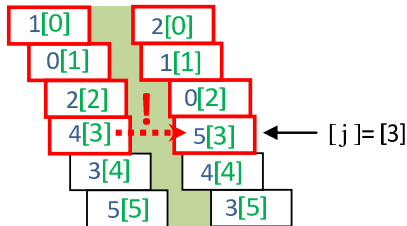
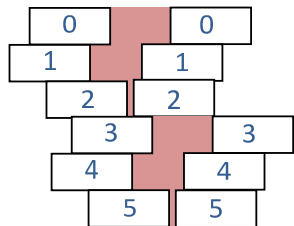
take successor $[j]$

by definition there exist $j + 1$ same or earlier successors
 their original predecessors finish before successor $[j]$:

$j + 1$ predecessors finish before, hence the earliest $j + 1$ ones as well



Proof Sketch



take successor $[j]$

by definition there exist $j + 1$ same or earlier successors

their original predecessors finish before successor $[j]$:

$j + 1$ predecessors finish before, hence the earliest $j + 1$ ones as well
 predecessor $[j]$ finishes before successor $[j]$



Outline

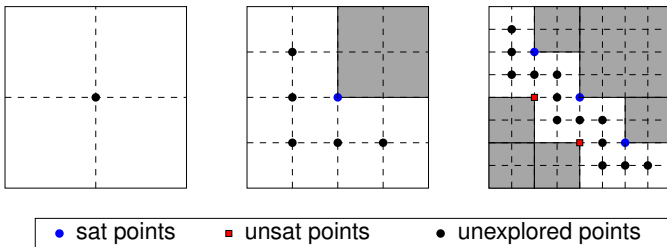
- 1 Motivation
- 2 Application Model
- 3 Problem Formulation - SMT
- 4 Symmetry Breaking
- 5 Cost Space Exploration**
- 6 Experiments and Results
- 7 Conclusions



Exploring the Design Space

One SMT query for a given point (C_L, C_M) in the cost space:

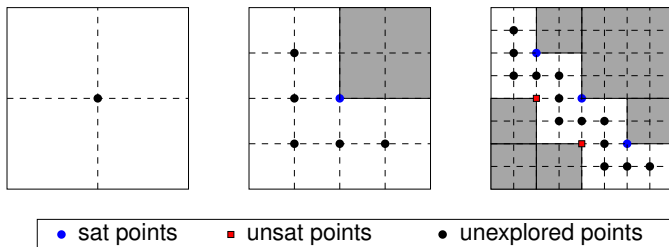
- C_L - latency
- C_M - processor count



Exploring the Design Space

One SMT query for a given point (C_L, C_M) in the cost space:

- C_L - latency
- C_M - processor count



- Precedence and Mutual Exclusion Constraints
- Cost Constraints

$$\bigwedge_{u \in U} s(u) + \delta(u) \leq C_L \wedge \bigwedge_{u \in U} \mu(u) \leq C_M$$

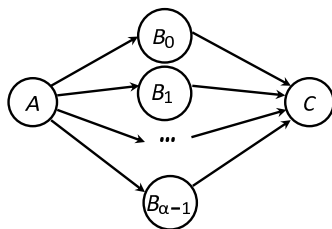
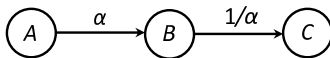


Outline

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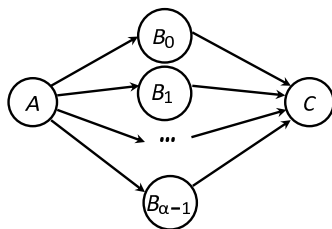
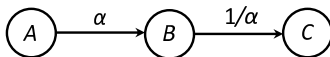
Synthetic-Graph Experiments



- Fix processor cost C_M and perform **binary search** for optimal C_L
- Increase α and measure increase in **computation time**
- With(out) breaking of **task symmetry** and **processor symmetry**



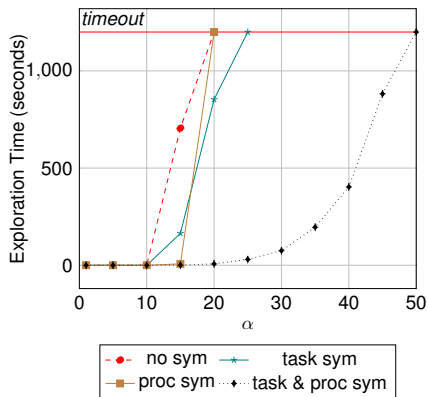
Synthetic-Graph Experiments



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- Z3 solver v4.1 on i7 core at 1.73GHz



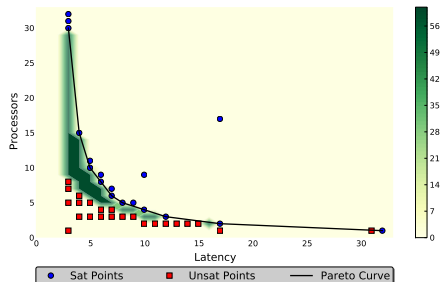
Synthetic-Graph Experiments



5-processor deployments



Pareto Exploration

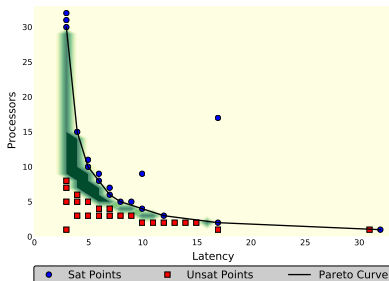


without symmetry breaking

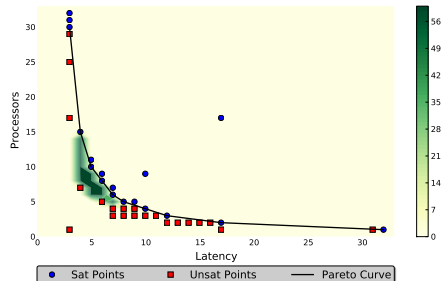
cost space (C_L, C_M) exploration for $\alpha = 30$
 evaluate task and processor symmetry breaking



Pareto Exploration



without symmetry breaking



with symmetry breaking

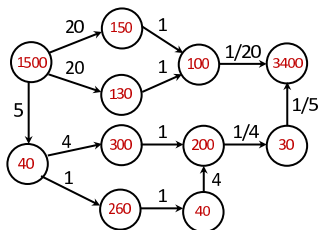
cost space (C_L, C_M) exploration for $\alpha = 30$
 evaluate **task and processor** symmetry breaking



Video Decoder

3D cost space (C_L , C_M , C_B) exploration, C_B - total buffer size

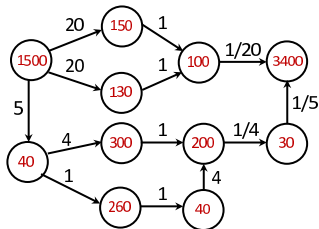
MPEG video decoder:



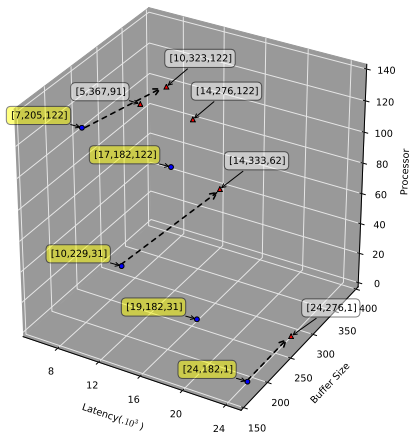
Video Decoder

3D cost space (C_L , C_M , C_B) exploration, C_B - total buffer size

MPEG video decoder:



▲ without symmetry constraints ● with symmetry constraints



Conclusions

- Symbolic representation of data-parallel programs
 - a useful subclass of SDF model
- Framework for **multi-criteria optimal deployment**
- Symmetry breaking: **prove task symmetry** and **use processor symmetry**



Conclusions

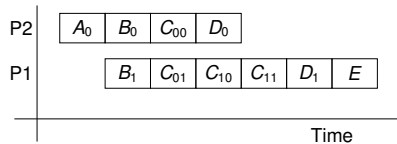
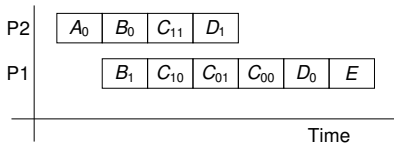
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- **Future work:**



Conclusions

- Symbolic representation of data-parallel programs
 - a useful subclass of SDF model
- Framework for **multi-criteria optimal deployment**
- Symmetry breaking: **prove task symmetry** and **use processor symmetry**
- **Future work:**
- More symmetry breaking, also approximation and heuristics
- More refined data communication: data transfer delays
- Pipelined scheduling
- Scheduling under uncertainty
- Multistage design flow





QUESTIONS?

