Multi-criteria scheduling of workflow applications

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Introduction and motivation

- Mapping applications onto parallel platforms
 Difficult challenge
- Heterogeneous clusters, fully heterogeneous platforms
 Even more difficult!
- Target platform
 - more or less heterogeneity
 - different communication models (overlap, one- vs multi-port)
- Target application
 - Workflow: several data sets are processed by a set of tasks
 - Structured: independent tasks, linear chains, ...
 - (Selectivity: some tasks filter data)



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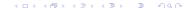
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Multi-criteria scheduling of workflow applications

Workflow:



Several consecutive data sets enter the application graph.

Criteria to optimize?

Period \mathcal{P} : time interval between the beginning of execution of two consecutive data sets (inverse of throughput)

Latency \mathcal{L} : maximal time elapsed between beginning and end of execution of a data set



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Major contributions

Definitions

Workflow applications Computational platforms and communication models Multi-criteria mappings

Theory

Problem complexity Linear programming formulation

Practice

Heuristics for sub-problems

Experiments: compare and evaluate heuristics

Simulation of real applications (JPEG encoder)



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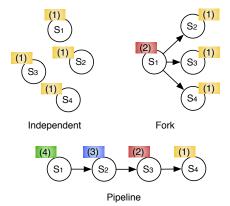


Outline

- 1 Definitions: Application, Platform and Mappings
- Working out examples
- Summary of complexity results
- 4 Conclusion

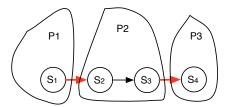
Application model

- Set of n application stages
- Workflow: each data set must be processed by all stages
- Computation cost of stage S_i : w_i
- Dependencies between stages

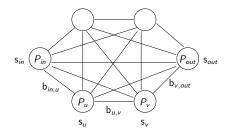


Application model: communication costs

- Two dependent stages $S_1 \rightarrow S_2$: data must be transferred from S_1 to S_2
- Fixed data size $\delta_{1,2}$, communication cost to pay only if S_1 and S_2 are mapped on different processors (i.e. red arrows in the example)



Platform model



- p processors P_u , $1 \le u \le p$, fully interconnected
- s_u : speed of processor P_u
- bidirectional link link_{u,v} : $P_u \rightarrow P_v$, bandwidth $b_{u,v}$
- fp_u : failure probability of processor P_u (independent of the duration of the application, meant to run for a long time)
- P_{in} : input data P_{out} : output data



Different platforms

Fully Homogeneous – Identical processors ($s_u = s$) and links ($b_{u,v} = b$): typical parallel machines

Communication Homogeneous – Different-speed processors $(s_u \neq s_v)$, identical links $(b_{u,v} = b)$: networks of workstations, clusters

Fully Heterogeneous – Fully heterogeneous architectures, $s_u \neq s_v$ and $b_{u,v} \neq b_{u',v'}$: hierarchical platforms, grids

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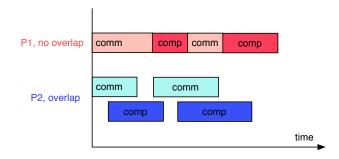
Failure Homogeneous – Identically reliable processors (fp_u = fp_v)
```

- Communication Homogeneous Different-speed processors $(s_u \neq s_v)$, identical links $(b_{u,v} = b)$: networks of workstations, clusters
- Fully Heterogeneous Fully heterogeneous architectures, $s_u \neq s_v$ and $b_{u,v} \neq b_{u',v'}$: hierarchical platforms, grids
- Failure Heterogeneous Different failure probabilities ($fp_u \neq fp_v$)

Platform model: communications

no overlap vs overlap

- no overlap: at each time step, either computation or communication
- overlap: a processor can simultaneously compute and communicate

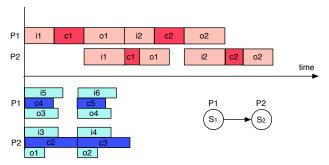




Platform model: communications

one-port vs multi-port

- one-port: each processor can either send or receive to/from a single other processor any time-step it is communicating
- bounded multi-port: simultaneous send and receive, but bound on the total outgoing/incoming communication (limitation of network card)



- Map each application stage onto one or more processors
- Goal: minimize period/latency and maximize reliability
- The pipeline case: several mapping strategies



- Other applications, one to one and general always defin
- Define connected-subgraph mapping (instead of interval)
- Replication: independent sets of processors, instead of a single processor as above

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Mapping: stage types

- Monolithic stages: must be mapped on one single processor since computation for a data set may depend on result of previous computation
- Replicable stages: can be replicated on several processors, but not parallel, i.e. a data set must be entirely processed on a single processor
- Data-parallel stages: inherently parallel stages, one data set can be computed in parallel by several processors
- Replication for reliability (also called duplication): one data set is processed several times on different processors.



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Mapping: objective function?

Mono-criterion

- Minimize period \mathcal{P} (inverse of throughput)
- ullet Minimize latency ${\cal L}$ (time to process a data set)
- ullet Minimize application failure probability \mathcal{FP}



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Multi-criteria

- How to define it? Minimize $\alpha.P + \beta.L + \gamma.FP$?
- Values which are not comparable

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Multi-criteria

- How to define it? Minimize $\alpha.P + \beta.L + \gamma.FP$?
- Values which are not comparable
- ullet Minimize ${\cal P}$ for a fixed latency and failure
- Minimize \mathcal{L} for a fixed period and failure
- Minimize \mathcal{FP} for a fixed period and latency



Mapping: objective function?

Mono-criterion

- Minimize period \mathcal{P} (inverse of throughput)
- ullet Minimize latency ${\cal L}$ (time to process a data set)
- ullet Minimize application failure probability \mathcal{FP}

Bi-criteria

- Period and Latency:
- Minimize \mathcal{P} for a fixed latency
- Minimize \mathcal{L} for a fixed period
- And so on...



An example of formal definitions

- Pipeline application, INTERVAL MAPPING
- Period/Latency problem with no replication
- Communication Homogeneous: one-port with no overlap

$$\mathcal{P} = \max_{1 \le j \le m} \left\{ \frac{\delta_{d_j - 1}}{\mathsf{b}} + \frac{\sum_{i = d_j}^{e_j} \mathsf{w}_i}{\mathsf{s}_{\mathsf{alloc}(j)}} + \frac{\delta_{e_j}}{\mathsf{b}} \right\}$$

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$$\mathcal{L} = \sum_{1 \le i \le n} \left\{ \frac{\delta_{d_j - 1}}{b} + \frac{\sum_{i = d_j}^{e_j} w_i}{s_{\text{alloc}(i)}} \right\} + \frac{\delta_n}{b}$$

An example of formal definitions

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 \mathcal{L} = the longest path of the mapping as without overlap, but does not necessarily respect previous period

 $\mathcal{L} = (2K + 1).\mathcal{P}$, where K is the number of changes of processors

ntroduction Definitions **Examples** Complexity results Conclusion

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Period - No communication, no replication

Optimal period?



2 processors of speed 1

Optimal period?

$$\mathcal{P} = 5$$
, $\mathcal{S}_1 \mathcal{S}_3 \to P_1$, $\mathcal{S}_2 \mathcal{S}_4 \to P_2$

Perfect load-balancing in this case, but NP-hard (2-PARTITION)

Interval mapping?



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Heterogeneous platform?

$$\mathcal{S}_1 \rightarrow \mathcal{S}_2 \rightarrow \mathcal{S}_3 \rightarrow \mathcal{S}_4$$

2 1 3 4
Speed of $P_1: 2 \mid P_2: 3$

Optimal period?

$$\mathcal{P} = 5$$
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, $\mathcal{S}_1\mathcal{S}_2\mathcal{S}_3\to P_1$, $\mathcal{S}_4\to P_2$ – Polynomial algorithm? Classical chains-on-chains problem, dynamic programming works

Heterogeneous platform?

$$\mathcal{P} = 2$$
, $\mathcal{S}_1 \mathcal{S}_2 \mathcal{S}_3 \rightarrow P_2$, $\mathcal{S}_4 \rightarrow P_1$

Heterogeneous chains-on-chains, NP-hard

2 processors of speed 1

With overlap: optimal period?



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2 processors of speed 1

With overlap: optimal period?

$$\mathcal{P} = 5$$
, $\mathcal{S}_1 \mathcal{S}_3 \rightarrow P_1$, $\mathcal{S}_2 \mathcal{S}_4 \rightarrow P_2$

Perfect load-balancing both for computation and comm.

Optimal latency?

2 processors of speed 1

With overlap: optimal period?

$$\mathcal{P} = 5$$
, $\mathcal{S}_1 \mathcal{S}_3 \rightarrow P_1$, $\mathcal{S}_2 \mathcal{S}_4 \rightarrow P_2$

Perfect load-balancing both for computation and comm.

Optimal latency?

With only one processor, $\mathcal{L}=12$

No internal communication to pay

2 processors of speed 1

With overlap: optimal period?

$$\mathcal{P} = 5$$
, $\mathcal{S}_1 \mathcal{S}_3 \rightarrow P_1$, $\mathcal{S}_2 \mathcal{S}_4 \rightarrow P_2$

Perfect load-balancing both for computation and comm.

Optimal latency?

Same mapping as above: $\mathcal{L}=21$ with no period constraint

$$\mathcal{P}=21$$
, no conflicts

2 processors of speed 1

With overlap: optimal period?

$$\mathcal{P} = 5$$
, $\mathcal{S}_1 \mathcal{S}_3 \rightarrow P_1$, $\mathcal{S}_2 \mathcal{S}_4 \rightarrow P_2$

Perfect load-balancing both for computation and comm.

Optimal latency? with P = 5?

Progress step-by-step in the pipeline \rightarrow no conflicts

$$K=4$$
 processor changes, $\mathcal{L}=(2K+1).\mathcal{P}=9\mathcal{P}=45$

2 processors of speed 1

With no overlap: optimal period and latency?



2 processors of speed 1

With no overlap: optimal period and latency?

General mappings too difficult to handle:

restrict to interval mappings

2 processors of speed 1

With no overlap: optimal period and latency?

General mappings too difficult to handle:

restrict to interval mappings

$$\mathcal{P} = 8: \ \ S_1, S_2, S_3 \to P_1, \ S_4 \to P_2$$

2 processors of speed 1

With no overlap: optimal period and latency?

General mappings too difficult to handle:

restrict to interval mappings

$$\mathcal{P}=8$$
: $S_1,S_2,S_3\rightarrow P_1,\ S_4\rightarrow P_2$

$$\mathcal{L} = 12$$
: $S_1, S_2, S_3, S_4 \rightarrow P_1$

$$\mathcal{S}_1 \rightarrow \mathcal{S}_2 \rightarrow \mathcal{S}_3 \rightarrow \mathcal{S}_4 \\
14 \quad 4 \quad 2 \quad 4$$

Interval mapping, 4 processors, $\mathsf{s}_1=2$ and $\mathsf{s}_2=\mathsf{s}_3=\mathsf{s}_4=1$

Replicate interval $[S_u..S_v]$ on $P_1,...,P_q$

$$\mathcal{S}_u \dots \mathcal{S}_v$$
 on P_1 : data sets $\mathbf{1}, \mathbf{4}, \mathbf{7}, \dots$ $\mathcal{S}_u \dots \mathcal{S}_v$ on P_2 : data sets $\mathbf{2}, \mathbf{5}, \mathbf{8}, \dots$ $- \mathcal{S}_u \dots \mathcal{S}_v$ on P_3 : data sets $\mathbf{3}, \mathbf{5}, \mathbf{9}, \dots$

$$\mathcal{P} = rac{\sum_{k=u}^{v} \mathsf{w}_k}{q imes \mathsf{min}_i(\mathsf{s}_i)}$$
 and $\mathcal{L} = q imes \mathcal{P}$

Interval mapping, 4 processors, $\mathsf{s}_1=2$ and $\mathsf{s}_2=\mathsf{s}_3=\mathsf{s}_4=1$

Data Parallelize single stage S_k on P_1, \ldots, P_q

$$S (w = 16) \qquad P_1 (s_1 = 2) : \bullet \bullet \bullet \bullet \bullet \bullet$$

$$\Rightarrow P_2 (s_2 = 1) : \bullet \bullet \bullet \bullet \bullet$$

$$P_3 (s_3 = 1) : \bullet \bullet \bullet \bullet \bullet$$

$$P = \frac{w_k}{\sum_{i=1}^{q} s_i} \text{ and } \mathcal{L} = \mathcal{P}$$

Interval mapping, 4 processors, $\mathsf{s}_1=2$ and $\mathsf{s}_2=\mathsf{s}_3=\mathsf{s}_4=1$

Optimal period?



Interval mapping, 4 processors, $\mathsf{s}_1=2$ and $\mathsf{s}_2=\mathsf{s}_3=\mathsf{s}_4=1$

Optimal period?

$$\mathcal{S}_1 \overset{\mathrm{DP}}{\underset{\rightarrow}{\longrightarrow}} P_1 P_2, \, \mathcal{S}_2 \mathcal{S}_3 \mathcal{S}_4 \overset{\mathrm{REP}}{\underset{\rightarrow}{\longrightarrow}} P_3 P_4$$

$$\mathcal{P} = \max(\frac{14}{2+1}, \frac{4+2+4}{2\times 1}) = 5$$
, $\mathcal{L} = 14.67$

Optimal latency?



Interval mapping, 4 processors, $\mathsf{s}_1=2$ and $\mathsf{s}_2=\mathsf{s}_3=\mathsf{s}_4=1$

Optimal period?

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Optimal latency?
$$S_1 \stackrel{\mathrm{DP}}{\rightarrow} P_2 P_3 P_4$$
, $S_2 S_3 S_4 \rightarrow P_1$

$$\mathcal{P} = \max(\frac{14}{1+1+1}, \frac{4+2+4}{2}) = 5, \ \mathcal{L} = 9.67 \text{ (optimal)}$$



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Pipeline: minimizing period or latency

	Period			Latency		
	o2o	int	gen	o2o	int	gen
noc hom	P(t)	P(DP)	NPC(2P)		P(t)	
het	P(g)	NPC(*)	NPC(-)	P(g)	P(t)
noo fhom	P(t)	P(DP)	NPC(-)		P(t)	
chom	P(bs)	NPC(-)		P(g)	(g) P(t)	
fhet	NPC(CT)	NPC(-)		NPC(T)	NPC(*)	P(DP)
wov fhom	P(t)	P(DP)	NPC(-)	similar		
chom	P(g)	NPC(-)		to		
fhet	NPC(TC)	NPC(-)		noo		

noc: No comm – noo: Comm, no overlap – wov: Comm, with overlap

P: Polynomial (t) trivial – (g) greedy algorithm – (DP) dynamic programming algorithm – (bs) binary search algorithm

NPC: NP-complete (-) comes from simpler case – (2P) 2-Partition – (CT) Chinese traveller – (T) TSP – (*) involved reduction



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Pipeline: minimizing period and latency

	Bi-criteria				
	o2o	int	gen		
noc hom	P(t)	P(DP)	NPC(-)		
het	P(g)	NPC(-)			
noo fhom	P(t)	P(DP)	NPC(-)		
chom	P(m)	NPC(-)			
fhet	NPC(-)				
wov fhom	P(t)		NPC(-)		
chom	P(g)	NPC(-)			
fhet	NPC(-)				

noc: No comm – noo: Comm, no overlap – wov: Comm, with overlap

P: Polynomial (t) trivial – (g) greedy algorithm – (DP) dynamic programming algorithm – (m) matching+binary search algorithm

NPC: NP-complete (-) comes from mono-criterion



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- ... more cases I did not talk about
- period: rapidly NP-hard
- latency: difficult to define
- reliability: non-linear formula
- replication for period or reliability, data-parallelism, ...
- mix everything: even more exciting problems 🙂
- ... please ask me for details and references ...



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Related work

Qishi Wu et al— Directed platform graphs (WAN); unbounded multi-port with overlap; mono-criterion problems

Subhlok and Vondran- Pipeline on hom platforms: extended

Chains-to-chains- Heterogeneous, replicate/data-parallelize

Mapping pipelined computations onto clusters and grids— DAG [Taura et al.], DataCutter [Saltz et al.]

Energy-aware mapping of pipelined computations— [Melhem et al.], three-criteria optimization

Scheduling task graphs on heterogeneous platforms— Acyclic task graphs scheduled on different speed processors [Topcuoglu et al.]. Communication contention: 1-port model [Beaumont et al.]

Mapping skeletons onto clusters and grids— Use of stochastic process algebra [Benoit et al.]



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Conclusion

Definitions: Applications, platforms, and multi-criteria mappings

Theoretical side: Working out examples to show insight of problem complexity, and full complexity study

Practical side: not showed in this talk

- Several polynomial heuristics and simulations
- JPEG application, good results of the heuristics (close to LP solution)

Future work:

- Extend to other application graphs
- In particular, define latency for general DAGs (order communications)
- New heuristics for NP-hard cases, further experiments

