Multi-criteria scheduling of workflow applications

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Introduction

Definitions

Examples

Complexity result

Conclusion

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, ...
- Filtering: some tasks filter data

Mapping workflow applications onto heterogeneous platforms

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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

Reliability: inverse of \mathcal{FP} , probability of failure of the application (i.e. some data sets will not be processed)

Multi-criteria!

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Multi-criteria!

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)

In this talk: small examples to illustrate problem complexity

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In this talk: small examples to illustrate problem complexity

1 Definitions: Application, Platform and Mappings

2 Working out examples

Summary of complexity results

4 Conclusion

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Introduction	Definitions	Examples	Complexity results	Conclusion
Applicatio	on 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



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 Application model: communication costs
 Complexity results
 Conclusion

- 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)



Application model: adding selectivity

- Stages with selectivity: stage S_i transforms (filters) data of size δ to size σ_i × δ (σ_i: stage selectivity)
- Computation cost depends on the data size (previous σ)
- May add dependencies to exploit selectivity



- S₁ and S₄ process file of initial size 1; S₁ removes even line numbers; S₂ removes two-third of the file
- Combined file of size $\frac{1}{2}.\frac{1}{3}=\frac{1}{6}$ (no cost for join)
- S₂ duplicates the file
- S₃ processes but does not alter the file

Platform model



- p processors P_u , $1 \le u \le p$, fully interconnected
- s_u : speed of processor P_u
- bidirectional link link_{$u,v} : <math>P_u \rightarrow P_v$, bandwidth b_{u,v}</sub></sub>
- 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



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

$$\label{eq:fully Heterogeneous} \begin{split} & \textit{Fully Heterogeneous} - \textit{Fully heterogeneous architectures, } \mathsf{s}_u \neq \mathsf{s}_v \\ & \text{and } \mathsf{b}_{u,v} \neq \mathsf{b}_{u',v'} \text{: hierarchical platforms, grids} \end{split}$$

Fully Homogeneous – Identical processors $(s_{\mu} = s)$ and links $(b_{\mu,\nu} = b)$: typical parallel machines

Failure Homogeneous – Identically reliable processors ($fp_{,i} = fp_{,v}$)

Communication Homogeneous - Different-speed processors $(s_{\mu} \neq s_{\nu})$, identical links $(b_{\mu,\nu} = b)$: networks of workstations, clusters

Fully Heterogeneous – Fully heterogeneous architectures, $s_{\mu} \neq s_{\nu}$ and $b_{\mu,\nu} \neq b_{\mu',\nu'}$: hierarchical platforms, grids *Failure Heterogeneous* – Different failure probabilities $(fp_{\mu} \neq fp_{\nu})$

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 Platform model: communications
 no overlap vs overlap
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 soverlap
 soverlap

no overlap vs overlap

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



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 Platform model: communications
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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



- Define connected-subgraph mapping (instead of interval)
- Replication: independent sets of processors, instead of a single processor as above



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- Other applications: one-to-one and general always defined
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- 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)
- Minimize latency \mathcal{L} (time to process a data set)
- \bullet Minimize application failure probability \mathcal{FP}

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- $\bullet\,$ Minimize latency ${\cal L}$ (time to process a data set)
- $\bullet\,$ Minimize application failure probability \mathcal{FP}

Multi-criteria

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

Mono-criterion

- Minimize period \mathcal{P} (inverse of throughput)
- Minimize latency \mathcal{L} (time to process a data set)
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- How to define it? Minimize $\alpha . \mathcal{P} + \beta . \mathcal{L} + \gamma . \mathcal{FP}$?
- Values which are not comparable
- \bullet Minimize ${\cal P}$ for a fixed latency and failure
- \bullet Minimize ${\cal L}$ for a fixed period and failure
- \bullet Minimize \mathcal{FP} for a fixed period and latency

Mono-criterion

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

Bi-criteria

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



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

$$\mathcal{P} = \max_{1 \le j \le m} \left\{ \frac{\delta_{d_j - 1}}{b} + \frac{\sum_{i = d_j}^{e_j} w_i}{s_{\text{alloc}(j)}} + \frac{\delta_{e_j}}{b} \right\}$$


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

An example of formal definitions

- Pipeline application, INTERVAL MAPPING
- Period/Latency problem with no replication
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$$\mathcal{P} = \max_{1 \le j \le m} \left\{ \max\left\{ \frac{\sum_{i=d_j}^{e_j} w_i}{\mathsf{s}_{\mathsf{alloc}(j)}}, \ \frac{\delta_{d_j-1}}{\mathsf{b}}, \ \frac{\delta_{d_j-1}}{\mathsf{B}^i}, \ \frac{\delta_{e_j}}{\mathsf{b}}, \ \frac{\delta_{e_j}}{\mathsf{B}^o} \right\} \right\}$$

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

Definitions: Application, Platform and Mappings

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Period - No communication, no replication

 $2\ \text{processors}$ of speed 1

Optimal period?

Period - No communication, no replication

2 processors of speed 1

Optimal period?

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

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

Interval mapping?

Examples

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 $\mathcal{P} = 5$, $\mathcal{S}_1 \mathcal{S}_3 \rightarrow P_1$, $\mathcal{S}_2 \mathcal{S}_4 \rightarrow P_2$ Perfect load-balancing in this case, but NP-hard (2-PARTITION)

Interval mapping?

 $\mathcal{P}=\text{6}, \ \ \mathcal{S}_1\mathcal{S}_2\mathcal{S}_3 \to \textit{P}_1, \ \ \mathcal{S}_4 \to \textit{P}_2 \ \ - \ \ \text{Polynomial algorithm}?$

Period - No communication, no replication

\mathcal{S}_1	\rightarrow	\mathcal{S}_2	\rightarrow	\mathcal{S}_3	\rightarrow	\mathcal{S}_4
2		1		3		4

2 processors of speed 1

Optimal period?

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Interval mapping?

 $\mathcal{P} = 6$, $\mathcal{S}_1 \mathcal{S}_2 \mathcal{S}_3 \rightarrow P_1$, $\mathcal{S}_4 \rightarrow P_2$ – Polynomial algorithm? Classical chains-on-chains problem, dynamic programming works IntroductionDefinitionsExamplesComplexity resultsConclusionPeriod - No communication, no replication $S_1 \rightarrow S_2 \rightarrow S_3 \rightarrow S_4$
 $2 \quad 1 \quad 3 \quad 4$
 $s_1 = 2$ and $s_2 = 3$ Optimal period?
 $\mathcal{P} = 5, S_1S_3 \rightarrow P_1, S_2S_4 \rightarrow P_2$

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Heterogeneous platform?

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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

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Perfect load-balancing both for computation and comm.

Achieved latency?



Perfect load-balancing both for computation and comm.

Achieved latency?

With only one processor, $\mathcal{L} = 12$ No internal communication to pay



 Introduction Definitions Examples Latency - No replication, different comm. models 2 processors of speed 1 With overlap: optimal period? $\mathcal{P} = 5$, $\mathcal{S}_1 \mathcal{S}_3 \to \mathcal{P}_1$, $\mathcal{S}_2 \mathcal{S}_4 \to \mathcal{P}_2$ Perfect load-balancing both for computation and comm. Achieved latency? with $\mathcal{P} = 5$? Progress step-by-step in the pipeline \rightarrow no conflicts K = 4 processor changes, $\mathcal{L} = (2K + 1) \mathcal{P} = 9\mathcal{P} = 45$ \dots period k | period k + 1 | period k + 2 | \dots | ds^(k-8) $ds^{(k-7)}$ $ds^{(k-6)}$ $P_2 \rightarrow out$



With no overlap: optimal period and latency?



With no overlap: optimal period and latency? General mappings too difficult to handle: restrict to interval mappings IntroductionDefinitionsExamplesComplexity resultsConclusionLatency - No replication, different comm. models $\frac{1}{\rightarrow}$ S_1 $\frac{4}{\rightarrow}$ S_2 $\frac{4}{\rightarrow}$ S_3 $\frac{1}{\rightarrow}$ S_4 $\frac{1}{\rightarrow}$ 213422processors of speed 1

With no overlap: optimal period and latency? General mappings too difficult to handle: restrict to interval mappings

$$\mathcal{P}=$$
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With no overlap: optimal period and latency? General mappings too difficult to handle: restrict to interval mappings

$$\begin{aligned} \mathcal{P} &= 8: \quad S_1, S_2, S_3 \rightarrow P_1, \ S_4 \rightarrow P_2 \\ \mathcal{L} &= 12: \quad S_1, S_2, S_3, S_4 \rightarrow P_1 \end{aligned}$$

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, \ldots, P_q

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

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?

$$\mathcal{S}_1 \stackrel{\mathrm{DP}}{\xrightarrow{}} \mathcal{P}_1 \mathcal{P}_2, \ \mathcal{S}_2 \mathcal{S}_3 \mathcal{S}_4 \stackrel{\mathrm{REP}}{\xrightarrow{}} \mathcal{P}_3 \mathcal{P}_4$$

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

Optimal latency?

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

Optimal period?

$$\mathcal{S}_1 \stackrel{\mathrm{DP}}{\xrightarrow{}} \mathcal{P}_1 \mathcal{P}_2, \ \mathcal{S}_2 \mathcal{S}_3 \mathcal{S}_4 \stackrel{\mathrm{REP}}{\xrightarrow{}} \mathcal{P}_3 \mathcal{P}_4$$

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

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

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

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Definitions: Application, Platform and Mappings

- 2 Working out examples
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Filters: stages with selectivity

- One-to-one mappings of independent tasks
 - No communication, homogeneous processors: period and latency polynomial
 - With heterogeneous processors: both problems NP-hard
 - With homogeneous communication, overlap or no-overlap: all problems NP-hard
- General mappings: everything is NP-hard (2-partition)
- For references, please ask me

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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)	P(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 or 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)	P(DP)	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

- ... 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 ③

- ... 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 🙂

Complexity results....

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Definitions: Application, Platform and Mappings

- 2 Working out examples
- Summary of complexity results



Examples

Complexity results

Conclusion

Related work

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 pipelined computations onto special-purpose architectures-FPGA arrays [Fabiani et al.]. Fault-tolerance for embedded systems [Zhu et al.]

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

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