

# Performance and energy optimization of concurrent pipelined applications

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École Normale Supérieure de Lyon, France

CCGSC 2010, Flat Rock, NC

# Motivations

- Mapping **concurrent pipelined applications** onto **distributed platforms**: practical applications, but difficult problems
- Assess problem hardness  $\Rightarrow$  different mapping rules and platform characteristics
- **Energy saving** is becoming a crucial problem
- Several **concurrent objective functions**: period, latency, power
- $\Rightarrow$  Multi-criteria approach: minimize power consumption while guaranteeing some performance
- Exhaustive complexity study
- Heuristics on most general (NP-complete) case

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# Why bother with energy?

- Minimizing total energy consumed by processors: very important objective (economic and environmental reasons)
- M. P. Mills, [The internet begins with coal](#), Environment and Climate News (1999)
- Algorithmic techniques:
  - Shut down idle processors
  - Dynamic speed scaling
  - The higher the speed, the higher the power consumption
  - $Power = f \times V^2$ , and  $V$  (voltage) increases with  $f$  (frequency)
  - Speed  $s$ :  $P(s) = s^\alpha + P_{static}$ , with  $2 \leq \alpha \leq 3$
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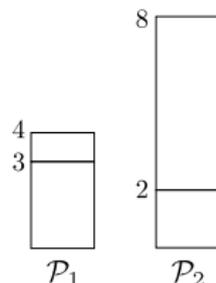
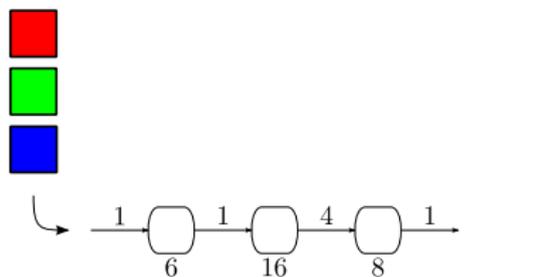
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# Motivating example

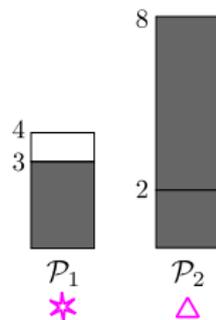
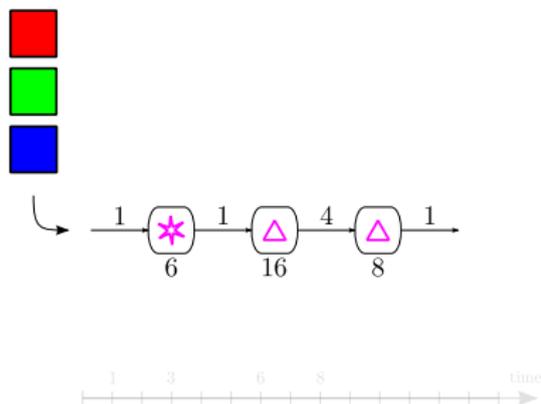


$\mathcal{P}_1$

$\mathcal{P}_2$

- Period:  $T = 3$
- Latency:  $L = 8$

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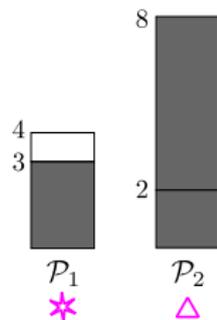
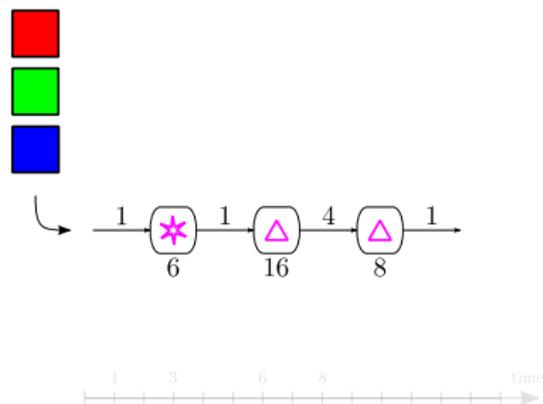


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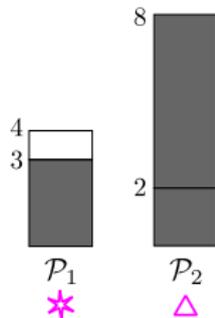
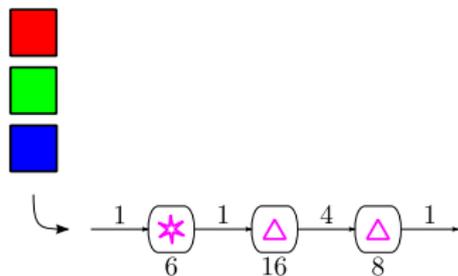
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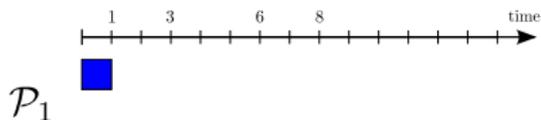
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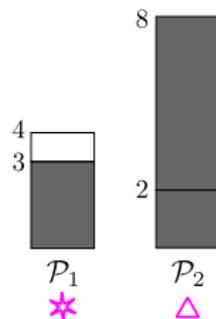
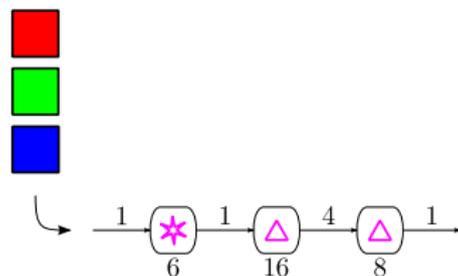
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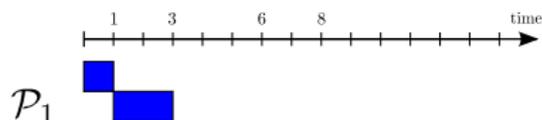
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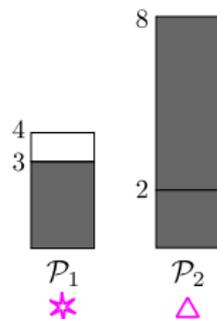
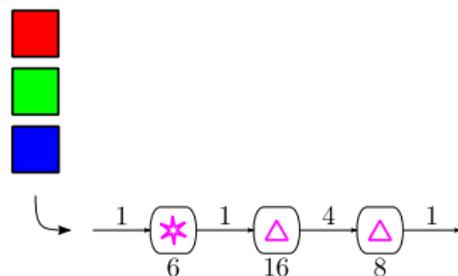
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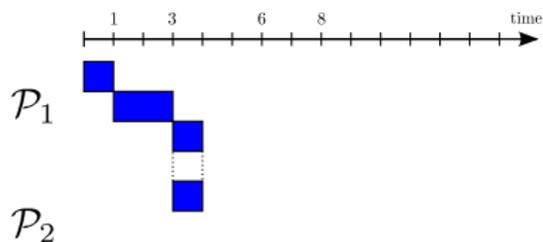
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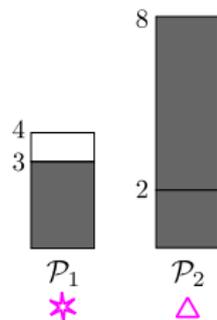
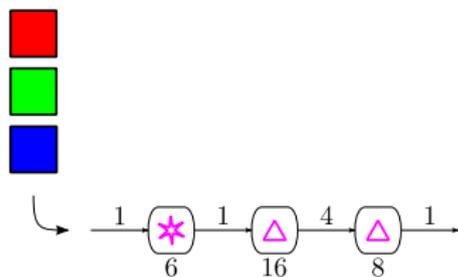


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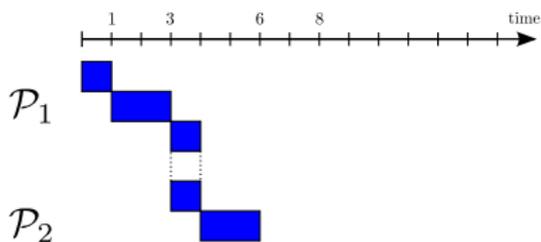


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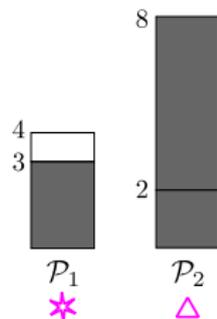
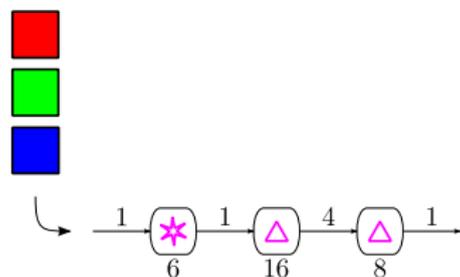


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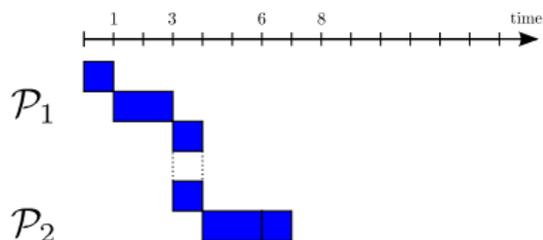


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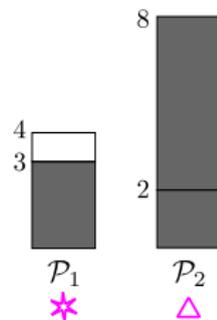
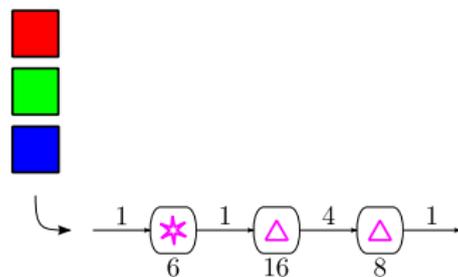


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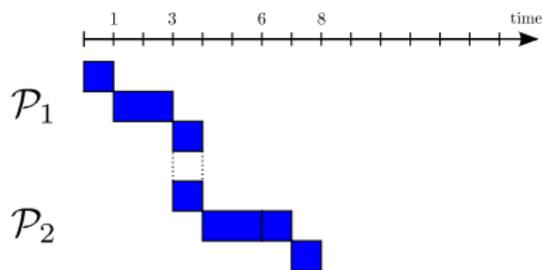


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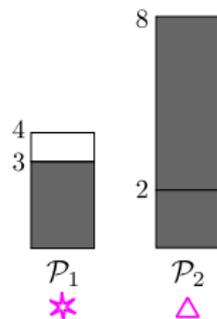
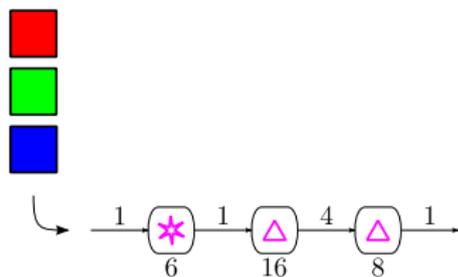


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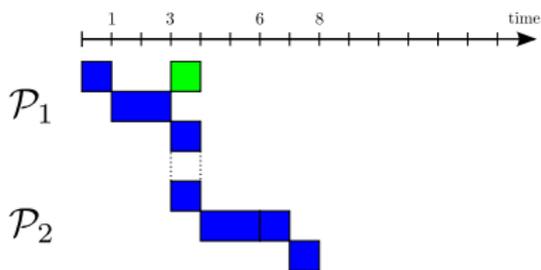


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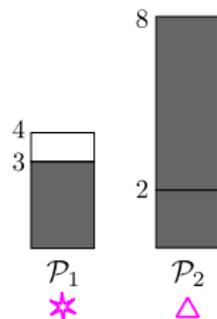
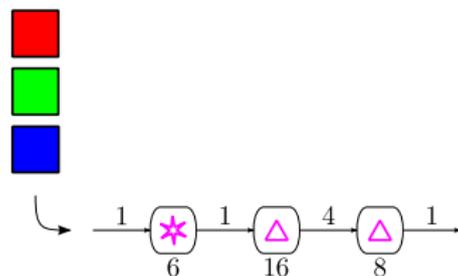


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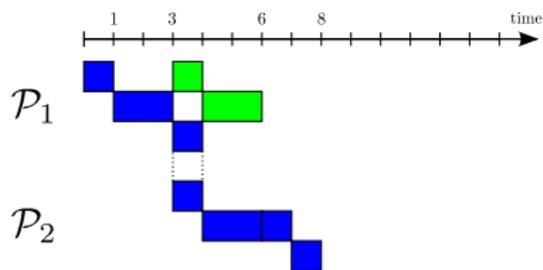


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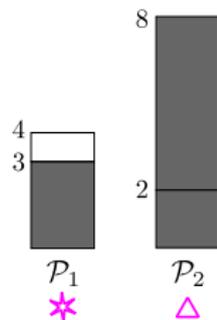
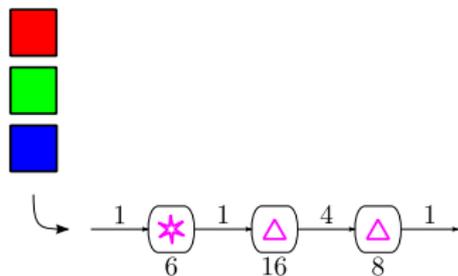


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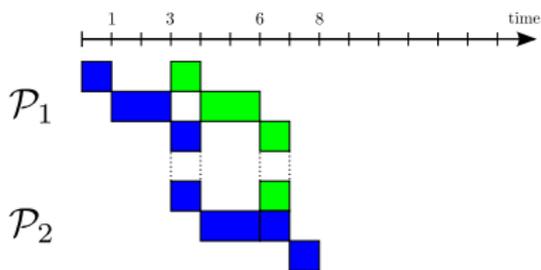


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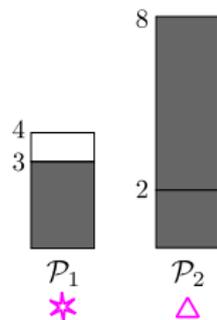
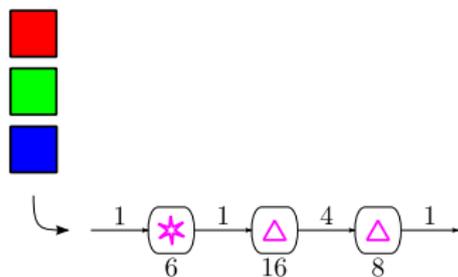


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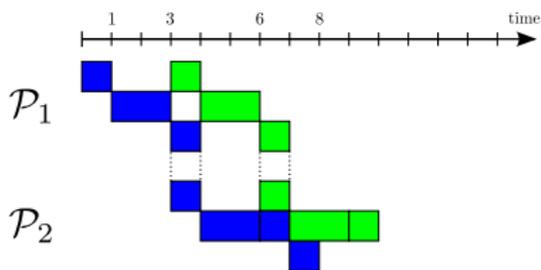


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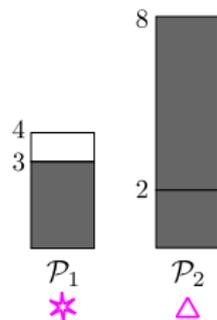
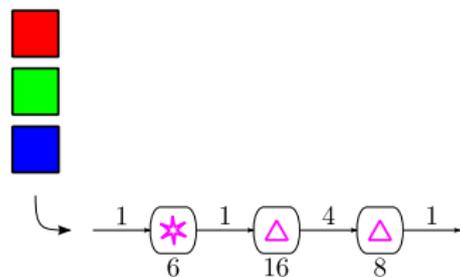


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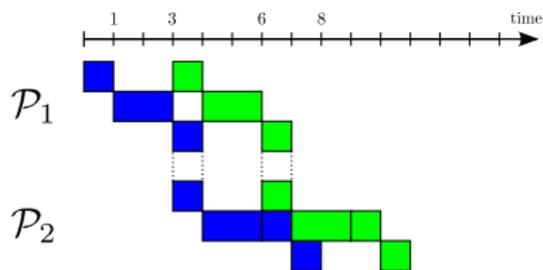


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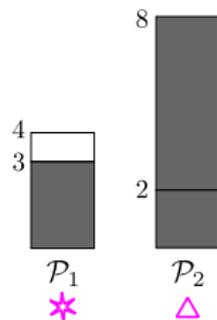
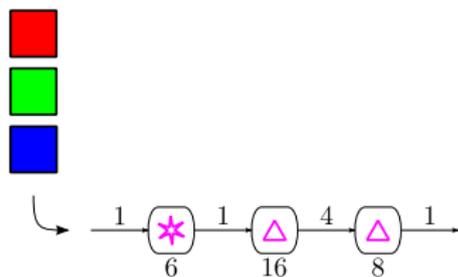
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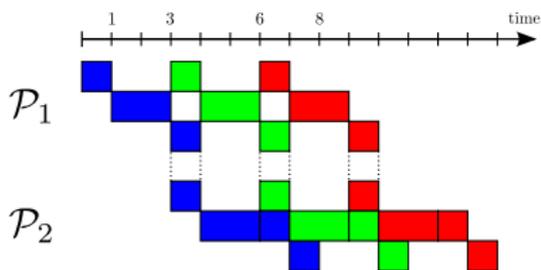
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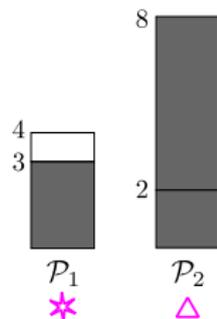
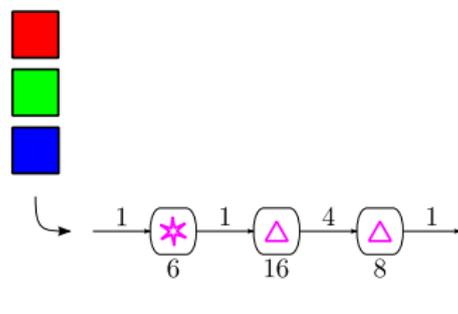


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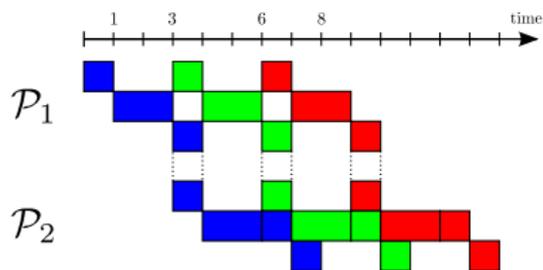


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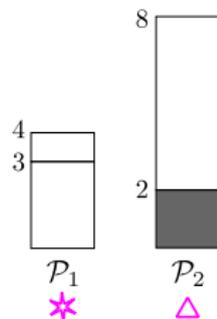
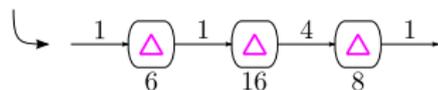


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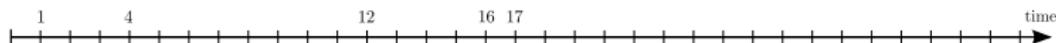


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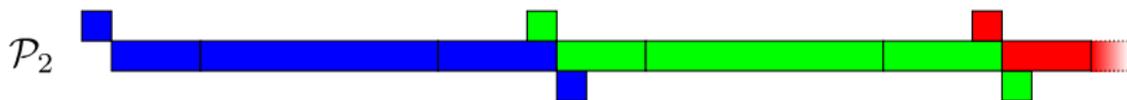
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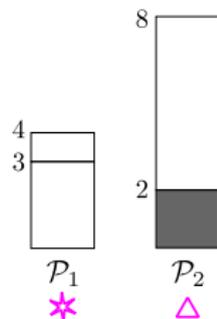
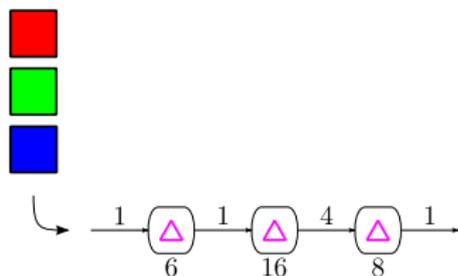


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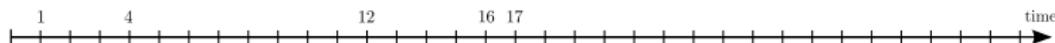
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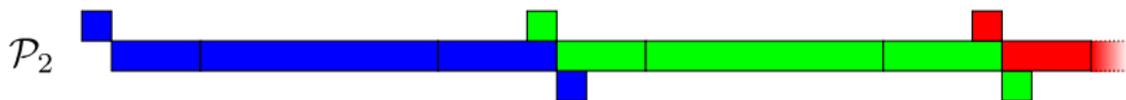


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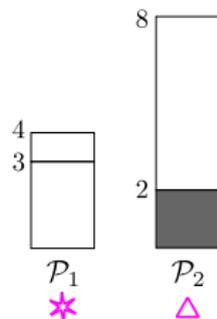
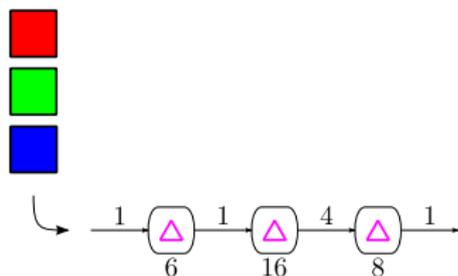


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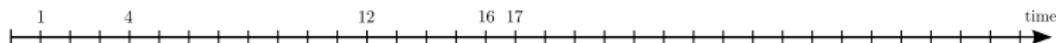
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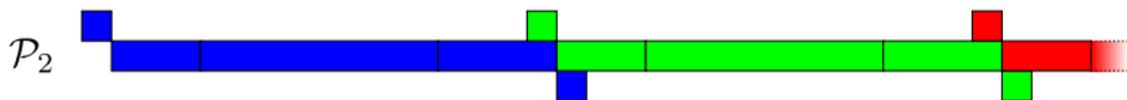


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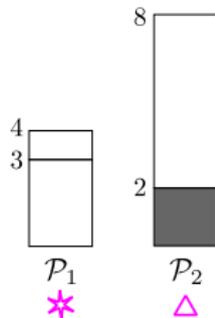
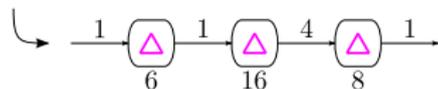


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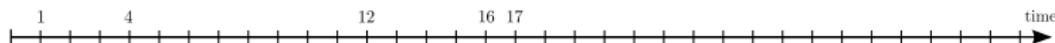
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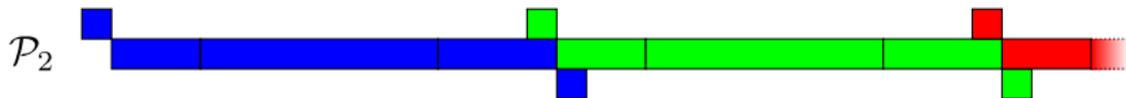


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$\mathcal{P}_1$



- Period:  $T = 3$      $T = 15$

- Latency:  $L = 8$      $L = 17$

# Outline of the talk

- 1 Framework
  - Application and platform
  - Mapping rules
  - Metrics
- 2 Complexity results
  - Mono-criterion problems
  - Bi-criteria problems
  - Tri-criteria problems
  - With resource sharing
- 3 Experiments
  - Heuristics
  - Experiments
  - Summary
- 4 Conclusion

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# Application model and execution platform

- **Concurrent pipelined applications**

- $w_a^i$ : weight of stage  $\mathcal{S}_a^i$  ( $i^{\text{th}}$  stage of application  $a$ )
- $\delta_a^i$ : size of outgoing data of  $\mathcal{S}_a^i$

- Processors with **multiple speeds** (or modes):  $\{s_{u,1}, \dots, s_{u,m_u}\}$   
Constant speed during the execution

- **Platform** fully interconnected;

$b_{u,v}$ : bandwidth between processors  $\mathcal{P}_u$  and  $\mathcal{P}_v$ ;

overlap or non-overlap of communications and computations

- Three platform types:

- Fully homogeneous, or speed homogeneous
- Communication homogeneous, or speed heterogeneous
- Fully heterogeneous

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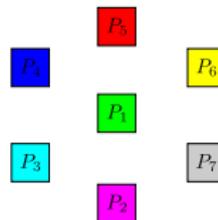
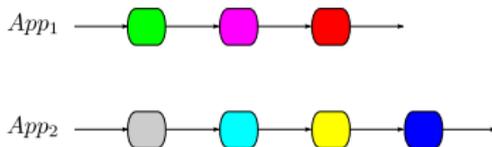
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- **Platform** fully interconnected;  
 $b_{u,v}$ : bandwidth between processors  $\mathcal{P}_u$  and  $\mathcal{P}_v$ ;  
overlap or non-overlap of communications and computations
- Three platform types:
  - Fully homogeneous, or speed homogeneous
  - Communication homogeneous, or speed heterogeneous
  - Fully heterogeneous

# Application model and execution platform

- **Concurrent pipelined applications**
  - $w_a^i$ : weight of stage  $\mathcal{S}_a^i$  ( $i^{\text{th}}$  stage of application  $a$ )
  - $\delta_a^i$ : size of outgoing data of  $\mathcal{S}_a^i$
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# Mapping rules

- Mapping with **no processor sharing**: relevant in practice (security rules)
  - One-to-one mapping



- Interval mapping

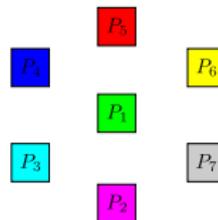


- General mapping **with resource sharing**: better resource utilization

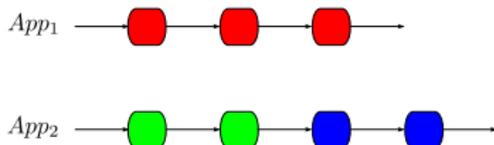


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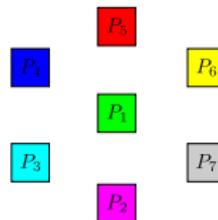


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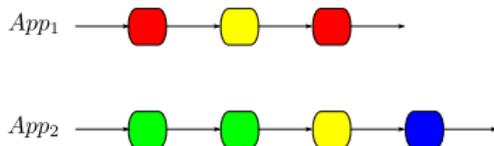
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# Metrics without resource sharing

Interval mapping on a single application with no resource sharing;  
 $k$  intervals  $I_j$  of stages from  $\mathcal{S}^{d_j}$  to  $\mathcal{S}^{e_j}$

- **Period  $T$**  of an application: minimum delay between the processing of two consecutive data sets

$$T^{(overlap)} = \max_{j \in \{1, \dots, k\}} \left( \max \left( \frac{\delta^{d_j-1}}{b_{\text{alloc}(d_j-1), \text{alloc}(d_j)}}, \frac{\sum_{i=d_j}^{e_j} w^i}{s_{\text{alloc}(d_j)}}, \frac{\delta^{e_j}}{b_{\text{alloc}(d_j), \text{alloc}(e_j+1)}} \right) \right)$$

- **Latency  $L$**  of an application: time, for a data set, to go through the whole pipeline

$$L = \frac{\delta^0}{b_{\text{alloc}(0), \text{alloc}(1)}} + \sum_{j=1}^m \left( \sum_{i=d_j}^{e_j} \frac{w^i}{s_{\text{alloc}(d_j)}} + \frac{\delta^{e_j}}{b_{\text{alloc}(d_j), \text{alloc}(e_j+1)}} \right)$$

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With classical latency definition, **NP-completeness of the execution scheduling**, given a mapping with a period/latency objective

⇒ for general mappings, **latency model of Özgüner**:

$L = (2m - 1)T$ , where  $m - 1$  is the number of processor changes, and  $T$  the period of the application

Period given ⇒ bound on number of processor changes

Given an application, we can **check if the mapping is valid**, given a bound on period and latency per application:

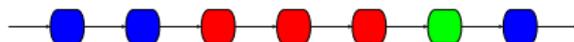
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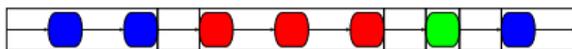
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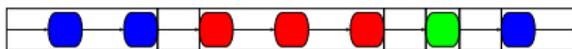
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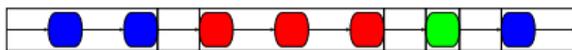
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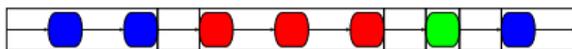
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# Optimization problems

- Minimizing **one criterion**:
  - Period or latency: minimize  $\max_a W_a \times T_a$  or  $\max_a W_a \times L_a$
  - Power: minimize  $P = \sum_u P(u)$
- **Fixing one criterion**:
  - Fix the period or latency of each application  
→ fix an array of periods or latencies
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- **Multi-criteria approach**: minimizing one criterion, fixing the other ones
- Energy criterion = power consumption, i.e., energy per time unit  $\Rightarrow$  combination power/period

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  - Application and platform
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- 2 Complexity results
  - Mono-criterion problems
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  - Tri-criteria problems
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# Mono-criterion complexity results

## Period minimization:

	proc-hom com-hom	special-app <sup>1</sup>	proc-het com-hom	com-het
one-to-one	polynomial (binary search)			NP-complete
interval	polynomial	NP-complete	NP-complete	

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# Latency minimization (1)

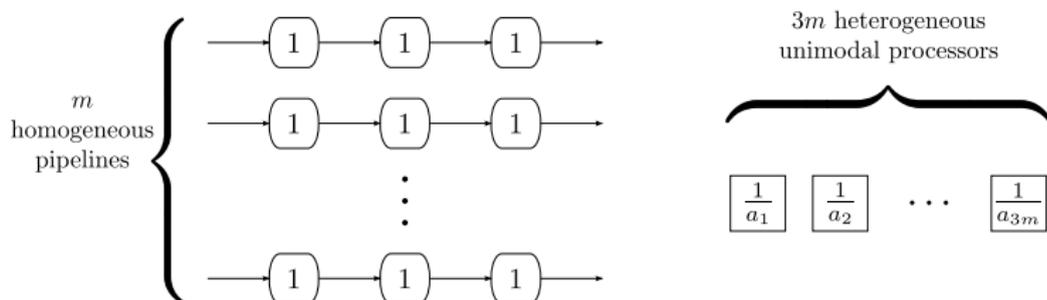
- Problem: one-to-one mapping - many applications - heterogeneous platform - no communication - homogeneous pipelines - minimize  $\max_a L_a$
- Single application: greedy polynomial algorithm
- Many applications: reduction from 3-PARTITION
- 3-PARTITION:
  - Input:  $3m + 1$  integers  $a_1, a_2, \dots, a_{3m}$  and  $B$  such that  $\sum_i a_i = mB$
  - Does there exist a partition  $I_1, \dots, I_m$  of  $\{1, \dots, 3m\}$  such that for all  $j \in \{1, \dots, m\}$ ,  $|I_j| = 3$  and  $\sum_{i \in I_j} a_i = B$ ?

# Latency minimization (2)

- 3-PARTITION: renumbering of the  $a_i$  such that:

$$\begin{cases} a_{1,1} + a_{1,2} + a_{1,3} = B \\ a_{2,1} + a_{2,2} + a_{2,3} = B \\ \vdots \\ a_{m,1} + a_{m,2} + a_{m,3} = B \end{cases}$$

- Reduction:



Can we obtain a latency  $L^0 \leq B$ ?

- Equivalence of problems

# Bi-criteria complexity results

## Period/latency minimization:

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one-to-one or interval	polynomial	NP-complete		

## Power/period minimization:

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interval	polynomial	NP-complete		

# Bi-criteria complexity results

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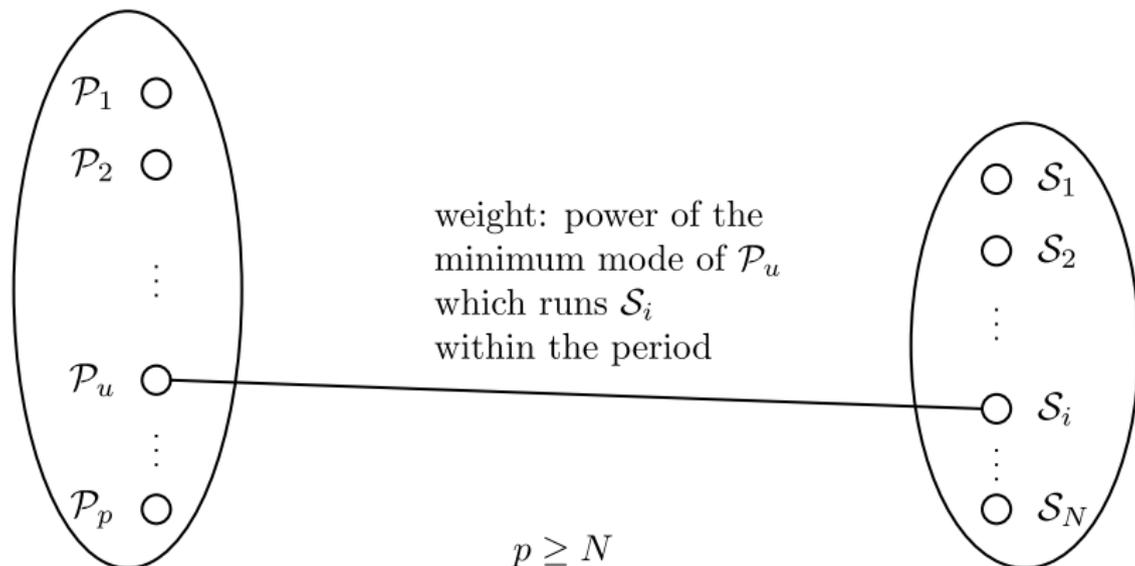
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## Power/period minimization:

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# Power/period minimization

- Problem: one-to-one mapping - many applications - communication homogeneous platform - power minimization for a given array of periods
- Minimum weighted matching of a bipartite graph



# Bi-criteria complexity results

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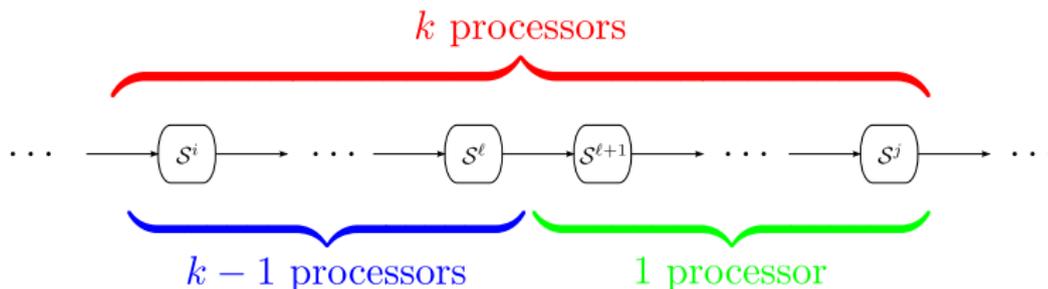
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one-to-one	polynomial	(minimum matching)		NP-complete
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# Single application (1)

- Problem: interval mapping - single application - fully homogeneous platform - power minimization for a given period
- $P(i, j, k)$ : minimum power to run stages  $\mathcal{S}^i$  to  $\mathcal{S}^j$  using exactly  $k$  processors  $\rightarrow$  looking for  $\min_{1 \leq k \leq p} P(1, n, k)$
- Recurrence relation:

$$P(i, j, k) = \min_{1 \leq \ell \leq j-1} (P(i, \ell, k-1) + P(\ell+1, j, 1))$$



# Single application (2)

- $P(i, i, q) = +\infty$  if  $q > 1$
- $\mathcal{F}_i^j$ : possible powers of a processor running the stages  $\mathcal{S}^i$  to  $\mathcal{S}^j$ , fulfilling the period constraint

$$\mathcal{F}_i^j = \left\{ P_{dyn}(s_\ell) + P_{stat}, \max \left( \frac{\delta^{i-1}}{b}, \frac{\sum_{k=i}^j w^k}{s_\ell}, \frac{\delta^j}{b} \right) \leq T, \ell \in \{1, \dots, m\} \right\}$$

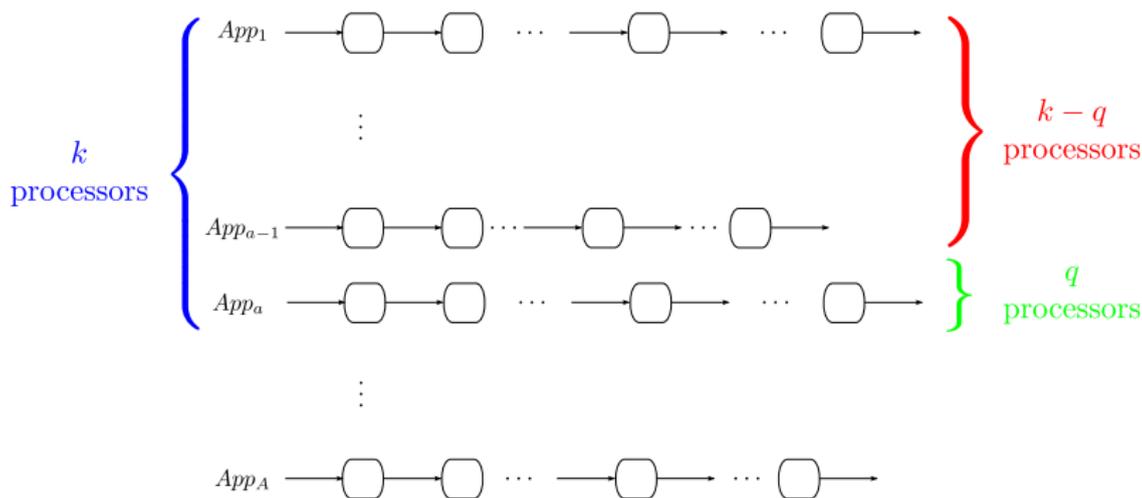
- $P(i, j, 1) = \begin{cases} \min \mathcal{F}_i^j & \text{if } \mathcal{F}_i^j \neq \emptyset \\ +\infty & \text{otherwise} \end{cases}$

# Many applications (1)

- Problem: interval mapping - fully homogeneous platform - power minimization for given periods by application
- $P_a^q$ : minimum power consumed by  $q$  processors so that the period constraint on the application  $a$  is met, found by the previous dynamic programming
- $P(a, k)$ : minimum power consumed by  $k$  processors on the applications  $1, \dots, a$ , unknown
- Initialization:  $\forall k \in \{1, \dots, p\} \quad P(1, k) = P_1^k$

# Many applications (2)

- Recurrence:  $P(a, k) = \min_{1 \leq q < k} (P(a-1, k-q) + P_a^q)$



# Tri-criteria complexity results

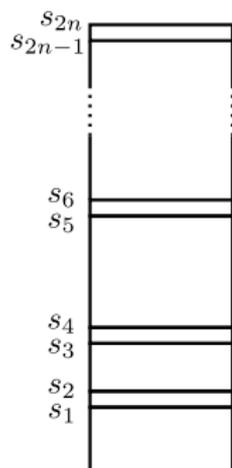
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Reduction from 2-PARTITION

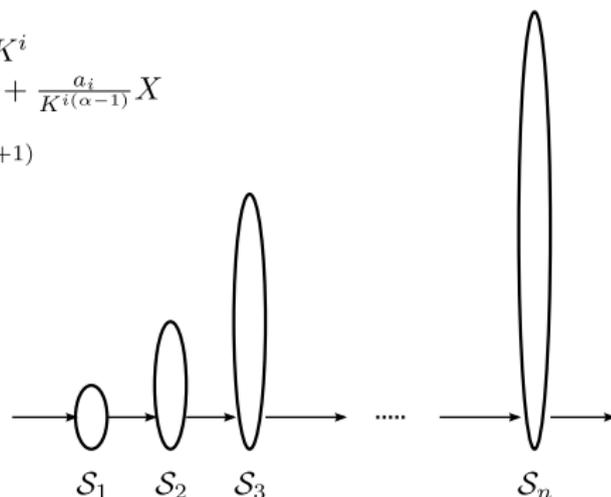
(Instance of 2-PARTITION:  $a_1, a_2, \dots, a_n$  with  $\sigma = \sum_{i=1}^n a_i$ )

# Problem instance

One-to-one mapping - fully homogeneous platform



$$\begin{cases} s_{2i-1} = K^i \\ s_{2i} = K^i + \frac{a_i}{K^{i(\alpha-1)}} X \\ w_i = K^{i(\alpha+1)} \end{cases}$$



$P^0 = P^* + \alpha X(\sigma/2 + 1/2)$ ,  $L^0 = L^* - X(\sigma/2 - 1/2)$ ,  $T^0 = L^0$   
 where  $P^*$  and  $L^*$  are power and latency when each  $S_i$  is run at speed  $s_{2i-1}$

# Main ideas

- $K$  big enough and  $X$  small enough so that the stage  $\mathcal{S}_i$  must be processed at speed  $s_{2i-1}$  or  $s_{2i}$
- For a subset  $\mathcal{I}$  of  $\{1, \dots, n\}$ , if ( $\mathcal{S}_i$  is run at speed  $s_{2i}$   $\Leftrightarrow i \in \mathcal{I}$ ),

$$P = P^* + \sum_{i \in \mathcal{I}} (\alpha a_i X + o(X)) \quad , \quad L = L^* - \sum_{i \in \mathcal{I}} (a_i X - o(X))$$

- Recall:

$$P^0 = P^* + \alpha X(\sigma/2 + 1/2) \quad , \quad L^0 = L^* - X(\sigma/2 - 1/2)$$

# And for general mappings with resource sharing?

- Exhaustive complexity study **with no resource sharing**: new polynomial algorithms for multiple applications and results of NP-completeness
- With the simplified latency model, **tri-criteria polynomial dynamic programming algorithm** with **no resource sharing** and **speed-homogeneous platforms**
- With **resource sharing** or **speed-heterogeneous platforms**, all problem instances are **NP-hard**, even for only **period minimization**

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# Heuristics

Tri-criteria problem: **power consumption minimization given a bound on period and latency per application**, on speed heterogeneous platform

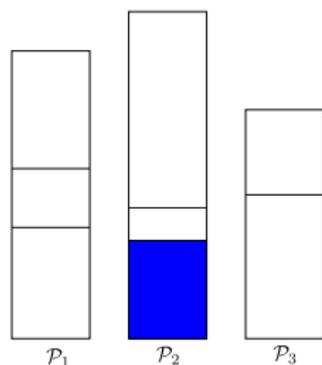
Each heuristic (except H2) exists in two variants: **interval mapping without resource sharing** and **general mapping with resource sharing** in order to evaluate the impact of processor reuse

Latency model of Özgüner:  $L = (2m - 1)T$

- H1: random cuts
- H2: one entire application per processor (assignment problem)
- H2-split: interval splitting
- H3: two-step heuristic: choose a speed distribution and find a valid mapping (variants on both steps)

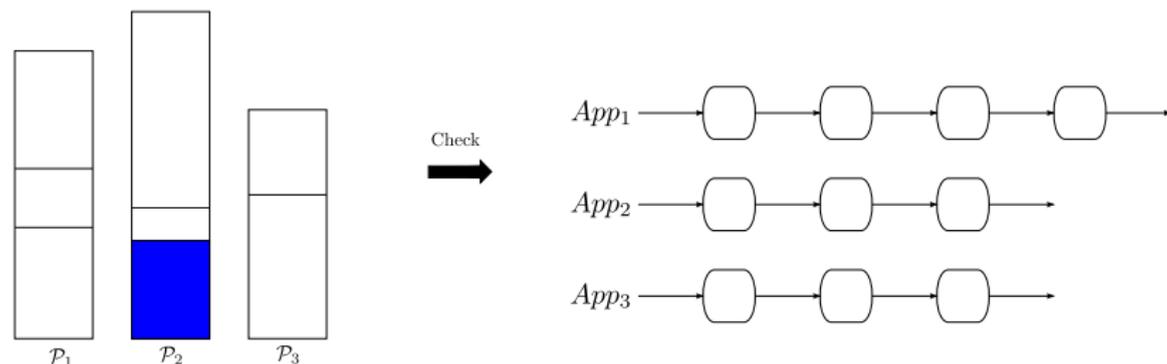
# H3-energy

Fix processor speeds



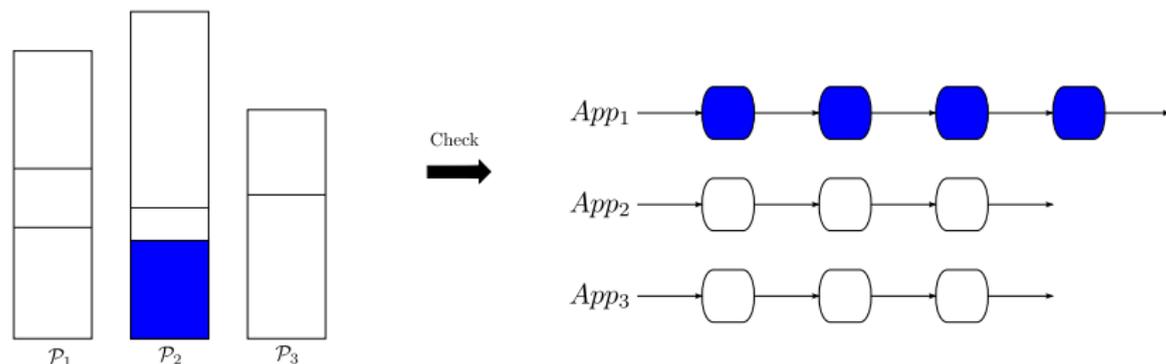
# H3-energy

Mapping heuristic: find a valid mapping



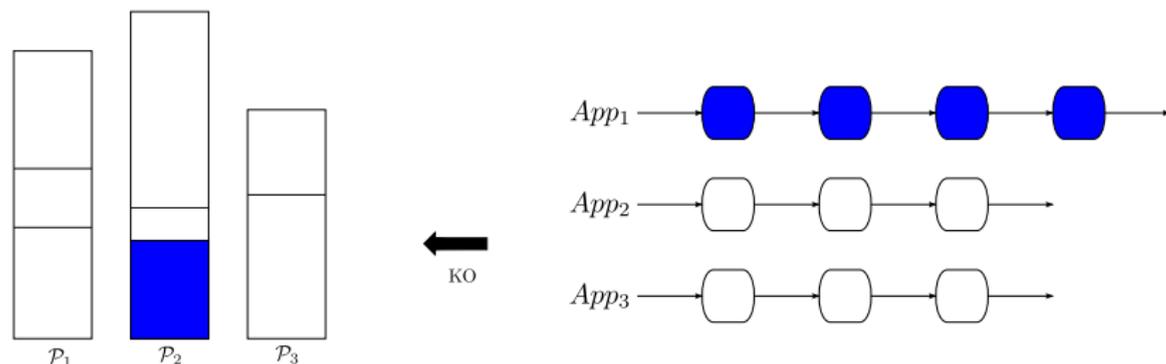
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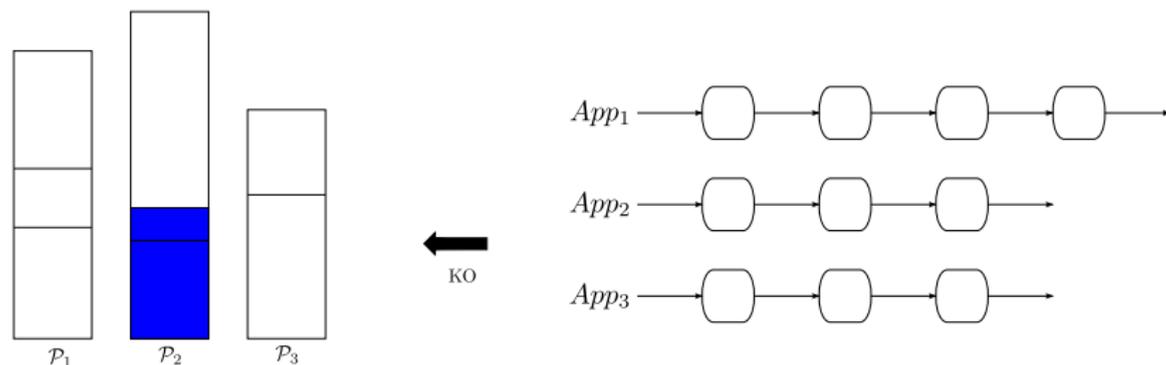
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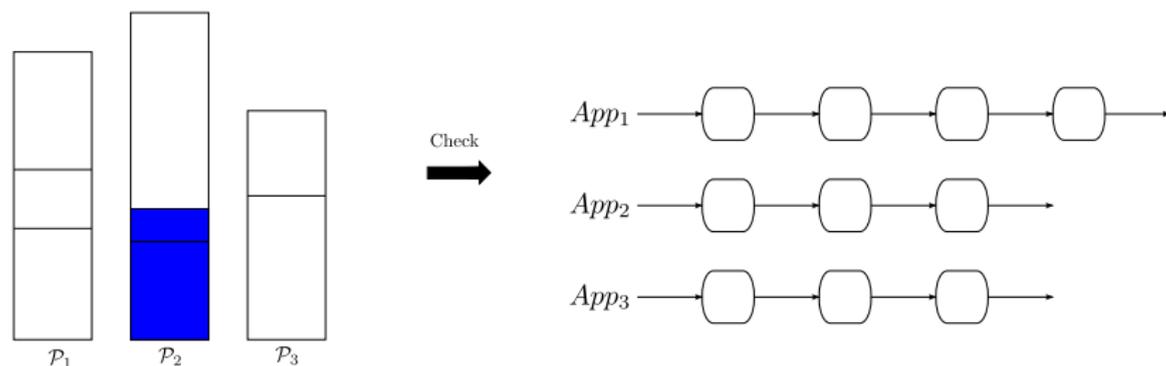
# H3-energy

Iterate the process: increase processor speeds



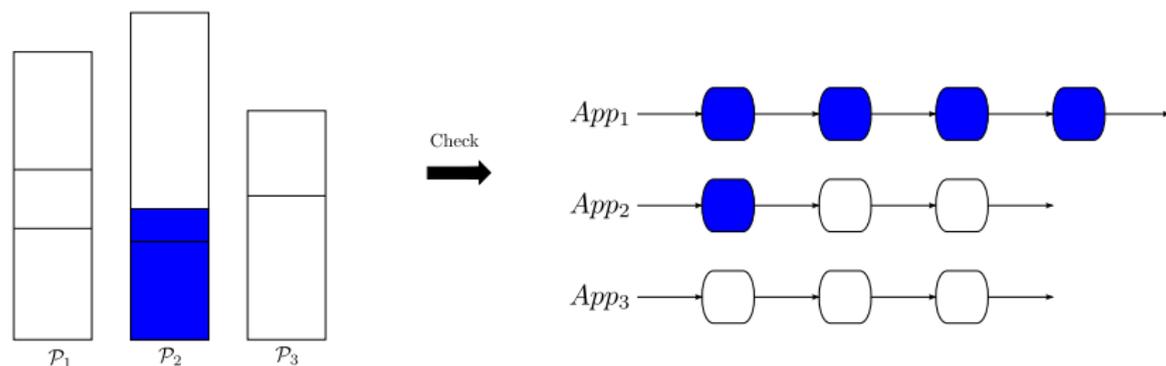
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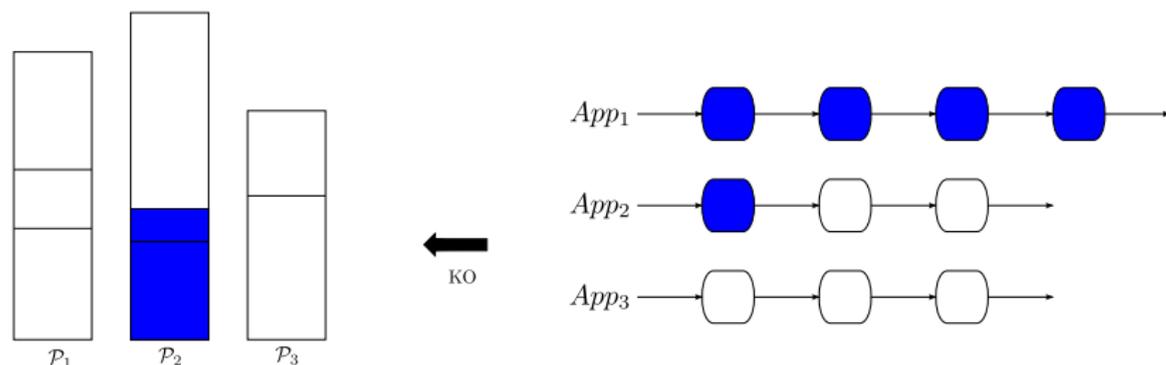
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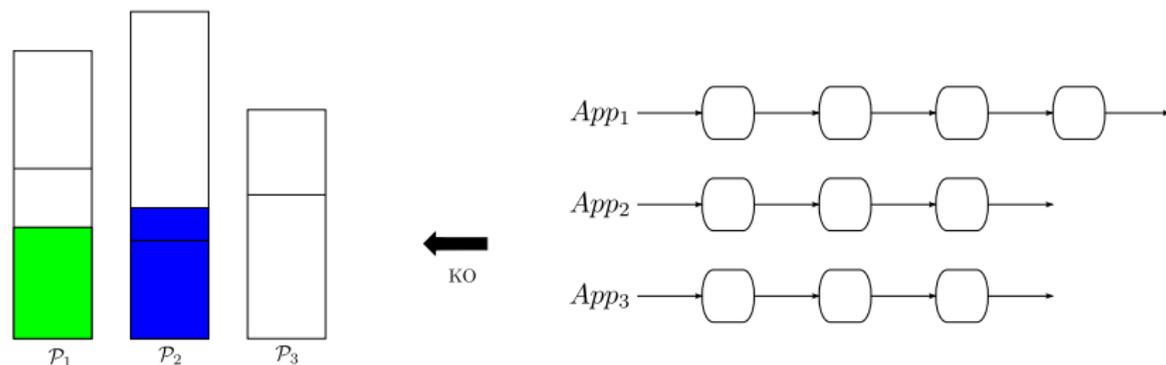
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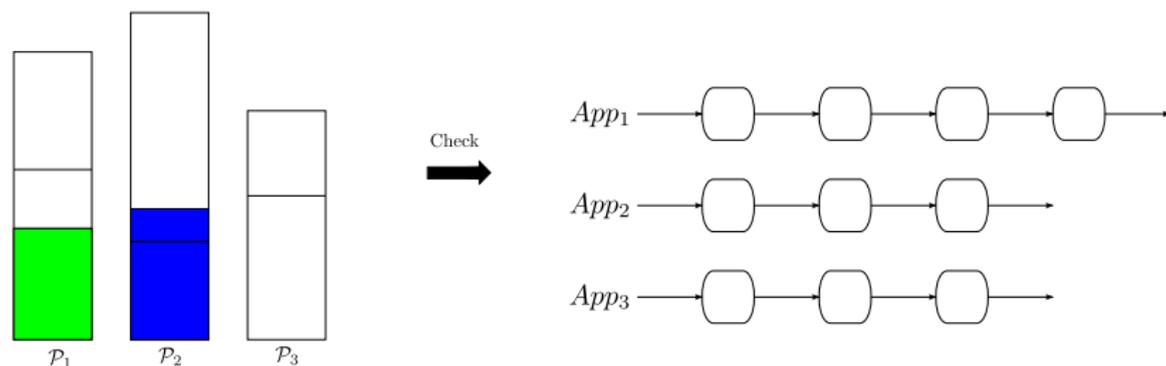
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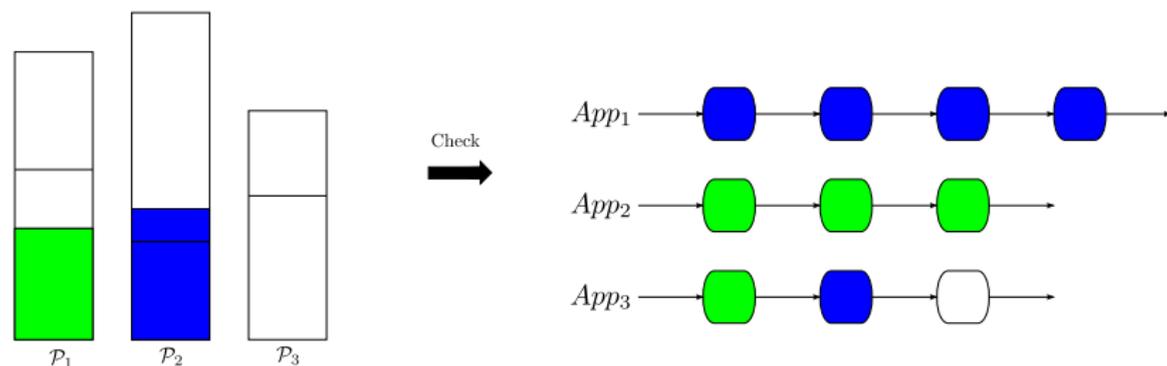
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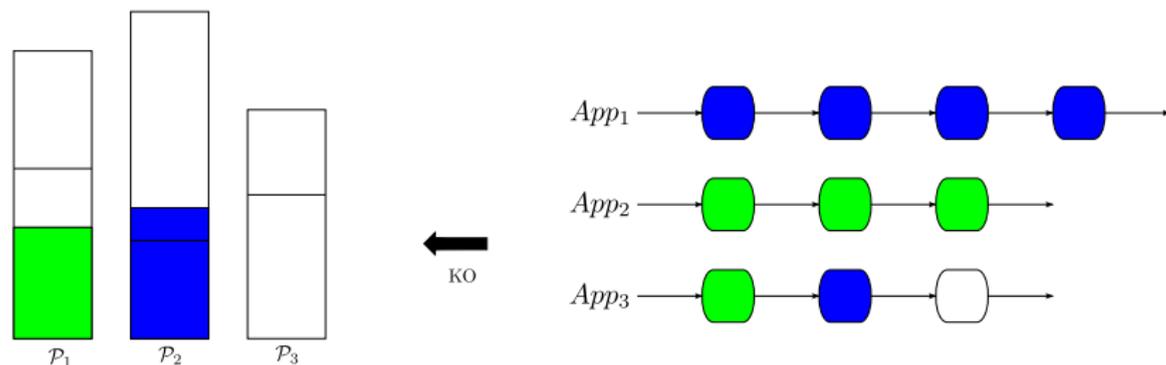
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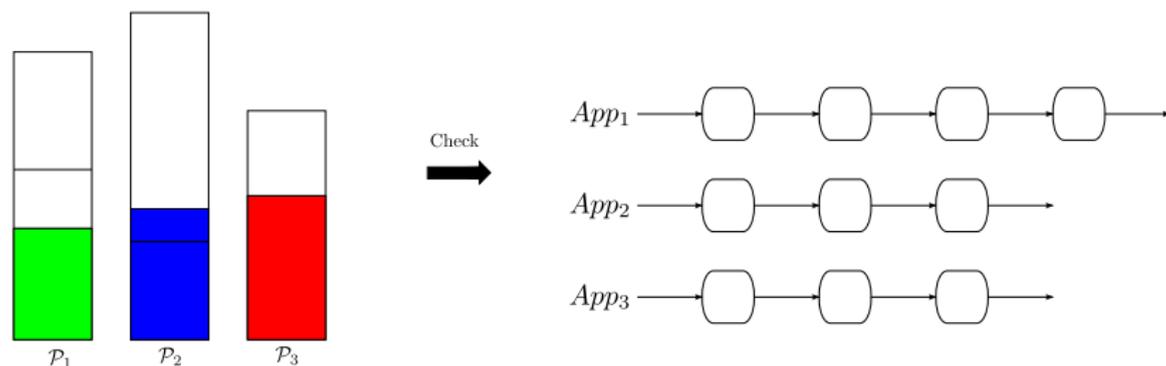
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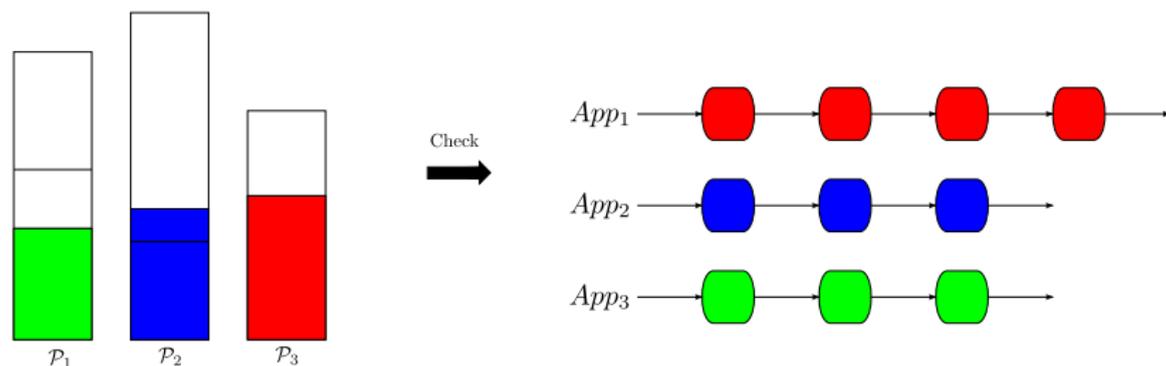
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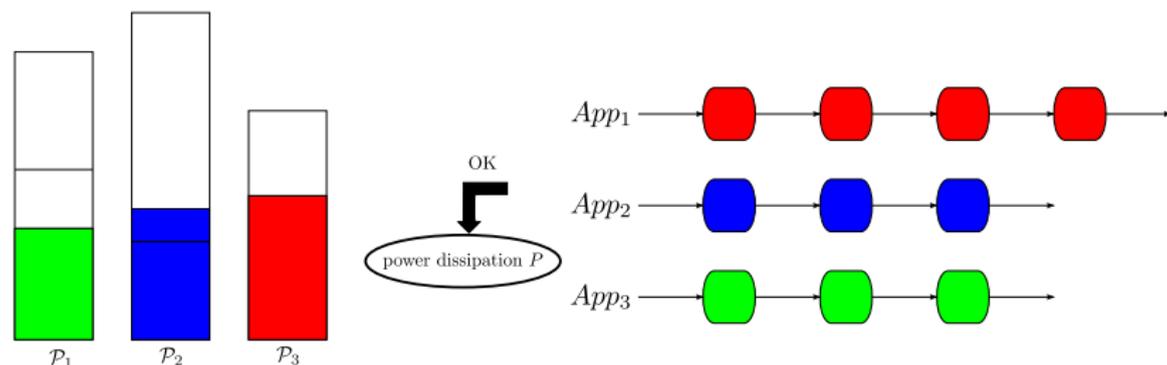
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Iterate the process: increase processor speeds



# Experimental plan

- **Integer linear program** to assess the absolute performance of the heuristics on small instances
- **Small instances**: two or three applications, around 15 stages per application, around 8 processors
- Execution time on 30 small instances: less than one second for all heuristics, one week for the ILP
- Each heuristic and the ILP: variant without sharing ("-n") and variant with sharing ("-r")
  - General behavior of heuristics
  - Impact of resource sharing
  - Scalability of heuristics

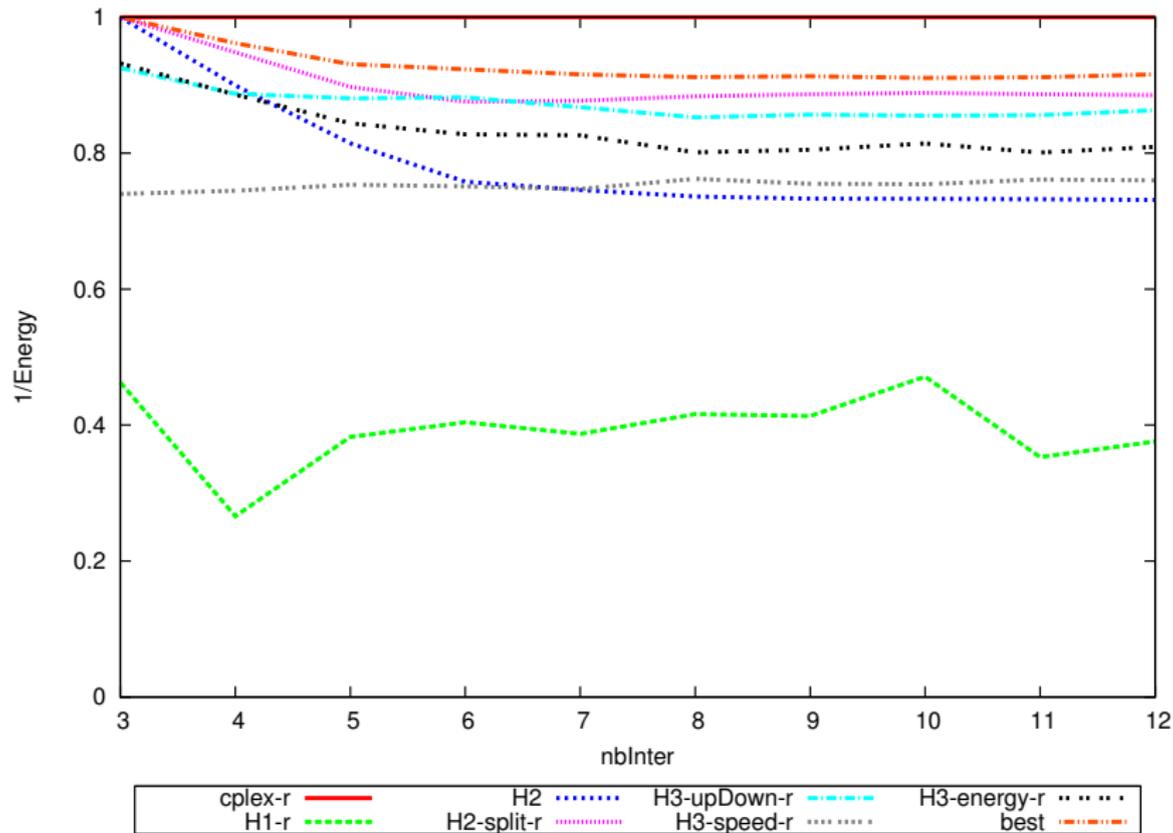
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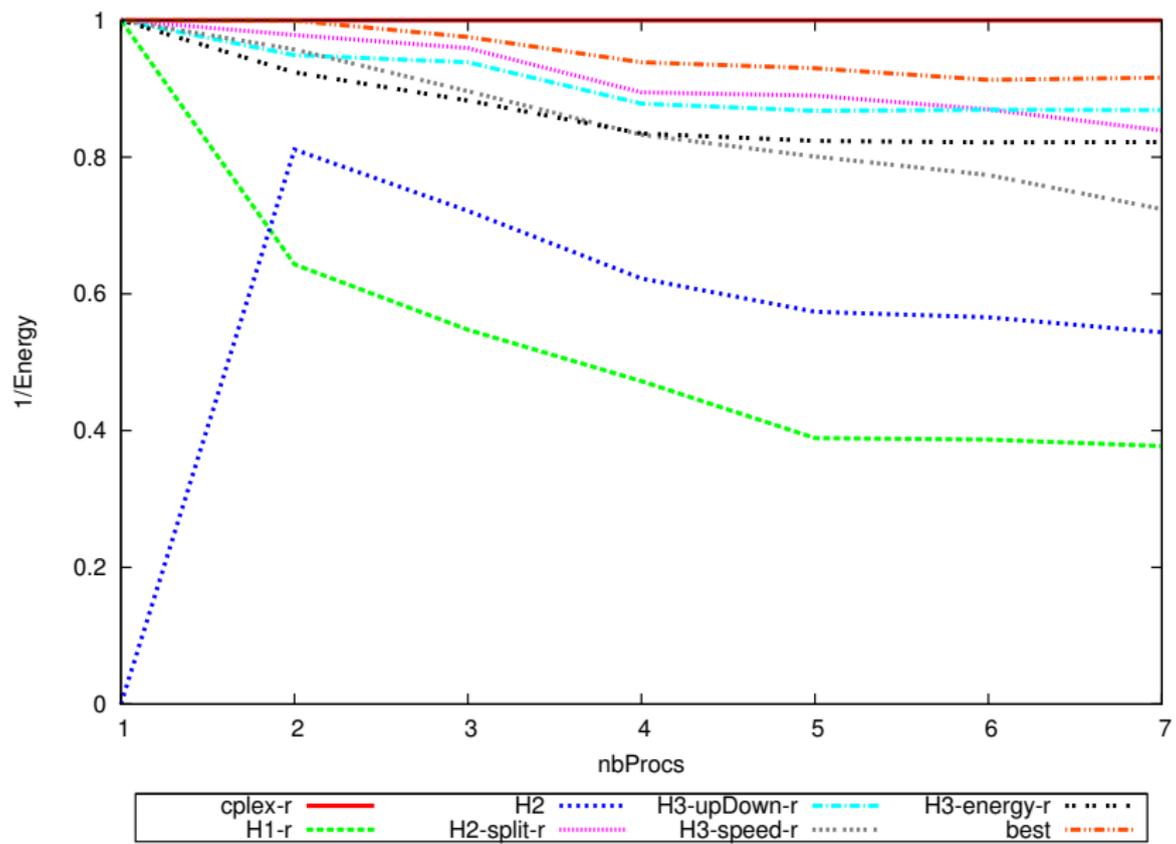
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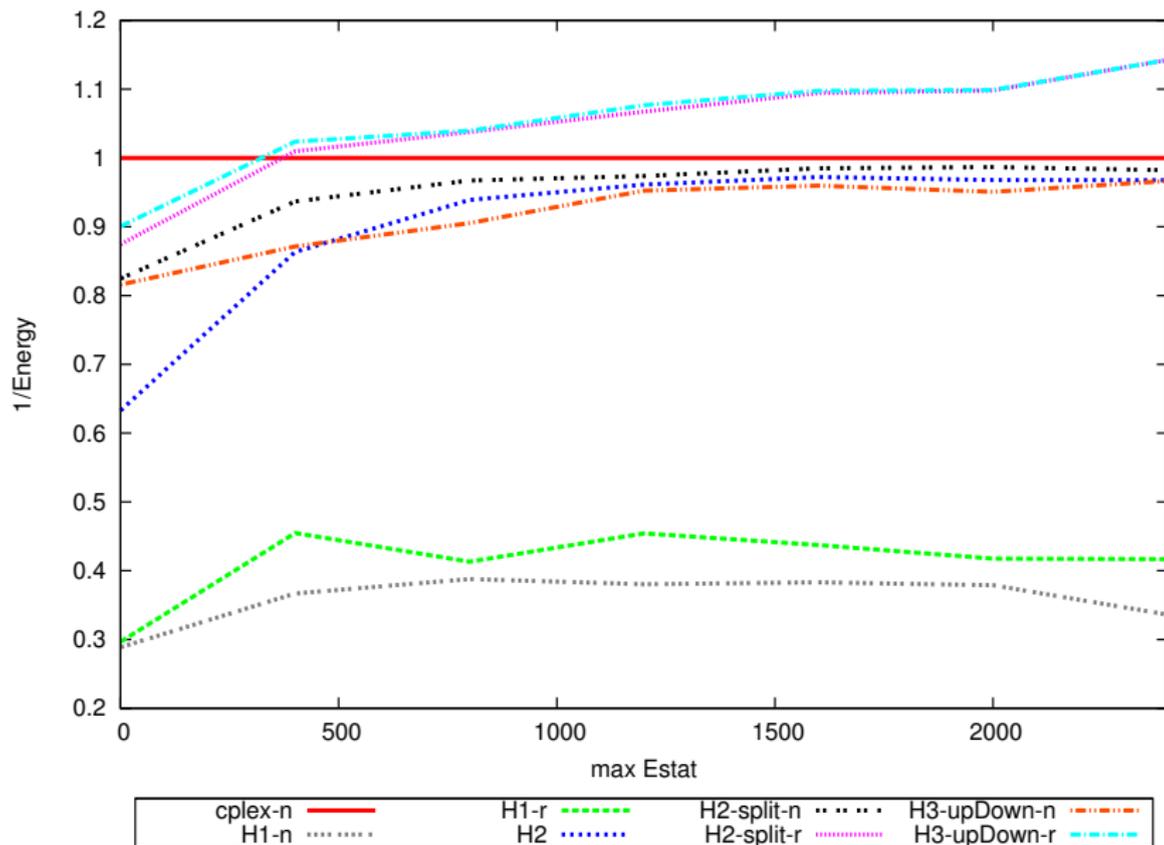
# Increasing latency



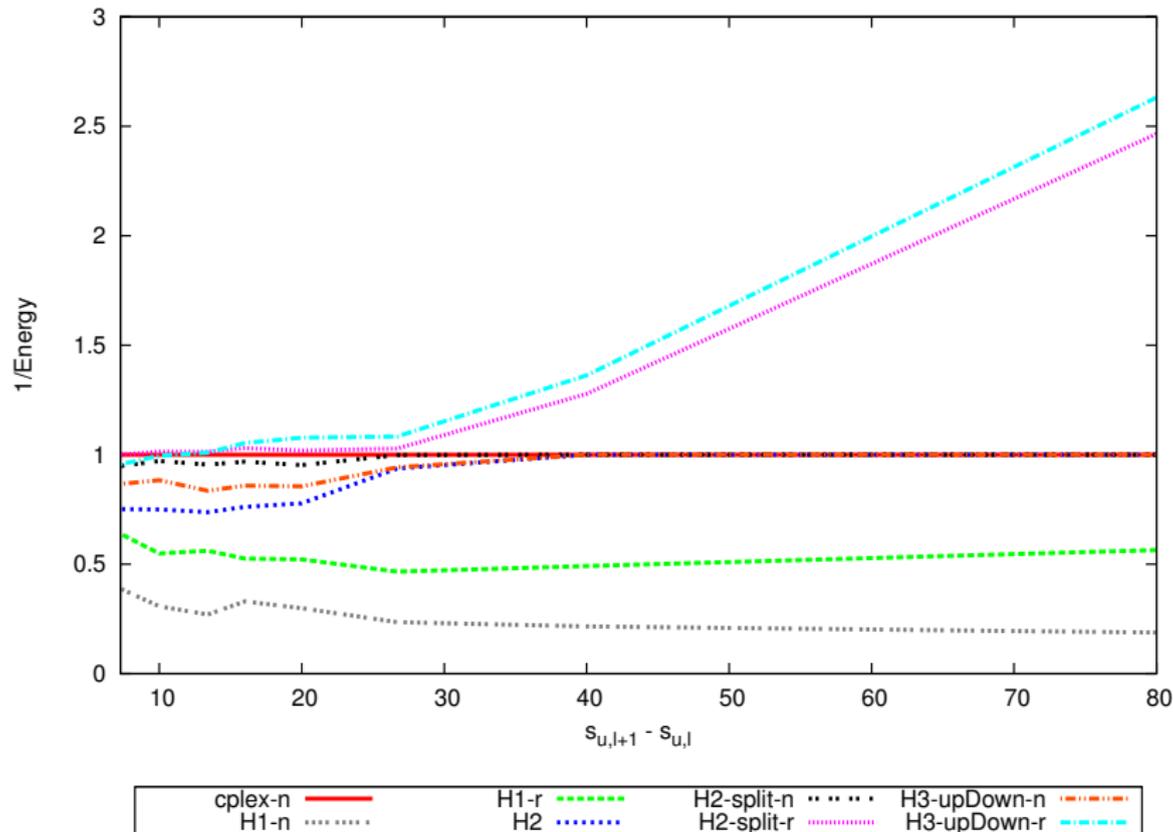
# Increasing number of processors



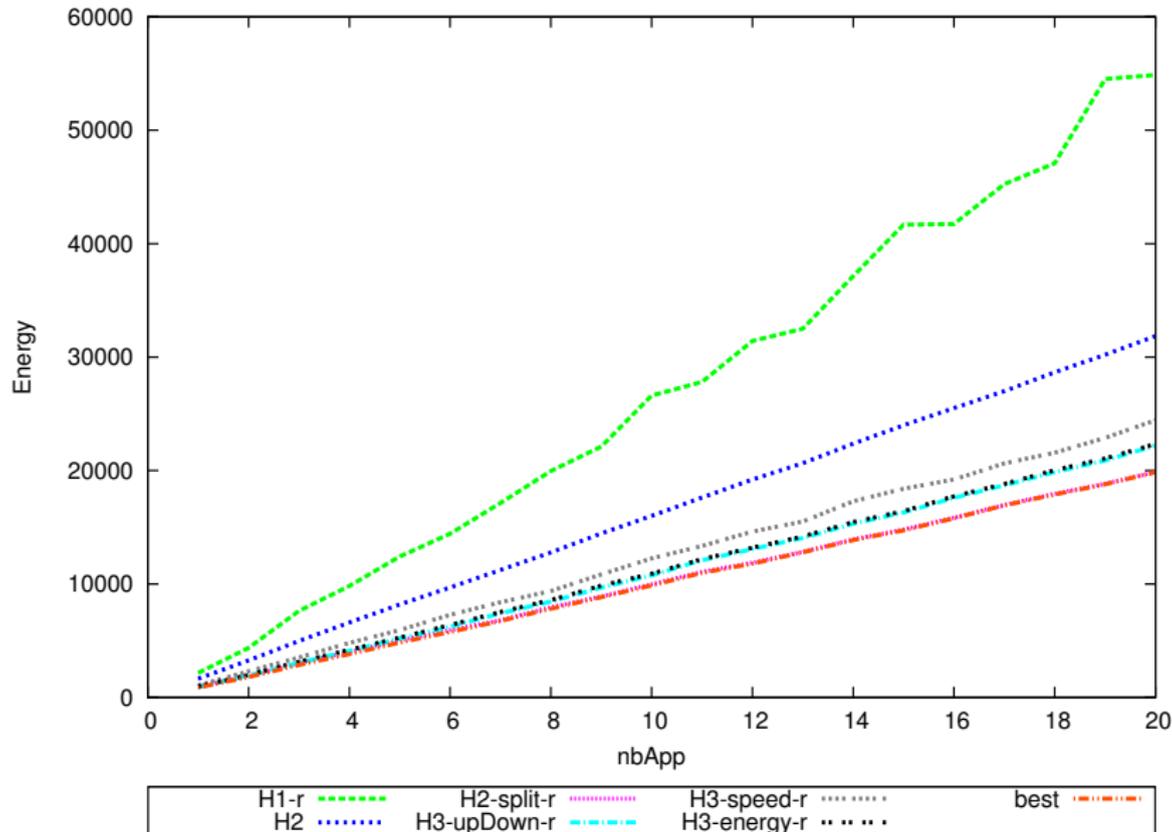
# Impact of static power



## Impact of mode distribution



# Scalability



# Summary of experiments

- **Efficient heuristics**: best heuristic always at 90% of the optimal solution on small instances
- Supremacy of H2-split-r, better in average, and gets even better when problem instances get larger
- H3 has smaller execution time (one second versus three minutes for 20 applications), ILP not usable in practice
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# Outline of the talk

- 1 Framework
  - Application and platform
  - Mapping rules
  - Metrics
- 2 Complexity results
  - Mono-criterion problems
  - Bi-criteria problems
  - Tri-criteria problems
  - With resource sharing
- 3 Experiments
  - Heuristics
  - Experiments
  - Summary
- 4 Conclusion

# Conclusion and future work

- Exhaustive complexity study
  - new polynomial algorithms
  - new NP-completeness proofs
  - impact of model on complexity (tri-criteria homogeneous)
- Experimental study
  - efficient heuristics
  - impact of resource reuse
- Current/future work
  - continuous speeds
  - approximation algorithms

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