Michaël Gabay Sofia Zaourar

Laboratoire G-SCOP LJK - Inria Grenoble France

Team J19 - Open source

July 10, 2012

Summary



Constraints

Costs



Our solution

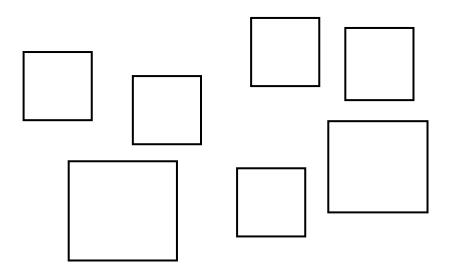
- Scheme
- Vector bin packing
- Another approach
- Local search



Results

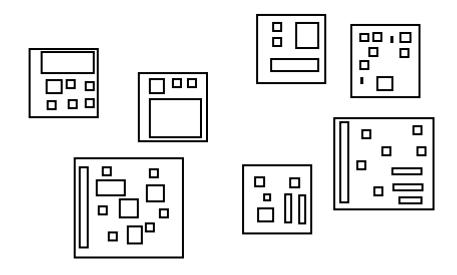
The problem

Overview



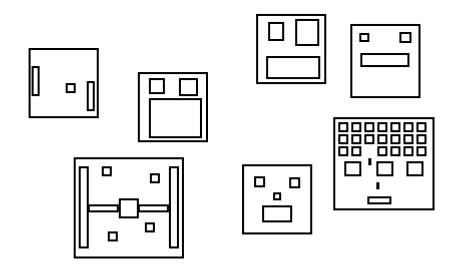
The problem

Overview



The problem

Overview



The problem

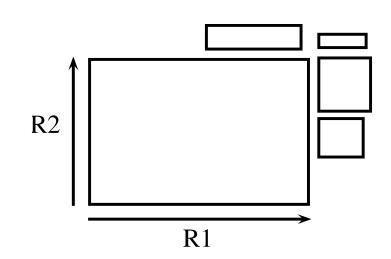
Constraints



- Capacity constraints and transient usage constraints
- Conflict constraints
- Spread constraints
- Dependency constraints

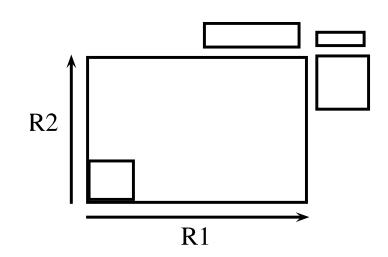
The problem

Constraints



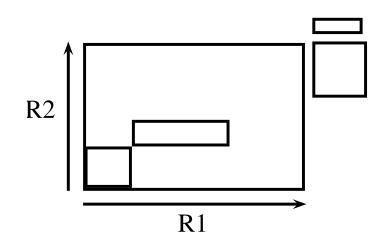
The problem

Constraints



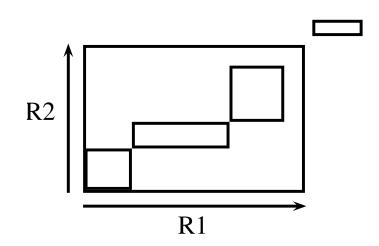
The problem

Constraints



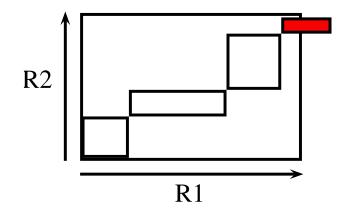
The problem

Constraints



The problem

Constraints

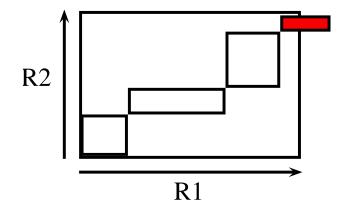


The problem

Constraints

Capacity constraints

Vector bin packing problem, Garey et al., 76

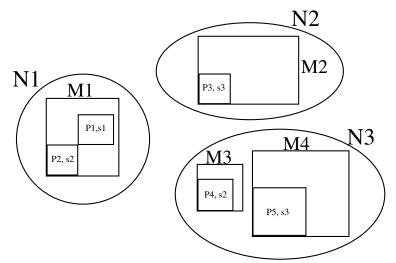


The problem

Constraints

Dependency constraints

 s_1 depends on s_2 :

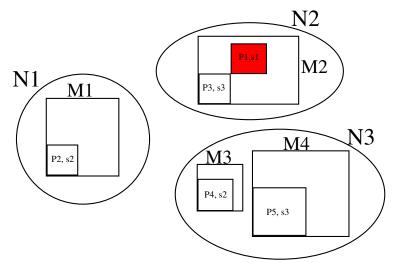


The problem

Constraints

Dependency constraints

 s_1 depends on s_2 :

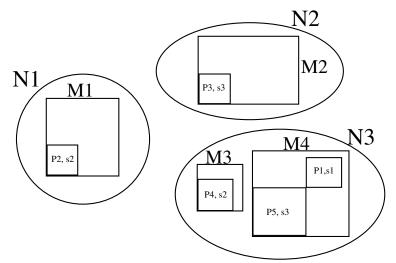


The problem

Constraints

Dependency constraints

 s_1 depends on s_2 :



6

The problem

Constraints

Dependency constraints

Idea : break up the problem by neighborhoods

 \Rightarrow smaller subproblems with no dependency constraints.

Solvable using integer programming \Rightarrow Matheuristic approach.

The problem

Costs

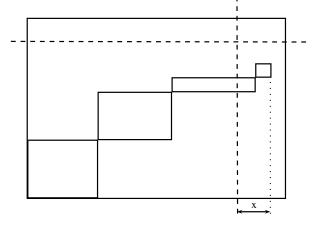


- Process / Service / Machine move costs
- Load costs
- Balance costs

The problem

Costs

Load costs

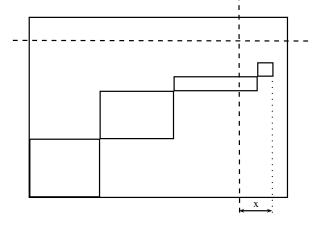


The problem

Costs

Load costs

Use load costs to guide search



Our solution

Scheme



GRASP = Greedy randomized adaptive search procedure

Optimization scheme:

- Create a solution using vector bin packing heuristic
- Optimize it with a local search



Our solution

Vector bin packing

Vector Bin Packing

 $\forall m \in \mathcal{M}, \forall p \in \mathcal{P}, \text{ volume } v:$

$$v(m) = \sum_{r \in \mathcal{R}} \frac{C(m,r)}{C(r)}, \quad v(p) = \sum_{r \in \mathcal{R}} \frac{R(p,r)}{R(r)}$$

Our solution

Vector bin packing

Vector Bin Packing

 $\forall m \in \mathcal{M}, \forall p \in \mathcal{P}, \text{ volume } v:$

$$v(m) = \sum_{r \in \mathcal{R}} \frac{C(m,r)}{C(r)}, \quad v(p) = \sum_{r \in \mathcal{R}} \frac{R(p,r)}{R(r)}$$

Greedy : First Fit Decreasing (FFD) FFD Bin-centric heuristic (Panigrahy et al., 2011):

- Sort machines by increasing v(m), processes by decreasing v(p)
- Pop the smallest remaining machine m
- While some processes fit into *m*, place the largest remaining one



Our solution

Vector bin packing

Vector Bin Packing

 $\forall m \in \mathcal{M}, \forall p \in \mathcal{P}, \text{ volume } v:$

$$v(m) = \sum_{r \in \mathcal{R}} \frac{C(m,r)}{C(r)}, \quad v(p) = \sum_{r \in \mathcal{R}} \frac{R(p,r)}{R(r)}$$

Greedy : First Fit (FF)

FF Bin-Balancing (item-centric) heuristic:

- Sort machines by increasing v(m), processes by decreasing v(p), i = 1
- Pop the largest remaining process p
- Solution While some machine can host p, place p on the smallest $j \ge i + 1$ (with a cyclic order), i = j



Our solution

Vector bin packing

Vector Bin Packing

 $\forall m \in \mathcal{M}, \forall p \in \mathcal{P}, \text{ volume } v:$

$$v(m) = \sum_{r \in \mathcal{R}} \frac{C(m,r)}{C(r)}, \quad v(p) = \sum_{r \in \mathcal{R}} \frac{R(p,r)}{R(r)}$$

Greedy Random

FF Mixed orderings heuristic:

- Sort machines by increasing v(m), processes by decreasing v(p), or random sort both
- Run FFD Bin-centric, run FF Bin-Balancing
- Goto

Our solution

Vector bin packing

Vector Bin Packing

- Capacity and transient usage constraints
- Conflict constraints
- Spread constraints
- Dependency constraints

Our solution

Vector bin packing

Vector Bin Packing

- Capacity and transient usage constraints \Rightarrow YES
- Conflict constraints ⇒ YES
- Spread constraints
- Dependency constraints

Our solution

Vector bin packing

Vector Bin Packing

- Capacity and transient usage constraints \Rightarrow YES
- Conflict constraints ⇒ YES
- Spread constraints
- Dependency constraints ⇒ YES Assign all processes of a neighborhood to the same neighborhood (possibly not the original one)

Our solution

Vector bin packing

Vector Bin Packing

- Capacity and transient usage constraints \Rightarrow YES
- Conflict constraints ⇒ YES
- Spread constraints ⇒ **REPAIR**
- Dependency constraints ⇒ YES Assign all processes of a neighborhood to the same neighborhood (possibly not the original one)

Our solution

Vector bin packing



- Use FF Mixed orderings to get a feasible solution
- 2 Local search on the solution
- Goto

Our solution

Vector bin packing

Problems with VBP

VBP heuristics are very fast but...

- Solutions' costs are too high
- Not so interesting when processes do not change neighborhood,... nor when they all change...
- 10 B instances, 6 feasible,... all violate spread constraints
- Repairing may be too difficult / time consuming

Our solution

Another approach

A matheuristic

ldea:

- For each neighborhood, solve the assignment problem using an IP (smaller subproblem with no dependency constraints)
- Optimize using local search

Goto

(implemented using Coin-Osi/Clp/Cbc)

Our solution

Another approach

A matheuristic: Problems

Our solution

Another approach

A matheuristic: Problems

Subproblems are still too big

Our solution

Another approach

A matheuristic: Problems

Subproblems are still too big

Solution:

Divide into smaller subproblems Randomly / Guided by load cost (most expensive machines with cheapest w.r.t. load costs)

Our solution

Another approach

A matheuristic: Problems

Tune size parameters:

Compromise between time consumption and feasibility

Our solution

Another approach

A matheuristic: Problems

Tune size parameters:

Compromise between time consumption and feasibility

Solution:

- Set maximum number of machines and processes to be considered at once
- Set maximum number of nodes

Our solution

Another approach

A matheuristic: Problems

Pure local search is better

Our solution

Another approach

A matheuristic: Problems

Pure local search is better (and deadline approaches)

Our solution

Another approach

A matheuristic: Problems

Pure local search is better (and deadline approaches)

Solution:

Forget about the Matheuristic :-(

Our solution

Local search



Two simple moves:

- Move *p* from *m*₁ to *m*₂
- Swap p_1 and p_2 on m_1 and m_2

Efficient Structures + randomization

Our solution

Local search



Two simple moves:

- Move p from m₁ to m₂
- Swap p_1 and p_2 on m_1 and m_2

Efficient Structures + randomization

Hill Climbing: if a move is feasible and decreases the total cost, do it!

Our solution

Local search



Two simple moves:

- Move p from m₁ to m₂
- Swap p_1 and p_2 on m_1 and m_2

Efficient Structures + randomization

Hill Climbing: if a move is feasible and decreases the total cost, do it!

Guided first step: Move processes from the machines with the highest load costs to cheaper machines

Our solution

Local search



Two simple moves:

- Move p from m₁ to m₂
- Swap p_1 and p_2 on m_1 and m_2

Efficient Structures + randomization

Hill Climbing: if a move is feasible and decreases the total cost, do it!

Guided first step: Move processes from the machines with the highest load costs to cheaper machines

Blocked ? Restart from a previous solution with a new seed

Final solution

Solution



Local search

Two parallel local search with different strategies

Final solution

Solution



Local search

Two parallel local search with different strategies

Open source under LGPL v3 license, available at:

```
https://github.com/TeamJ19ROADEF2012
```

(Includes VBP and the matheuristic)

Final solution

Results

Intel Core 2 duo P9400, RAM 4 Go

Instance	Cost	Real time	CPU time
B 1	3 609 228 327	4m58.5s	9m53.4s
B 2	1 017 459 868	4m58.5s	9m54.4s
B 3	170 139 317	4m58.5s	9m53.4s
B 4	4 677 960 720	4m58.5s	9m54.2s
B 5	930 031 137	4m58.5s	9m53.8s
B 6	9 525 886 513	4m58.5s	9m53.7s
B 7	14 911 018 492	4m58.5s	9m52.4s
B 8	1 217 854 951	4m58.5s	9m53.0s
B 9	15 886 884 119	4m58.5s	9m54.1s
B 10	18 391 528 354	4m58.5s	9m51.0s

Final solution

Results

Questions ?