

# Modeling Patterns for LocalSolver

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# Who we are



Large industrial group with businesses in construction, telecom, media

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Operation Research subsidiary of the Bouygues group

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**LocalSolver**

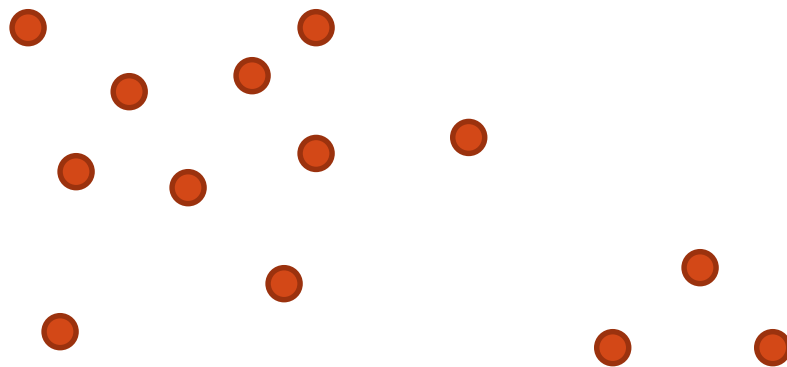
Flagship product of Innovation 24

*[www.localsolver.com](http://www.localsolver.com)*



# LocalSolver in one slide

*Select a set  $S$  of  $P$  cities among  $N$   
Minimizing the sum of distances  
from each city to the closest city  
in  $S$*



```
function model() {  
  x[1..N] <- bool();  
  constraint sum[i in 1..N] (x[i]) == P;  
  
  minDistance[i in 1..N] <- min[j in 1..N] (x[j] ? distance[i][j] : +inf);  
  minimize sum[i in 1..N] (minDistance[i]);  
}
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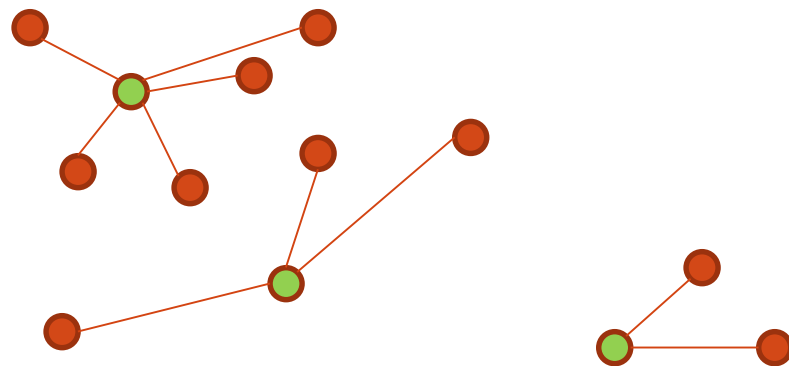
## Results on the OR Library

- 28 optimal solutions on the 40 instances of the OR Lib
- an average gap of 0.6%
- with 1 minute per instance



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# LocalSolver in one slide

An hybrid math  
programming solver

For large-scale mixed-variable  
non-convex optimization problems

Providing high-quality solutions  
in short running time  
without any tuning

# Outline

- LocalSolver
- Modeling Patterns for Local Search ?
- Six Modeling Patterns



# Modeling Patterns for LocalSolver

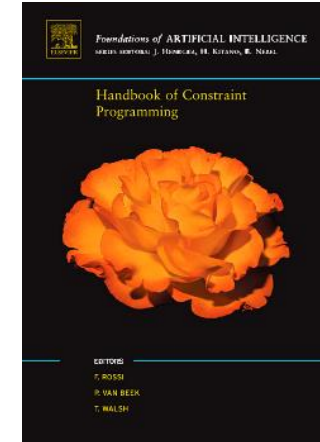
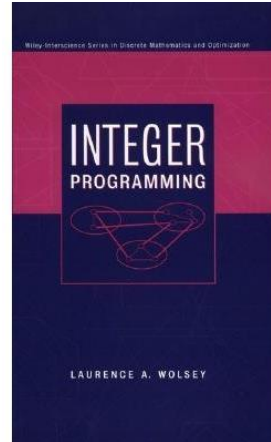
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Why?



# Modeling patterns ?

A classic topic in MIP or CP



Very little literature on modeling for Local Search...

...**because of** the absence of model-and-run solver

→ models and algorithms were designed together and not always clearly separated





# Modeling Pattern #1

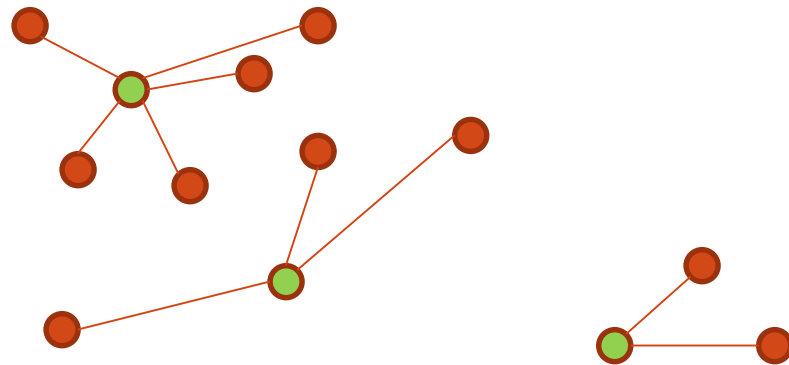
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Choose the right set of decision variables



# Choose the right set of decision variables

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# Modeling Pattern #2

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Precompute what can be precomputed



# Precompute what can be precomputed

Document processing : dans un tableau une case de texte a plusieurs configurations *hauteur x largeur* possibles.

LocalSolver : mathematical programming by local search

29 x 82

LocalSolver : mathematical programming by local search

34x 61

LocalSolver : mathematical programming by local search

45x 43

Comment choisir la configurations de chaque case de façon à minimiser la hauteur du tableau (sa largeur étant limitée) ?




UNIVERSITY of LIMERICK

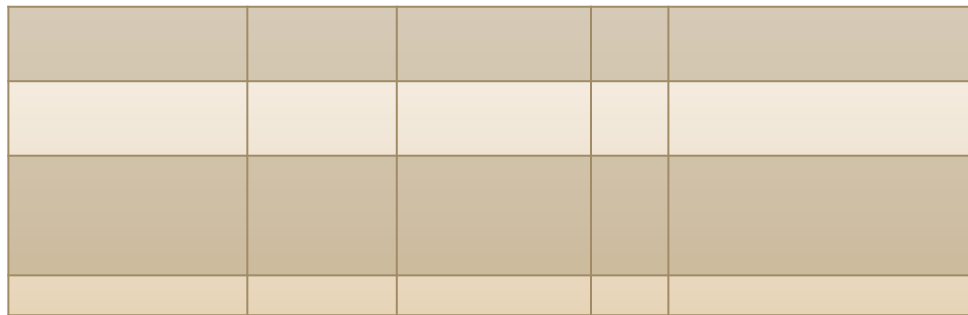
OLLSCOIL LUIMNIGH

# Precompute what can be precomputed

Première modélisation : 1 variable de décision par configuration (largeur, hauteur) possible pour chaque cellule

Formulation étendue :

- On remarque qu'à partir de la largeur d'une colonne on peut déterminer la hauteur minimum de chacune de ses cellules.
- 1 variable de décision par largeur possible pour chaque colonne
- Conséquence : en changeant une variable de décision, LocalSolver va changer la hauteur et la largeur de toutes les cellules dans la colonne




-> R. Megel (Roadef 2013).

Modélisations LocalSolver de type « génération de colonnes ».

# Modeling Pattern #3

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Do not limit yourself to linear operators



# Do not limit yourself to linear operators

## TRAVELING SALESMAN PROBLEM

MIP approach:  $X_{ij}=1$  if city  $j$  is after city  $i$  in the tour

- Matching constraints  $\sum_j X_{ij} = 1$  and  $\sum_i X_{ij} = 1$
- Plus an exponential number of subtour elimination constraints
- Minimize  $\sum_{ij} c_{ij} X_{ij}$

Polynomial non-linear model:  $X_{ik}=1$  if city  $i$  is in position  $k$  in the tour

- Matching constraints  $\sum_k X_{ik} = 1$  and  $\sum_i X_{ik} = 1$
- $Y_k \leftarrow \sum_i i X_{ik}$  the index of the  $k^{\text{th}}$  city of the tour
- Minimize  $\sum_k c_{[Y_k, Y_{k+1}]}$

↑  
“at” operator

TSP Lib: average gap  
after 10mn = 2.6%



# Modeling Pattern #4

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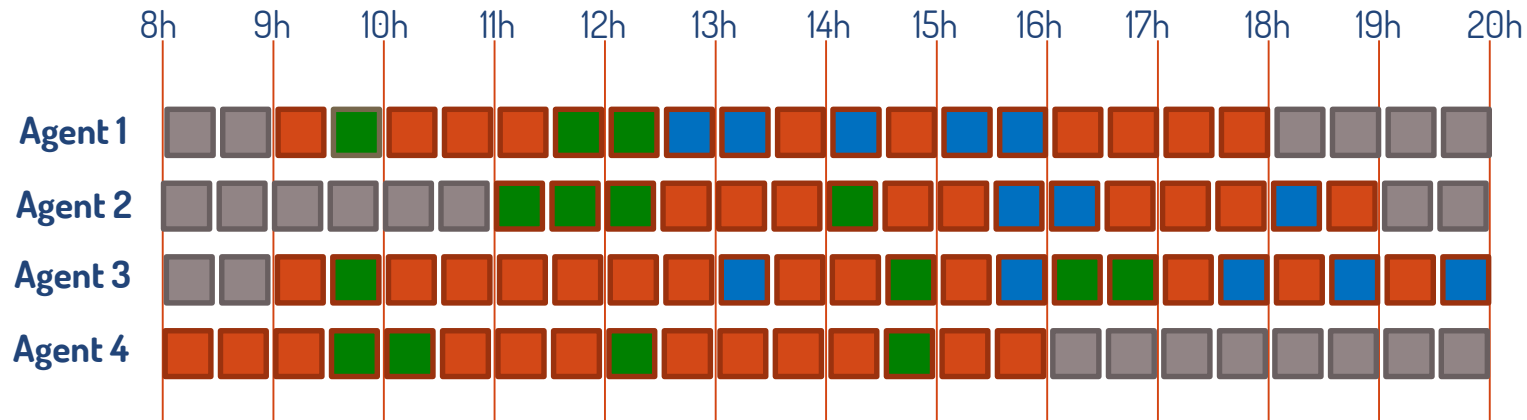
Separate commands and effects





# Separate commands and effects

## Multi-skill workforce scheduling



## Candidate model

$\text{Skill}_{atk} = 1 \Leftrightarrow$  agent  $a$  works on skill  $k$  at timestep  $t$

*Constraint*  $\text{SUM}_k (\text{Skill}_{atk}) \leq 1$

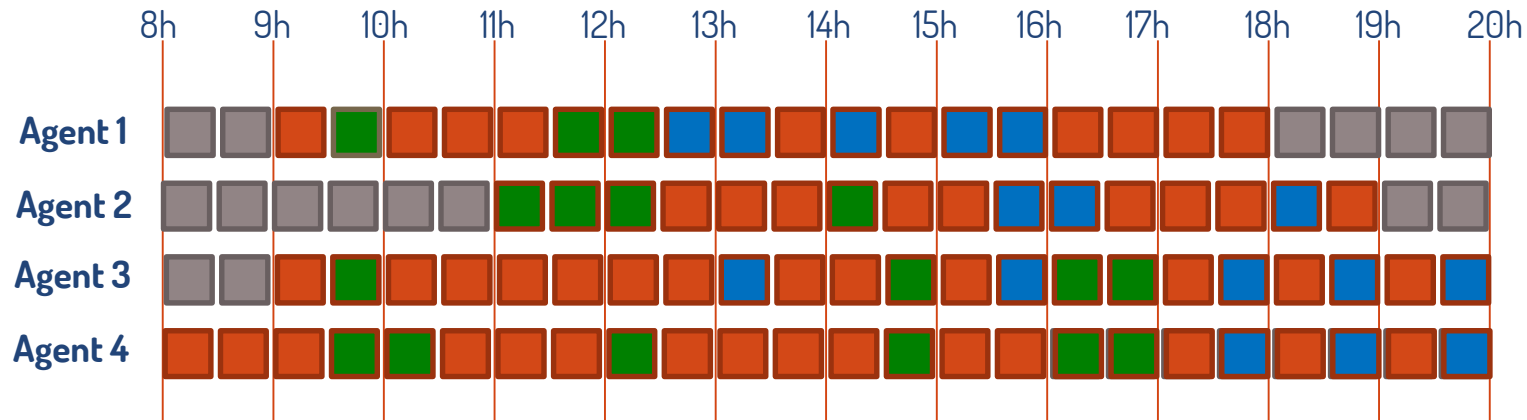
*Constraint*  $\text{OR}_k (\text{Skill}_{atk}) == (t \in [\text{Start}_a, \text{End}_a[ )$

Problem: any change of  $\text{Start}_a$  will be rejected unless skills are updated for all impacted timesteps



# Separate commands and effects

## Multi-skill workforce scheduling



## Alternative model

$\text{SkillReady}_{atk} = 1 \Leftrightarrow$  agent  $a$  will work on skill  $k$  at timestep  $t$  **if present**

*Constraint*  $\text{SUM}_k (\text{SkillReady}_{atk}) == 1$

$\text{Skill}_{atk} \leftarrow \text{AND}(\text{SkillReady}_{atk}, t \in [\text{Start}_a, \text{End}_a])$

Now we have no constraint between skills and worked hours  
-> for any change of  $\text{Start}_a$  skills are automatically updated



# Separate commands and effects

## Similar case: Unit Commitment Problems

- A generator is active or not, but when active the production is in  $[P_{\min}, P_{\max}]$
- Better modeled without any constraint

$$\text{ProdReady}_{\text{gt}} \leftarrow \text{float}(P_{\min}, P_{\max})$$
$$\text{Active}_{\text{gt}} \leftarrow \text{bool}()$$
$$\text{Prod}_{\text{gt}} \leftarrow \text{Active}_{\text{gt}} \times \text{ProdReady}_{\text{gt}}$$


# Modeling Pattern #5

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Invert constraints and objectives ?



# Invert constraints and objectives

***Clément Pajean***

*Modèle LocalSolver d'ordonnancement d'une machine unique  
sous contraintes de Bin Packing*

Vendredi 28

14h

Bât B TD 35



# Modeling Pattern #6

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Use dominance properties



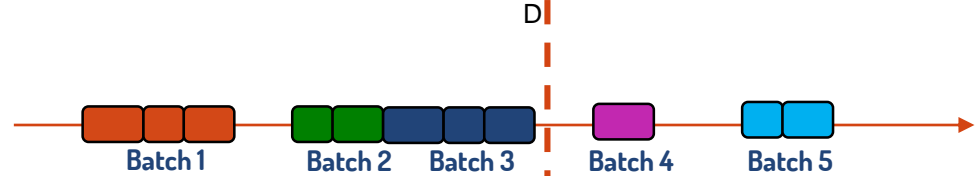
# Use dominance properties

Batch scheduling for  $N$  jobs having the same due date  $D$ .

- Completion time of each job will be that of the batch selected for this job
- Linear late or early cost ( $\alpha_k \beta_k$ )

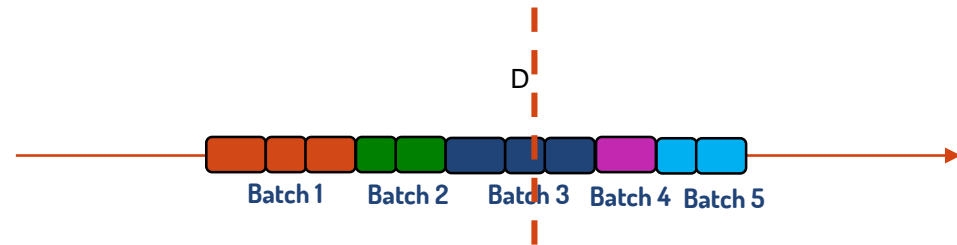
We can **minimize a minimum**

As if starting date was automatically adjusted after each move



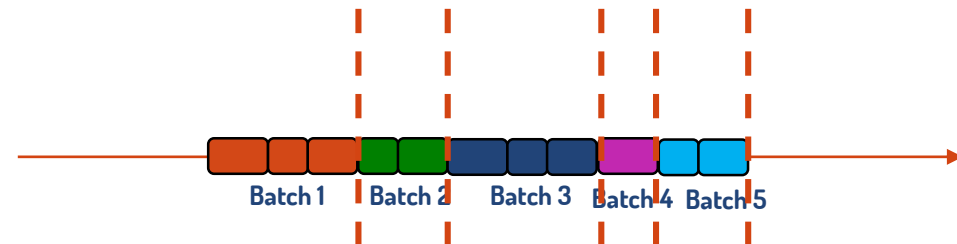
**Basic Model**

Date variable for each batch  
+ assignment of jobs to batches  
+ Precedence constraints



**No idle time**

Only one starting date variable  
+ assignment of jobs to batches



**Optimal start will position due date at the end of a batch**

No start date variable  
+ assignment of jobs to batches  
+ penalty[k] if due date at the end of batch k  
Minimize  $\min[k \text{ in } 1..5](\text{penalty}[k])$

# Summary

1. Choose the right set of decision variables
2. Precompute what can be precomputed
3. Do not limit yourself to linear operators
4. Separate commands and effects
5. Invert constraints and objectives ?
6. Use dominance properties





# Modeling Pattern #7

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Your turn!

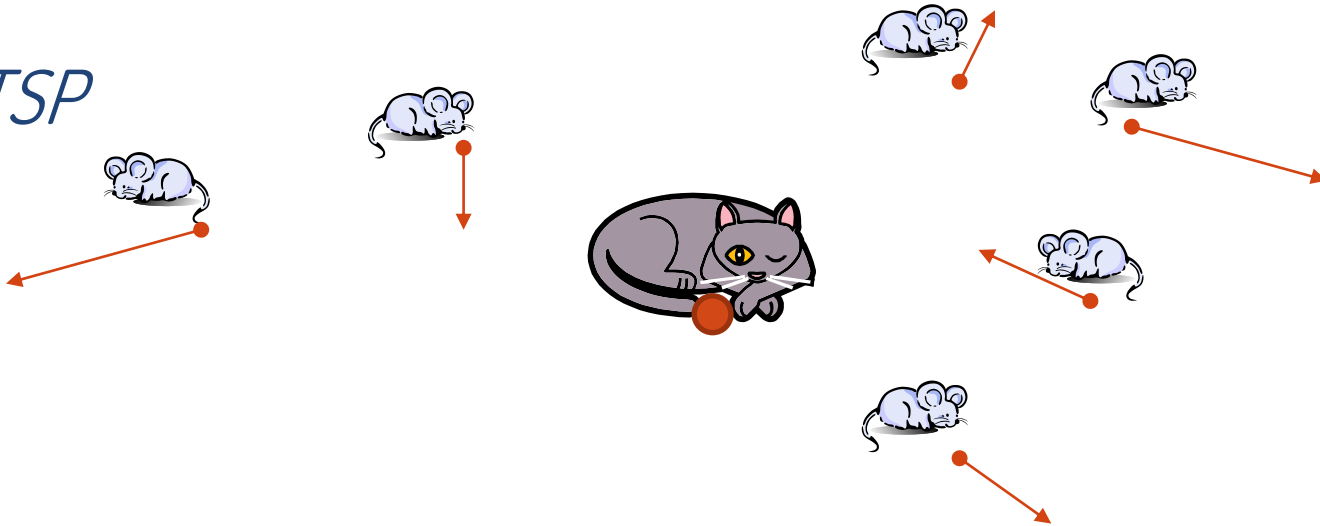


# Why solving a TSP with LocalSolver ?

$$\text{Time}[A, B, T] = \frac{2(\alpha_x^2 + \alpha_y^2)}{-2(\alpha_x\beta_x + \alpha_y\beta_y) + \sqrt{4(\alpha_x\beta_x + \alpha_y\beta_y)^2 - 4(\alpha_x^2 + \alpha_y^2)(\beta_x^2 + \beta_y^2 - V^2)}} + T$$

With  $\alpha_x, \beta_x, \alpha_y, \beta_y$  function of A,B and T

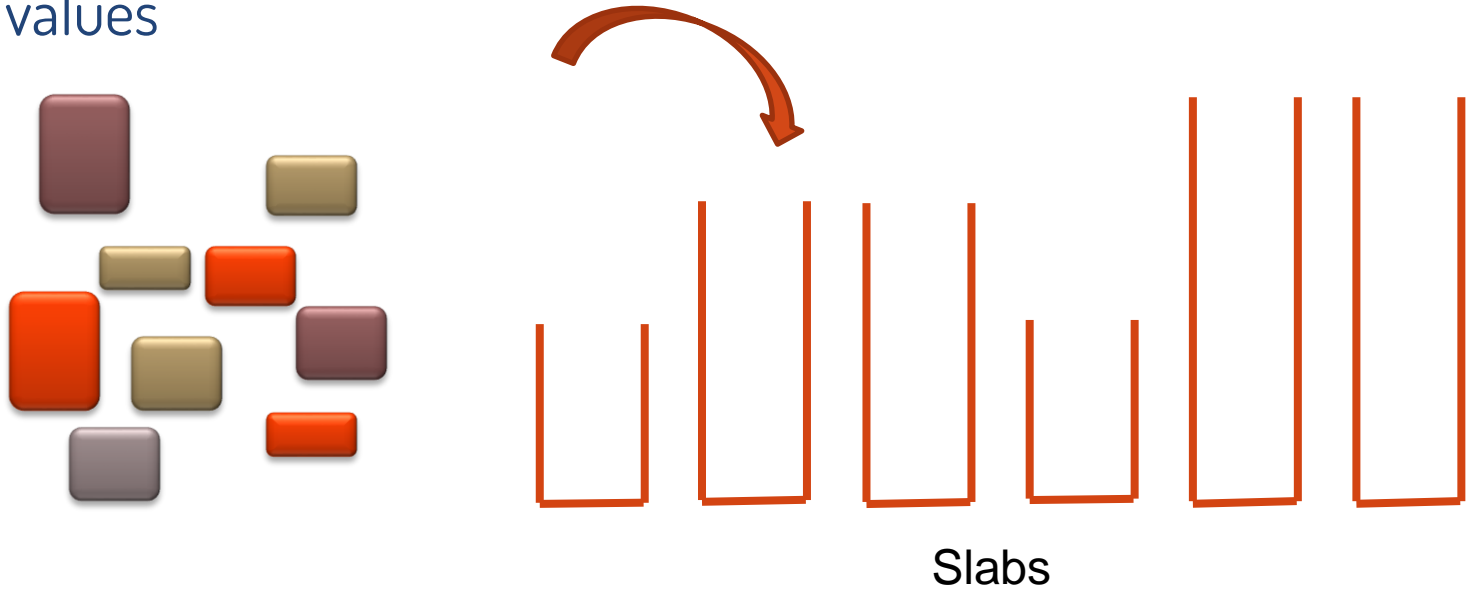
*Kinetic TSP*



# Choose the right set of decision variables

Industrial « Bin-packing »

Assignment of steel orders to « slabs » whose capacity can take only 5 different values



Model

$X_{ij} = 1 \Leftrightarrow$  Order  $i$  is assigned to slab  $j$

$Y_{jk} = 1 \Leftrightarrow$  Slab  $j$  takes capacity  $k$

Minimize total size of slabs

# Choose the right set of decision variables

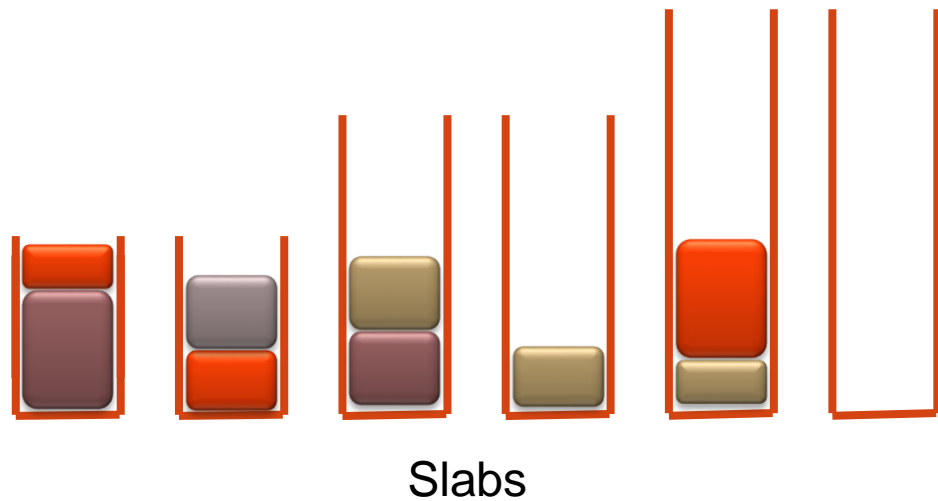
Industrial « Bin-packing »

Assignment of steel orders to « slabs » whose capacity can take only 5 different values

Change

Exchange

Ejection chain



In a good model:

- when a value can be computed from others it is defined with operator `<-` (it is an **intermediate** variable )
- moving from a feasible solution to another feasible solution only requires modifying a small number of **decision** variables.

# Choose the right set of decision variables

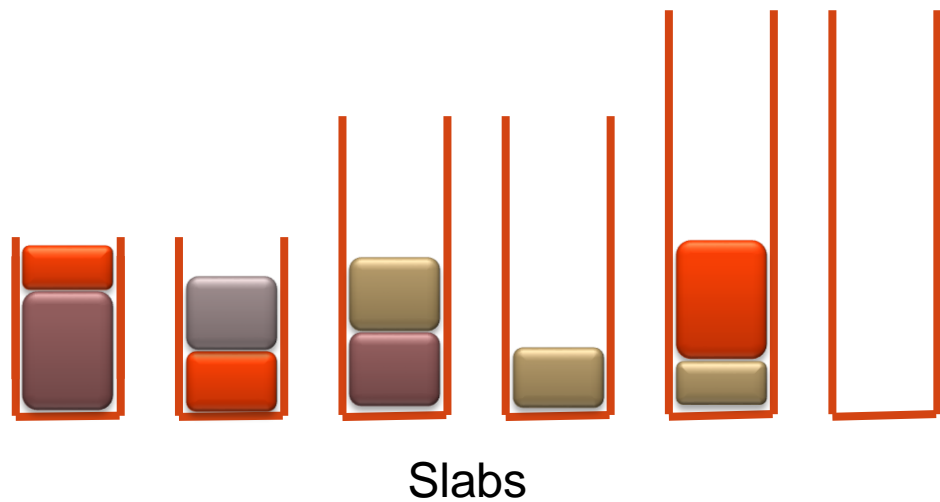
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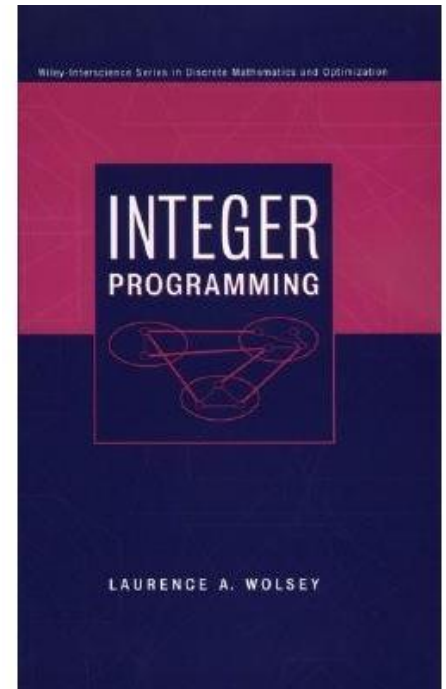
$Capa_k \leftarrow MinCapa[content_k]$

↑  
“at” operator



# Modeling patterns for Mixed-Integer Programming

- MIP requires linearizing non linear structures of the problem
- The polyhedron should be kept as close as possible to the convex hull -> valid inequalities, cuts, and so on
- Symmetries should be avoided (or not...)



# Modeling patterns for constraint programming

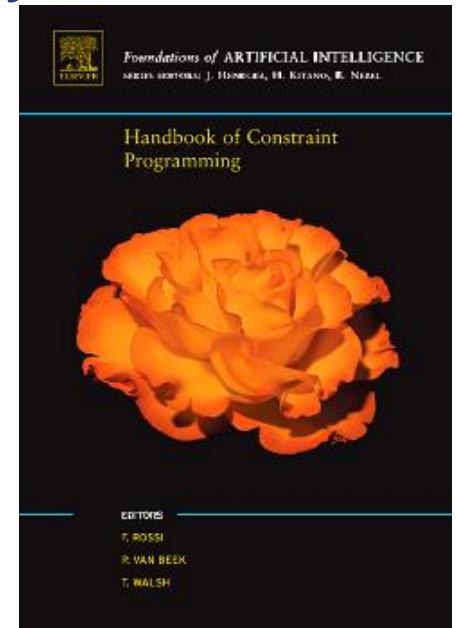
Choice of variables (integer, set variables, continuous...)

Choice of (global) constraints

Redundant constraints

Double point of view (with channeling constraints)

And so on



# Modeling patterns for Local Search ?

Very little literature on this topic...

...**because of** the absence of model-and-run solver

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