

# A Branch-and-Price Algorithm for the Bin Packing Problem with Conflicts

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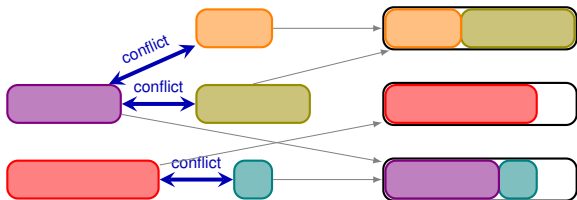
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# Problem definition

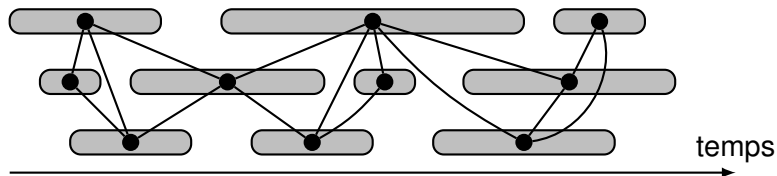
- ▶ Given
  - ▶ an infinite number of bins of size  $W$ ,
  - ▶ a set  $N = \{1, \dots, n\}$  of items  $i$  of sizes  $w_i$ ,
  - ▶ a graph  $G = (N, E)$  of conflicts between items.
- ▶ Pack the items into a minimum number of bins.
- ▶ Two items in conflict cannot be in the same bin.



# Motivations

## Mutual exclusion scheduling (generalisation)

(Baker & Coffman, 96), (Gardi, 05)



- ▶ Total length of tasks assigned to a person  $\leq W$
- ▶ Overlapping tasks cannot be assigned to the same person : conflict graph is **interval**

## Other applications

- ▶ Examination scheduling (Laporte & Desroches, 84)
- ▶ Parallel computation (Jansen, 99), (Beaumont et al., 08)
- ▶ Product delivery (Christofides et al., 79)

# IP formulation

Variables :

- ▶  $y_k = 1$  if bin  $k$  is used, otherwise  $y_k = 0$
- ▶  $x_{ik} = 1$  if item  $i$  is put to bin  $k$ , otherwise  $x_{ik} = 0$

$$\begin{aligned} \min \quad & \sum_{k \in K} y_k \\ & \sum_{k \in K} x_{ik} = 1, \quad i \in N, \\ & \sum_{i=1}^n w_i x_{ik} \leq W y_k, \quad k \in K, \\ & x_{ik} + x_{jk} \leq y_k, \quad (i, j) \in E, k \in K, \\ & y_k \in \{0, 1\}, \quad k \in K, \\ & x_{ik} \in \{0, 1\}, \quad i \in N, k \in K. \end{aligned}$$

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# Set covering formulation

$\mathcal{B}$  = set of valid single bin packings

$\lambda_b = 1$  if subset  $b \in \mathcal{B}$  of items occupies a bin

## Formulation

$$\begin{aligned} \min \quad & \sum_{b \in \mathcal{B}} \lambda_b \\ & \sum_{b \in \mathcal{B}: i \in b} \lambda_b \geq 1, \quad i \in N, \\ & \lambda_b \in \{0, 1\}, \quad b \in \mathcal{B}. \end{aligned}$$

## Pricing problem

$$\begin{aligned} \max \quad & \sum_{i \in N} \pi_i x_i \\ & \sum_{i \in N} w_i x_i \leq W, \\ & x_i + x_j \leq 1, (i, j) \in E, \\ & x_i \in \{0, 1\}, i \in N. \end{aligned}$$

(Knapsack with conflicts)

# Knapsack with conflicts : existent approaches

## Arbitrary conflict graphs

NP-hard,  $\approx 100$  seconds for solving exactly a 1000-items instance (Hifi, Michrafy, 07) — **slow**.

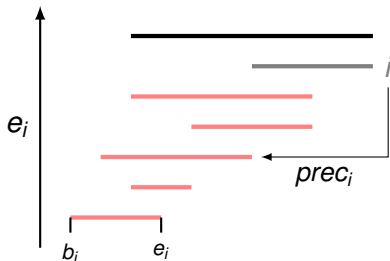
## Structured conflict graphs

- ▶ **Trees** and **chordal** graphs : dynamic programming algorithm with complexity  $O(nW^2)$  (Pferschy, Shauer, 09) — **slow**.
- ▶ **Interval** graphs : these instances we can solve **fast**.



## Interval conflict graphs : dynamic programming

$P(i, w)$  — solution value of the subproblem with the first  $i$  items and bin size  $w$



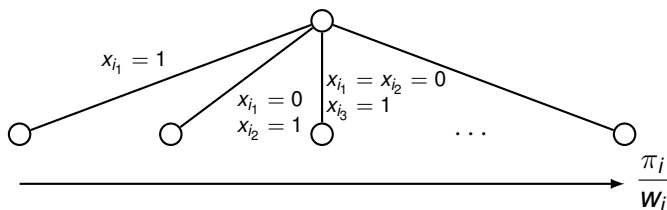
$$P(i, w) = \max \left\{ P(\text{prec}_i, w - w_i) + p_i, P(i - 1, w) \right\}$$

Complexity :  $O(nW)$ , same as for the knapsack without conflicts !

# Interval conflict graphs : enumeration algorithm

A simple Branch-and-Bound algorithm, extension of (Carraghan & Pardalos, 90) :

- ▶ Dual bound : relaxation of conflicts between non-fixed items and integrality.
- ▶ Depth-first search.
- ▶ Branching :



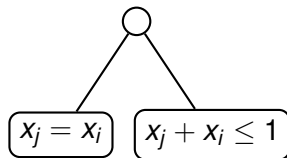
- ▶ Time we spent at each node explored is  $O(n)$ .

## Generic branching scheme

$\lambda$  is fractional  $\Leftrightarrow$  exists a pair  $i, j$  such that  $\sum_{i,j \in b} \lambda_b \neq \{0, 1\}$

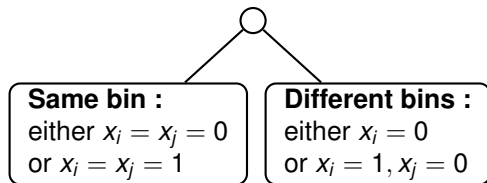
**Branching** : either item  $i$  and  $j$  are in the same bin or not, we add constraints to the pricing subproblem.

### Ryan&Foster scheme



The interval conflict graph structure can be broken.

### Generic branching scheme



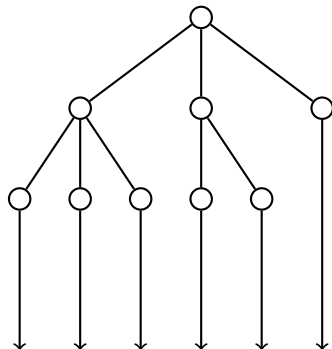
- ▶ Does not break graph structure.
- ▶ Better LP bound after branching.
- ▶ More subproblems to solve.

# “Diving” rounding generic heuristic

Will be presented at ISCO'10 : (Joncour et al., 10)

Combines :

- ▶ Depth-first search : we choose a column ( $\lambda_B \leftarrow 1$ ) and change the right-hand side of the formulation
- ▶ The number of generated columns at each node (except the root is limited)
- ▶ Diversification (Limited Discrepancy Search)
- ▶ Pre-processing



MaxDepth = 2

MaxDiscrepancy = 2

# Implementation

We used **BaPCod** (a generic Branch-and-Price Code) being developed by ReAIOpt

By default, we have :

- ▶ Stabilized column generation procedure
- ▶ Generic branching
- ▶ Generic “diving” heuristic

The user supplies

- ▶ Compact problem formulation
- ▶ An oracle for solving the pricing problem

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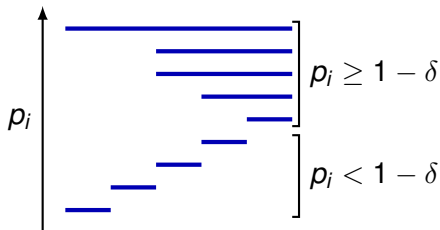
Résultats

# Instances de test

Due to (Gendreau, Laporte, and Semet, 04) :

- ▶ Sizes (integer) :
  - ▶ **Uniforms**(u) :  $w_i \in U[20, 100]$ ,  $W = 150$ .
  - ▶ **Triples**(t) :  $w_i \in U[250, 500]$  (in triples),  $W = 1000$ .
- ▶ “Conflictness” of items :  $p_i \in U[0, 1)$ .
- ▶ Conflict graph density :  $\delta \in \{0, 0.1, \dots, 0.9\}$ .
- ▶ Conflict graphs structure :

$$(i, j) \in E \text{ iff } \frac{p_i + p_j}{2} \geq 1 - \delta$$



# Comparison of exact algorithms

**MIMT** : (Muritiba, Iori, Malaguti, and Toth, 09)

**ELGN** : (Elhedhli, Li, Gzara, and Naoum-Sawaya, 09)

The times adjusted using [www.spec.org](http://www.spec.org)

class	MIMT		ELGN		Notre	
	not opt.	time	not opt.	time	not opt.	time
t60	0%	61	0%	3	0%	2
t120	5%	2976	3%	119	0%	24
t249	4%	2581	11%	398	0%	51
t501	4%	5060	–	–	0%	280
u120	0%	46	0%	47	0%	6
u250	0%	171	1%	183	0%	23
u500	5%	3512	14%	1254	0%	120
u1000	2%	3057	–	–	0%	884



## Comparison of heuristics

**PH** : population heuristic based on tabu search (Muritiba et al., 09)

**DH** : “diving heuristic” (without LDS)

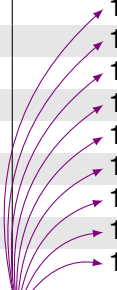
**DH LDS** : “diving heuristic” (with LDS)

class	PH		DH		DH LDS	
	gap	time	gap	time	gap	time
t60	0.45%	60	0.11%	2	0.00%	2
t120	0.62%	64	0.66%	4	0.00%	9
t249	0.39%	83	0.35%	8	0.00%	51
t501	0.21%	94	0.16%	17	0.00%	280
u120	0.10%	36	0.18%	4	0.00%	5
u250	0.21%	83	0.07%	9	0.00%	23
u500	0.20%	112	0.03%	17	0.00%	120
u1000	0.22%	172	0.01%	34	0.002%	822

# Open instances

10 instances which was not solved by (Muritiba et al., 09).

Name	class	# nodes	Sol.	Improv.	Time
6_3_6	t120	609	41	0	3m45s
7_3_4	t249	1	83	-1	29s
8_3_4	t501	1	167	-2	53s
3_4_4	u500	1	204	-3	30s
3_4_5	u500	1	206	-1	29s
3_4_7	u500	1	208	-4	21s
3_4_8	u500	1	205	-1	25s
3_4_9	u500	1	196	-4	36s
4_4_8	u1000	1	404	-7	4m5s
4_4_10	u1000	1	397	-9	7m9s



Effect of the “diving” heuristic

## Instances with arbitrary conflict graphs

- ▶ The same instances except that the conflicts were generated arbitrarily.
- ▶ Time limit : 1 hour.

class	not opt.	time (opt.)	gap (not opt.)	nodes enum.
t60	0%	0	–	384
t120	0%	3	–	1316
t249	5.6%	135	1.1%	3418
t501	27.8%	476	0.6%	16479
u120	0%	1	–	255
u250	1.1%	11	1.0%	405
u500	8.9%	47	0.5%	481
u1000	11.1%	331	0.3%	398

# Conclusions and perspectives

## Conclusions

- ▶ Knapsack problem with interval conflict graph can be solved efficiently (and fast !) by dynamic programming.
- ▶ Generic branch-and-price solver BaPCod is competitive with specialized oracle.
- ▶ Very good performance of the generic “diving” heuristic.

## Future research

- ▶ We need fast algorithms for the knapsack problem with arbitrary conflict graphs (can we do noticeably better than Cplex ?)
- ▶ Improvement of BaPCod (other generic primal heuristics, pre-processing,...)