

# Exact Algorithms for the Two Dimensional Cutting Stock Problem

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19<sup>th</sup> June 2008

# Outline

- 1 Introduction
- 2 Gilmore and Gomory Model
- 3 Branch-and-price-and-cut Algorithm
- 4 Computational Results
- 5 Conclusions and Future Work
- 6 Acknowledgements



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- 1 Introduction
  - Two Dimensional Cutting Stock Problem
  - Literature Review
- 2 Gilmore and Gomory Model
- 3 Branch-and-price-and-cut Algorithm
- 4 Computational Results
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# Cutting Stock Problem

- Combinatorial optimization problem, belonging to the wider family of Cutting and Packing problems
- NP-hard



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## ... two dimensional

- set of items, each item  $i \in \{1, \dots, m\}$  of width  $w_i$ , height  $h_i$  and demand of  $b_i$  pieces
- set of stock sheets of width  $W$  and height  $H$  ( $0 < w_i \leq W$  and  $0 < h_i \leq H, \forall i \in \{1, \dots, m\}$ )
- **Objective:** to minimize the number of used stock sheets



# Guillotine Constraint

- Patterns with uninterrupted cuts, going from one side of the sheet (or one of its already cut fragments) to its opposite side, dividing it in two



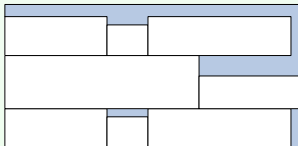
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  - ▶ A cutting pattern is called *n-staged* if it is cut in  $n$  phases. The cuts of each stage are of guillotine type, with the same direction, and two adjacent stages correspond to perpendicular directions

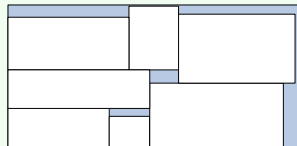


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2-staged guillotine  
cutting pattern



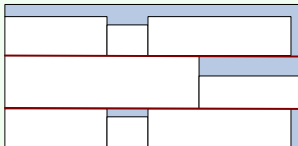
Non guillotine  
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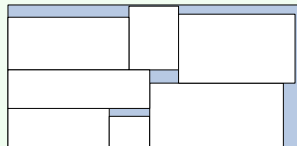


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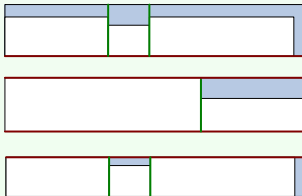


Non guillotine  
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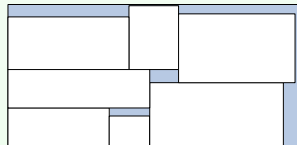


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# Two dimensional cuts with the guillotine constraint

- Gilmore and Gomory (1965)

Multistage cutting stock problems of two and more dimensions.  
*Operations Research*, 13:94-120

- Vanderbeck (2001)

A nested decomposition approach to a three-stage, two-dimensional cutting stock problem. *Management Science*, 47(6):864-879

- Amossen (2005)

Constructive algorithms and lower bounds for guillotine cuttable orthogonal bin packing problems. *Master's thesis*, Department of Computer Science, University of Copenhagen

- Puchinger and Raidl (2007)

Models and algorithms for three-stage two-dimensional bin packing.  
*European Journal of Operational Research*, 127(3):1304-1327



# New algorithm

- Exact solution method for the *Two dimensional cutting stock problem* with the guillotine constraint and two stages



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# New algorithm

- Exact solution method for the *Two dimensional cutting stock problem* with the guillotine constraint and two stages
- Branch-and-price-and-cut Algorithm
  - ▶ model proposed by Gilmore and Gomory (1965)
  - ▶ branching scheme based on the extended arc-flow model for the two dimensional problem with guillotine constraints
  - ▶ new cutting planes



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- 2 Gilmore and Gomory Model**
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# One-dimensional Gilmore and Gomory Model

## Master Problem

$$\begin{aligned}
 \min \quad & \sum_{j \in J} \lambda_j \\
 \text{s.t.} \quad & \sum_{j \in J} a_{ij} \lambda_j \geq b_i \quad \forall i \in \{1, \dots, m\} \\
 & \lambda_j \geq 0 \quad \text{and integer} \quad \forall j \in J
 \end{aligned}$$

$J$ : set of valid cutting patterns

$a_{ij}$ : n° of items  $i$  in cutting pattern  $j$

$\lambda_j$ : n° of times cutting pattern  $j$  is used



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## Pricing Problem

$$\begin{aligned}
 \max \quad & \sum_{i=1}^m \pi_i y_i \\
 \text{s.a.} \quad & w_i y_i \leq W \quad \forall i \in \{1, \dots, m\} \\
 & y_i \geq 0 \quad \text{and integer} \quad \forall i \in \{1, \dots, m\}
 \end{aligned}$$

$\pi_i$ : dual variable associated with constraint  $i$  from the master problem

$y_i$ : n° of times item  $i$  is selected in the new cutting pattern



# Two dimensional Gilmore and Gomory Model

$$\begin{aligned} \min \quad & \sum_{j \in J_0} \lambda_j^0 \\ \text{s.a} \quad & M' \cdot \bar{\lambda} = 0 \\ & M'' \cdot \bar{\lambda} \geq B \\ & \bar{\lambda} \geq 0 \quad \text{and integer} \end{aligned}$$

$J_0$ : set of valid cutting patterns for the first stage

$\lambda_j^0$ :  $j^{th}$  cutting pattern associated to the first stage

$\lambda_j^s$ :  $j^{th}$  cutting pattern associated to the  $s^{th}$  set of patterns of the second stage

$\bar{\lambda}$ :  $(\lambda_1^0, \dots, \lambda_1^1, \dots, \lambda_1^{m'}, \dots)^T$

$B$ :  $(b_1, \dots, b_m)^T$

$M', M''$ : first  $m'$  rows and last  $m$  rows of matrix  $M$ , respectively

$M_0, M_s$ : submatrix of feasible cutting patterns for the first stage and  $s^{th}$  set of the second stage, respectively

$$M = \left( \begin{array}{c|ccc|ccc|ccc|ccc} & & -1 & \dots & -1 & & 0 & & & & & & & & & \\ & & & 0 & & & \dots & & -1 & & & & & & & \\ M_0 & & & \vdots & & & 0 & & & & \dots & & & & \vdots & \\ & & & \vdots & & & \vdots & & & & & & & & 0 & \\ & & & & & & \vdots & & & & & & -1 & \dots & -1 & \\ \hline & 0 & & M_1 & & & M_2 & & & \dots & & & & & M_{m'} & \end{array} \right)$$



# Example

Consider an instance with stock sheets of height  $H = 20$  and width  $W = 30$  and a set of items  $\{(h_i, w_i) : i \in \{1, \dots, 5\}\} = \{(5, 7), (5, 10), (7, 12), (10, 8), (12, 10)\}$ , with demands  $b = (4, 3, 5, 3, 5)$ .



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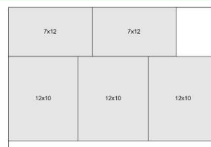
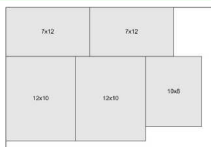
		$\lambda^0_0 \lambda^0_2 \lambda^0_3 \lambda^0_4 \lambda^0_5 \lambda^0_6 \lambda^0_7 \lambda^0_8$	$\lambda^1_1 \lambda^1_2 \lambda^1_3 \lambda^1_4$	$\lambda^2_1 \lambda^2_2 \lambda^2_3$	$\lambda^3_1 \lambda^3_2 \lambda^3_3 \lambda^3_4 \lambda^3_5 \lambda^3_6 \lambda^3_7 \lambda^3_8 \lambda^3_9$	$\lambda^4_1 \lambda^4_2 \lambda^4_3 \lambda^4_4 \lambda^4_5 \lambda^4_6 \lambda^4_7 \lambda^4_8 \lambda^4_9 \lambda^4_{10} \lambda^4_{11} \lambda^4_{12}$	
$(h_i, w_i) =$	$(5, [7, 10])$	4 2 2 1 1	-1 -1 -1 -1	-1 -1 -1	-1 -1 -1 -1 -1 -1 -1 -1	-1 -1 -1 -1 -1 -1 -1 -1 -1 -1 -1 -1	$\equiv 0$
	$(7, 12)$	1 2 1 1					$\equiv 0$
	$(10, 8)$	1 1 2					$\equiv 0$
	$(12, 10)$	1 1					$\equiv 0$
$(h_i, w_i) =$	$(5, 7)$	4 2 1	2 1	3 2	1 1	2 1	$\nabla 4$
	$(5, 10)$	1 2 3	1	2 1	1 1	2 1	$\nabla 3$
	$(7, 12)$	1 1 2	1 1 1	1 1 1	1 1 1	1 1 1	$\nabla 5$
	$(10, 8)$		1 2 1 2 2 1 1 1 3			2 1 1 1 1	$\nabla 5$
	$(12, 10)$					1 2 1 2 1 1 1 1 1 3	$\nabla 5$
F. O.		1 1 1 1 1 1 1 1					

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Optimal Integer Solution: 0 0 1 0 0 0 2 0 1 0 0 1 0 0 2 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 0 1

	$\lambda_1^0$	$\lambda_2^0$	$\lambda_3^0$	$\lambda_4^0$	$\lambda_5^0$	$\lambda_6^0$	$\lambda_7^0$	$\lambda_8^0$	$\lambda_1^1$	$\lambda_2^1$	$\lambda_3^1$	$\lambda_4^1$	$\lambda_5^1$	$\lambda_6^1$	$\lambda_7^1$	$\lambda_8^1$	$\lambda_9^1$	$\lambda_{10}^1$	$\lambda_{11}^1$	$\lambda_{12}^1$	
$(h_i, w_i) =$																					
(5, [7, 10])	4	2	2	1					-1	-1	-1	-1									= 0
(7, 12)		1	2		1	1				-1	-1	-1									= 0
(10, 8)			1			1	2						-1	-1	-1	-1	-1	-1	-1	-1	= 0
(12, 10)					1	1											-1	-1	-1	-1	= 0
$(h_i, w_i) =$																					
(5, 7)									4	2	1		2	1		3	2		1	1	= 4
(5, 10)										1	2	3		1			2	1		1	= 3
(7, 12)										1	1	2				1		1	1		= 5
(10, 8)													1	2	1	2	2	1	1	1	= 3
(12, 10)																	2	1		1	= 5
F. O.	1	1	1	1	1	1	1	1									1	2	1	2	= 5



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  - Branching Scheme
  - Two dimensional Arc-flow Model
  - Some details
  - Cutting Planes
- 4 Computational Results
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# Branching Scheme

Based on variables of an arc-flow model for the two-dimensional guillotine cutting problem, which is an extension of a model for the one dimensional cutting stock problem (VC'1999).

## One dimensional Arc-flow Model:



# Branching Scheme

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## One dimensional Arc-flow Model:

- minimum cost flow model with side constraints

# Branching Scheme

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## One dimensional Arc-flow Model:

- minimum cost flow model with side constraints
- each cutting pattern corresponds to a path in an acyclic directed graph  $G = (V, A)$ 
  - ▶  $V = \{0, 1, \dots, W\}$ : set of  $W + 1$  vertices, which define the positions in the stock sheet
  - ▶  $A = \{(a, b) : 0 \leq a < b \leq W \text{ and } b - a = w_i, \forall i = 1, \dots, m\}$ : set of arcs

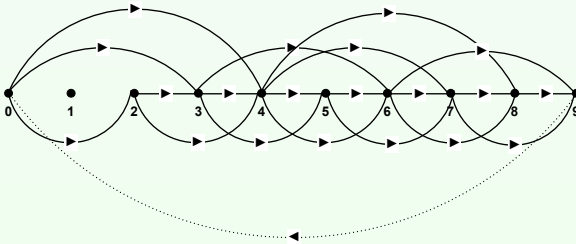
# Example

Consider an instance with stock sheets of width  $W = 9$  and a set of items with widths  $(4, 3, 2)$ .



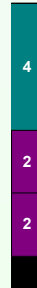
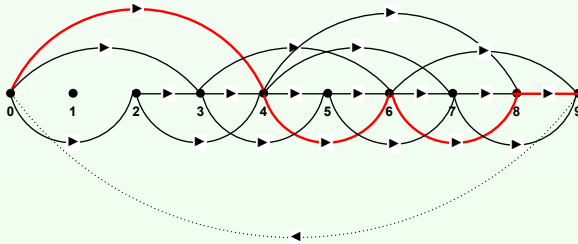
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# One dimensional Arc-Flow Model

$$\begin{aligned}
 \min \quad & z \\
 \text{s.t.} \quad & \sum_{(a,b) \in A} x_{ab} - \sum_{(b,c) \in A} x_{bc} = \begin{cases} -z & , \text{ if } b = 0 \\ 0 & , \text{ if } b = 1, 2, \dots, W-1 \\ z & , \text{ if } b = W \end{cases} \\
 & \sum_{(c,c+l_i) \in A} x_{c,c+l_i} \geq b_i, \quad \forall i \in \{1, \dots, m\} \\
 & x_{ab} \geq 0 \text{ and integer}, \quad \forall (a,b) \in A
 \end{aligned}$$

$x_{ab}$  arc's  $(a, b)$  flow, i.e.,  $n^\circ$  of items of width  $b - a$  placed at a distance of  $a$  units from the beginning of a given stock sheet

$z$  total flow that goes through the graph (return flow from vertex  $W$  to vertex  $0$ )

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## Constraints

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## Constraints

- flow conservation



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$z$  total flow that goes through the graph (return flow from vertex  $W$  to vertex  $0$ )

## Constraints

- flow conservation
- demands fulfillment



# Two dimensional Arc-Flow Model

The one dimensional arc-flow formulation was extended to the two dimensional guillotine case:



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The one dimensional arc-flow formulation was extended to the two dimensional guillotine case:

- for the first stage
  - ▶  $G^0 = (V^0, A^0)$
  - ▶  $V^0 = \{0, 1, \dots, H\}$
  - ▶  $A^0 = \{(a, b) : 0 \leq a < b \leq H \text{ and } b - a = h_i, \forall h_i^* \in H^*\}$
  - ▶  $H^* = \{h_1^*, \dots, h_{m'}^*\}$ : set of  $m'$  different heights ordered by their increasing values

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- for the second stage
  - ▶  $G^s = (V^s, A^s)$
  - ▶  $V^s = \{0, 1, \dots, W\}$
  - ▶  $A^s = \{(d, e) : 0 \leq d < e \leq W \text{ and } d - e = w_i, \forall i : h_i \leq h_s\}$
  - ▶  $s \in \{1, \dots, m'\}$

# Two dimensional Arc-Flow Model

$$\begin{aligned}
 \min \quad & z^0 \\
 \text{s.t.} \quad & \sum_{(a,b) \in A^0} x_{ab}^0 - \sum_{(b,c) \in A^0} x_{bc}^0 = \begin{cases} -z^0 & , \text{ if } b = 0 \\ 0 & , \text{ if } b = 1, 2, \dots, H-1 \\ z^0 & , \text{ if } b = H \end{cases} \\
 & \sum_{(c, c+h_i^*) \in A^0} x_{c, c+h_i^*}^0 - z^i = 0, \quad \forall i \in \{1, \dots, m'\} \\
 & \sum_{\substack{(d,e) \in A^s \\ h^* \in H^*}} x_{deh^*}^s - \sum_{\substack{(e,f) \in A^s \\ h^* \in H^*}} x_{efh^*}^s = \begin{cases} -z^s, & \text{if } e = 0 \\ 0 & , \text{if } e = 1, 2, \dots, W-1, \forall s \in \{1, \dots, m'\} \\ z^s & , \text{if } e = W \end{cases} \\
 & \sum_{s=1}^{m'} \sum_{(f, f+w_i) \in A^s} x_{f, f+w_i, h_i}^s \geq b_i, \quad \forall i \in \{1, \dots, m\} \\
 & x_{ab}^0 \geq 0 \text{ and integer}, \quad \forall (a, b) \in A^0 \\
 & x_{deh^*}^s \geq 0 \text{ and integer}, \quad \forall (d, e) \in A^s, \quad \forall s \in \{1, \dots, m'\}, \quad \forall h^* \in H^*
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# Two dimensional Arc-Flow Model

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$$x_{deh^*}^s \geq 0 \text{ and integer}, \quad \forall (d, e) \in A^s, \quad \forall s \in \{1, \dots, m'\}, \quad \forall h^* \in H^*$$



# Two dimensional Arc-Flow Model

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$$\text{s.t.} \quad \sum_{(a,b) \in A^0} x_{ab}^0 - \sum_{(b,c) \in A^0} x_{bc}^0 = \begin{cases} -z^0 & , \text{ if } b = 0 \\ 0 & , \text{ if } b = 1, 2, \dots, H-1 \\ z^0 & , \text{ if } b = H \end{cases}$$

$$\sum_{(c, c+h_i^*) \in A^0} x_{c, c+h_i^*}^0 - z^i = 0, \quad \forall i \in \{1, \dots, m'\}$$

$$\sum_{\substack{(d,e) \in A^s \\ h^* \in H^*}} x_{deh^*}^s - \sum_{\substack{(e,f) \in A^s \\ h^* \in H^*}} x_{efh^*}^s = \begin{cases} -z^s, & \text{if } e = 0 \\ 0 & , \text{if } e = 1, 2, \dots, W-1, \forall s \in \{1, \dots, m'\} \\ z^s & , \text{if } e = W \end{cases}$$

$$\sum_{s=1}^{m'} \sum_{(f, f+w_i) \in A^s} x_{f, f+w_i, h_i}^s \geq b_i, \quad \forall i \in \{1, \dots, m\}$$

$$x_{ab}^0 \geq 0 \text{ and integer, } \forall (a, b) \in A^0$$

$$x_{deh^*}^s \geq 0 \text{ and integer, } \forall (d, e) \in A^s, \quad \forall s \in \{1, \dots, m'\}, \quad \forall h^* \in H^*$$



# Two dimensional Arc-Flow Model

$$\begin{aligned}
 \min \quad & z^0 \\
 \text{s.t.} \quad & \sum_{(a,b) \in A^0} x_{ab}^0 - \sum_{(b,c) \in A^0} x_{bc}^0 = \begin{cases} -z^0 & , \text{ if } b = 0 \\ 0 & , \text{ if } b = 1, 2, \dots, H-1 \\ z^0 & , \text{ if } b = H \end{cases} \\
 & \sum_{(c, c+h_i^*) \in A^0} x_{c, c+h_i^*}^0 - z^i = 0, \quad \forall i \in \{1, \dots, m'\} \\
 & \sum_{\substack{(d,e) \in A^s \\ h^* \in H^*}} x_{deh^*}^s - \sum_{\substack{(e,f) \in A^s \\ h^* \in H^*}} x_{efh^*}^s = \begin{cases} -z^s, & \text{if } e = 0 \\ 0 & , \text{if } e = 1, 2, \dots, W-1, \forall s \in \{1, \dots, m'\} \\ z^s & , \text{if } e = W \end{cases} \\
 & \sum_{s=1}^{m'} \sum_{(f, f+w_i) \in A^s} x_{f, f+w_i, h_i}^s \geq b_i, \quad \forall i \in \{1, \dots, m\} \\
 & x_{ab}^0 \geq 0 \text{ and integer}, \quad \forall (a, b) \in A^0 \\
 & x_{deh^*}^s \geq 0 \text{ and integer}, \quad \forall (d, e) \in A^s, \quad \forall s \in \{1, \dots, m'\}, \quad \forall h^* \in H^*
 \end{aligned}$$



# Two dimensional Arc-Flow Model

min  $z^0$

$$\text{s.t.} \quad \sum_{(a,b) \in A^0} x_{ab}^0 - \sum_{(b,c) \in A^0} x_{bc}^0 = \begin{cases} -z^0 & , \text{ if } b = 0 \\ 0 & , \text{ if } b = 1, 2, \dots, H-1 \\ z^0 & , \text{ if } b = H \end{cases}$$

$$\sum_{(c, c+h_i^*) \in A^0} x_{c, c+h_i^*}^0 - z^i = 0, \quad \forall i \in \{1, \dots, m'\}$$

$$\sum_{\substack{(d,e) \in A^s \\ h^* \in H^*}} x_{deh^*}^s - \sum_{\substack{(e,f) \in A^s \\ h^* \in H^*}} x_{efh^*}^s = \begin{cases} -z^s, & \text{if } e = 0 \\ 0 & , \text{if } e = 1, 2, \dots, W-1, \forall s \in \{1, \dots, m'\} \\ z^s & , \text{if } e = W \end{cases}$$

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    - $3^{rd}$  (for the second stage's graphs) the one corresponding to the highest item.
- ▶ Branching strategy: *Depth-First-Search*.
- ▶ Pricing problem is *always* a knapsack problem, solved with dynamic programming.



# Cutting Planes

- All items with height greater than or equal to  $h_j$  must be cut out of strips of height greater than or equal to  $h_j$



# Cutting Planes

- All items with height greater than or equal to  $h_j$  must be cut out of strips of height greater than or equal to  $h_j$
- Considering the trivial lower bound,  $\forall j \in \{1, \dots, m'\}$

$$\sum_{l=j}^{m'} z^l \geq \left\lceil \frac{\sum_{i \in I_j} w_i b_i}{W} \right\rceil$$

$$I_j = \{i \in \{1, \dots, m\} : h_i \geq h_j\}$$

$m$  = number of different items

$m'$  = number of different heights



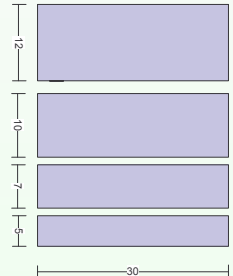
# Example

	$(h,w)$	$b$
<i>items 1</i>	(5,7)	4
<i>items 2</i>	(5,10)	3
<i>items 3</i>	(7,12)	5
<i>items 4</i>	(10,8)	3
<i>items 5</i>	(12,10)	5

$H = 20$



$W = 30$



# Example

	$(h,w)$	$b$	$\left\lceil \frac{\sum_{i \in I_j} w_i b_i}{W} \right\rceil$
<i>items 1</i>	(5,7)	4	
<i>items 2</i>	(5,10)	3	
<i>items 3</i>	(7,12)	5	
<i>items 4</i>	(10,8)	3	
<i>items 5</i>	(12,10)	5	



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<i>items 5</i>	(12,10)	5	$\left\lceil \frac{10 \times 5}{30} \right\rceil = 2$



# Example

	$(h,w)$	$b$	$\left\lceil \frac{\sum_{i \in I_j} w_i b_i}{W} \right\rceil$
<i>items 1</i>	(5,7)	4	
<i>items 2</i>	(5,10)	3	
<i>items 3</i>	(7,12)	5	
<i>items 4</i>	(10,8)	3	
<i>items 5</i>	(12,10)	5	$\left\lceil \frac{10 \times 5}{30} \right\rceil = 2 \Rightarrow$ at least 2 strips with heights equal to 12 are required



# Example

	$(h,w)$	$b$	$\left\lceil \frac{\sum_{i \in I_j} w_i b_i}{W} \right\rceil$		
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<i>items 2</i>	(5,10)	3			
<i>items 3</i>	(7,12)	5			
<i>items 4</i>	(10,8)	3	$\left\lceil \frac{10 \times 5 + 8 \times 3}{30} \right\rceil$	= 3	
<i>items 5</i>	(12,10)	5	$\left\lceil \frac{10 \times 5}{30} \right\rceil$	= 2	$\Rightarrow$ at least 2 strips with heights equal to 12 are required

# Example

	$(h,w)$	$b$	$\left\lceil \frac{\sum_{i \in I_j} w_i b_i}{W} \right\rceil$		
<i>items 1</i>	(5,7)	4			
<i>items 2</i>	(5,10)	3			
<i>items 3</i>	(7,12)	5			
<i>items 4</i>	(10,8)	3	$\left\lceil \frac{10 \times 5 + 8 \times 3}{30} \right\rceil$	$= 3 \Rightarrow$	at least 3 strips with heights greater than or equal to 10 are required
<i>items 5</i>	(12,10)	5	$\left\lceil \frac{10 \times 5}{30} \right\rceil$	$= 2 \Rightarrow$	at least 2 strips with heights equal to 12 are required



# Example

	$(h,w)$	$b$	$\left\lceil \frac{\sum_{i \in I_j} w_i b_i}{W} \right\rceil$		
<i>items 1</i>	(5,7)	4			
<i>items 2</i>	(5,10)	3	...		
<i>items 3</i>	(7,12)	5			
<i>items 4</i>	(10,8)	3	$\left\lceil \frac{10 \times 5 + 8 \times 3}{30} \right\rceil$	$= 3 \Rightarrow$	at least 3 strips with heights greater than or equal to 10 are required
<i>items 5</i>	(12,10)	5	$\left\lceil \frac{10 \times 5}{30} \right\rceil$	$= 2 \Rightarrow$	at least 2 strips with heights equal to 12 are required

# Outline

- 1 Introduction
- 2 Gilmore and Gomory Model
- 3 Branch-and-price-and-cut Algorithm
- 4 Computational Results**
  - Branch-and-price-and-cut
  - Comparison with Arc-flow model
  - Cutting planes
- 5 Conclusions and Future Work
- 6 Acknowledgements

# Implementation

- The branch-and-price algorithm was coded in C++
- Plain column generation (no heuristics, no stabilization, ...).
- Some of the optimization subroutines were implemented by using the CPLEX 8.0 Callable Library
- The computational tests were run on a PC with a 1.83 GHz Intel Core Duo processor and a 512MB RAM
- Set of 43 real instances from the furniture industry



# Terminology

- $m$ : number of different items
- $area_{items}$ : sum of the item's areas
- $cols_{ini}$ : initial number of columns
- $cols_{LP}$ : number of columns generated during the solution of the linear relaxation
- $cols_{BB}$ : number of columns generated during the branch-and-bound process
- $sp_{LP}$ : number of pricing problems solved before branching
- $sp_{BB}$ : number of pricing problems solved during the branch-and-bound process
- $nodes_{BB}$ : number of searched branching nodes
- $t_{LR}$ : computational time (in seconds) spent with the LP relaxation
- $t_{BB}$ : computational time (in seconds) spent with the branch-and-bound process
- $t_{tot}$ : total time (in seconds)
- $z_{LR}$ : linear relaxation solution
- $z_{IP}/UB$ : integer solution or the reached upper bound

⇒ rows marked with an \* represent instances in which the algorithm didn't find the optimal solution after 7200 seconds or 5000 searched branching nodes

Name	m	area <sub>items</sub>	cols <sub>ini</sub>	spLP	cols <sub>LP</sub>	spBB	cols <sub>BB</sub>	nodes <sub>BB</sub>	t <sub>LR</sub>	t <sub>BB</sub>	t <sub>tot</sub>	z <sub>LR</sub>	z <sub>IP</sub> /UB
AP-9-3MM-4MM-1	2	7,45E+06	1	15	7	21	0	7	0,02	0	0,02	3,3	4
AP-9-3MM-4MM-2	4	7,51E+07	1	17	6	0	0	0	0,02	0	0,02	36	36
AP-9-3MM-4MM-3	2	1,63E+07	1	10	4	0	0	0	0,05	0	0,05	8	8
AP-9-3MM-4MM-4	2	7,31E+06	1	13	5	3	0	1	0	0	0	2,667	3
AP-9-3MM-4MM-5	8	5,73E+07	1	252	73	544	25	62	0,27	0,48	0,75	12,525	13
AP-9-3MM-4MM-6	2	3,96E+06	1	7	3	2	0	1	0,02	0	0,02	1,889	2
AP-9-3MM-4MM-7	5	3,15E+07	1	57	20	19	1	3	0,02	0,02	0,03	13,125	14
AP-9-3MM-4MM-8	1	2,41E+06	1	5	2	2	0	1	0,01	0	0,01	1,067	2
AP-9-1	30	2,93E+08	1					5000			665,67	60,671	62 *
AP-9-2	3	9,88E+06	1	32	14	70	4	16	0,01	0,05	0,06	2	3
AP-9-3	20	2,32E+08	1	1440	317	12147	107	744	9,59	45,2	54,8	45,758	46
AP-9-4	3	5,82E+07	1	20	7	0	0	0	0,01	0	0,01	14	14
AP-9-5	8	6,35E+07	1	242	83	74	2	9	0,63	0,13	0,75	13,533	14
AP-9-6	31	3,38E+08	1					5000			572,89	66,824	74 *
AP-9-7	12	1,88E+08	1	318	93	2306	17	254	0,53	2,5	3,03	38,942	39
AP-9-8	27	4,17E+08	1	1532	364	1425	50	71	10,22	5,39	15,61	82,256	83
AP-9-9	3	2,39E+07	1					5000			159,2	4,704	116 *
AP-9-10	20	3,16E+08	1					5000			2936,95	64,681	66 *
AP-9-11	27	2,73E+08	1	2532	458	630	9	31	13,67	1,91	15,58	57,234	58
AP-9-12	10	1,20E+08	1	293	89	341	31	33	0,72	0,56	1,28	26,098	27
AP-9-13	21	1,36E+08	1	1091	280	2107	112	128	4,44	5,64	10,08	27,235	28
AP-9-14	1	1,04E+07	1	5	2	2	0	1	0,01	0	0,01	2,4	3
AP-9-15	8	6,32E+07	1					5000			39,11	12,922	14 *
FA+AA-9-1	107	1,65E+08	1	65419	3888	57013	1851	786	1036,99	702,2	1739,19	34,356	35
FA+AA-9-2	75	8,54E+07	1					3614			7200	17,327	19 *
FA+AA-9-4	34	3,42E+07	1	4399	597	1188	57	33	20,38	4	24,38	7,08	8 *
FA+AA-9-6	79	9,62E+07	1					4211			7201,86	19,2	23 *
FA+AA-9-7	54	5,30E+07	1	14454	1872	118676	1007	3253	154,84	830,17	985,02	11,167	12 *
FA+AA-9-8	82	1,35E+08	1					5000			5643,8	27,287	30 *
FA+AA-9-9	24	1,76E+07	1					5000			4115,2	3,677	5 *
FA+AA-9-10	36	3,64E+07	1	6196	903	14228	559	453	30,7	51,27	81,97	7,487	8 *
FA+AA-9-11	99	1,30E+08	1	28939	2724	168670	883	3976	467,88	1789,22	2257,09	26,897	27 *
FA+AA-9-13	134	1,75E+08	1					2196			7206,91	34,67	38 *
FA+AA-9-14	26	2,46E+07	1	1844	411	868	48	50	5,06	1,92	6,99	5,221	6
FA+AA-9-15	68	8,09E+07	1	13572	1675	29282	293	984	147,08	211,63	358,7	16,401	17
FA+AP-9-10MM-1	16	3,48E+07	1	204	74	226	12	15	0,39	0,31	0,7	8,875	9 *
FA+AP-9-10MM-2	8	2,22E+07	1					5000			293,03	4,631	35
FA+AP-9-10MM-3	42	1,08E+08	1	3649	481	11173	238	395	17,69	37,05	54,73	22,123	23
FA+AP-9-10MM-4	11	1,46E+07	1	227	78	213	7	22	0,23	0,14	0,38	3,058	4 *
TRAS-BC-2	40	4,81E+07	1					5000			322,3	15,896	20 *
TRAS-BC-3	32	4,73E+07	1	2130	324	454	17	12	10,39	1,58	11,97	18,5	19
TRAS-BC-4	8	1,48E+07	1	85	23	54	3	7	0,06	0,01	0,08	7,167	8
TRAS-BC-5	11	1,59E+07	1	205	58	7969	17	722	0,16	4,88	5,03	6,417	7

# Arc-flow model

- The procedure for generating the arcs was coded in C++
- The arc-flow model was run in the ILOG CPLEX 10.2
- The computational tests were run on a PC with a 1.87 GHz Intel Core Duo processor and a 2GB RAM
- Set of 43 real instances from the furniture industry

# Comparison of computational results

Name	n	Branch-and-price				Arc-flow model	
		nodes	LR	z	t (s)	z	t (s)
AP-9-3MM-4MM-1	2	7	3,3	4	0,02	4	0,156
AP-9-3MM-4MM-2	4	0	36	36	0,02	36	0,313
AP-9-3MM-4MM-3	2	0	8	8	0,05	8	0,156
AP-9-3MM-4MM-4	2	1	2,667	3	0	3	0,172
AP-9-3MM-4MM-5	8	62	12,525	13	0,75	13	0,688
AP-9-3MM-4MM-6	2	1	1,889	2	0,02	2	0,094
AP-9-3MM-4MM-7	5	3	13,125	14	0,03	14	0,625
AP-9-3MM-4MM-8	1	1	1,067	2	0,01	2	0,078
AP-9-1	30	5000	60,671	62	665,67 *	61	27,109
AP-9-2	3	16	2	3	0,06	3	0,250
AP-9-3	20	744	45,758	46	54,8	46	7,531
AP-9-4	3	0	14	14	0,01	14	0,297
AP-9-5	8	9	13,533	14	0,75	14	0,781
AP-9-6	31	5000	66,824	74	572,89 *	67	183,516
AP-9-7	12	254	38,942	39	3,03	39	1,094
AP-9-8	27	71	82,256	83	15,61	83	12,031
AP-9-9	3	5000	4,704	116	159,2 *	5	0,219
AP-9-10	20	5000	64,681	66	2936,95 *	65	2,953
AP-9-11	27	31	57,234	58	15,58	58	17,297
AP-9-12	10	33	26,098	27	1,28	27	1,047
AP-9-13	21	128	27,235	28	10,08	28	57,578
AP-9-14	1	1	2,4	3	0,01	3	0,094

Name	n	Branch-and-price				Arc-flow model	
		nodes	LR	z	t (s)	z	t (s)
AP-9-15	8	5000	12,922	14	39,11 *	14	0,750
FA+AA-9-1	107	786	34,356	35	1739,19	35	1010,891
FA+AA-9-2	75	3614	17,327	19	7200 *	18	2595,375
FA+AA-9-4	34	33	7,08	8	24,38	8	18,578
FA+AA-9-6	79	4211	19,2	23	7201,86 *	20	3963,422
FA+AA-9-7	54	3253	11,167	12	985,02	12	118,797
FA+AA-9-8	82	5000	27,287	30	5643,8 *	28	3155,500
FA+AA-9-9	24	5000	3,677	5	4115,2 *	4	13,719
FA+AA-9-10	36	453	7,487	8	81,97	8	23,031
FA+AA-9-11	99	3976	26,897	27	2257,09	27	81,938
FA+AA-9-13	134	2196	34,67	38	7206,91 *	35	4793,156
FA+AA-9-14	26	50	5,221	6	6,99	6	2,094
FA+AA-9-15	68	984	16,401	17	358,7	17	44,578
FA+AP-9-10MM-1	16	15	8,875	9	0,7	9	1,641
FA+AP-9-10MM-2	8	5000	4,631	35	293,03 *	5	0,828
FA+AP-9-10MM-3	42	395	22,123	23	54,73	23	11,531
FA+AP-9-10MM-4	11	22	3,058	4	0,38	4	0,969
TRAS-BC-2	40	5000	15,896	20	322,3 *	17	79,047
TRAS-BC-3	32	12	18,5	19	11,97	19	4,594
TRAS-BC-4	8	7	7,167	8	0,08	8	0,922
TRAS-BC-5	11	722	6,417	7	5,03	7	0,953



# Strengthening bounds

Name	n	Arc-flow model (LR)	
		without cut	with cut
AP-9-3MM-4MM-1	2	3.375	3.375
AP-9-3MM-4MM-2	4	36.000	36.000
AP-9-3MM-4MM-3	2	8.000	8.000
AP-9-3MM-4MM-4	2	2.667	2.667
AP-9-3MM-4MM-5	8	12.525	12.525
AP-9-3MM-4MM-6	2	1.889	2.000
AP-9-3MM-4MM-7	5	13.125	13.125
AP-9-3MM-4MM-8	1	2.000	2.000
AP-9-1	30	60.671	60.671
AP-9-2	3	1.969	2.308
AP-9-3	20	45.758	45.790
AP-9-4	3	14.000	14.000
AP-9-5	8	13.533	13.533
AP-9-6	31	66.824	66.824
AP-9-7	12	38.942	38.942
AP-9-8	27	82.256	82.256
AP-9-9	3	4.704	4.726
AP-9-10	20	64.649	64.654
AP-9-11	27	57.234	57.234
AP-9-12	10	26.098	26.098
AP-9-13	21	27.235	27.289
AP-9-14	1	3.000	3.000

Name	n	Arc-flow model (LR)	
		without cut	with cut
AP-9-15	8	12.922	13.120
FA+AA-9-1	107	34.321	34.321
FA+AA-9-2	75	17.015	17.294
FA+AA-9-4	34	7.049	7.088
FA+AA-9-6	79	19.191	19.191
FA+AA-9-7	54	10.839	11.125
FA+AA-9-8	82	27.287	27.287
FA+AA-9-9	24	3.597	3.658
FA+AA-9-10	36	7.474	7.474
FA+AA-9-11	99	26.849	26.849
FA+AA-9-13	134	34.634	34.634
FA+AA-9-14	26	5.195	5.242
FA+AA-9-15	68	16.360	16.373
FA+AP-9-10MM-1	16	8.875	8.875
FA+AP-9-10MM-2	8	4.631	4.631
FA+AP-9-10MM-3	42	22.110	22.110
FA+AP-9-10MM-4	11	3.047	3.050
TRAS-BC-2	40	15.813	15.813
TRAS-BC-3	32	18.500	18.500
TRAS-BC-4	8	7.167	7.375
TRAS-BC-5	11	6.375	6.375



# Strengthening bounds

Name	n	Arc-flow model (LR)	
		without cut	with cut
AP-9-3MM-4MM-1	2	3.375	3.375
AP-9-3MM-4MM-2	4	36.000	36.000
AP-9-3MM-4MM-3	2	8.000	8.000
AP-9-3MM-4MM-4	2	2.667	2.667
AP-9-3MM-4MM-5	8	12.525	12.525
AP-9-3MM-4MM-6	2	1.889	2.000
AP-9-3MM-4MM-7	5	13.125	13.125
AP-9-3MM-4MM-8	1	2.000	2.000
AP-9-1	30	60.671	60.671
AP-9-2	3	1.969	2.308
AP-9-3	20	45.758	45.790
AP-9-4	3	14.000	14.000
AP-9-5	8	13.533	13.533
AP-9-6	31	66.824	66.824
AP-9-7	12	38.942	38.942
AP-9-8	27	82.256	82.256
AP-9-9	3	4.704	4.726
AP-9-10	20	64.469	64.654
AP-9-11	27	57.234	57.234
AP-9-12	10	26.098	26.098
AP-9-13	21	27.235	27.289
AP-9-14	1	3.000	3.000

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# Outline

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- 2 Gilmore and Gomory Model
- 3 Branch-and-price-and-cut Algorithm
- 4 Computational Results
- 5 Conclusions and Future Work**
- 6 Acknowledgements

## Conclusions

- new exact algorithm for the 2-dim.cutting stock problem
- ongoing research
- original pseudo-polynomial arc-flow model solves faster with fewer nodes.
- There is room for improvement of plain column generation algorithm (heuristics, dual cuts, other stabilization, ...)



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- new exact algorithm for the 2-dim.cutting stock problem
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## Future Work

- new cutting planes from maximal Dual Feasible Functions (DFF) (compatible with subproblem).
- rotation of items.
- both first cut vertical and first cut horizontal.

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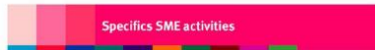
# Acknowledgements

- Project SCOOP (<http://www.scoop-project.net/>).
- other CSP approaches also being developed (heuristic,...).
- other issues being addressed (open stacks,...).

This research was done in project SCOOP (Sheet cutting and process optimization for furniture enterprises) (Contract N° COOP-CT- 006-032998), funded by the European Commission, 6th Framework Programme on Research, Technological Development and Demonstration, specific actions for SMEs, Cooperative Research Projects.



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