

# Explaining the `ATMOSTSEQCARD` constraint

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## Outline

Hybrid CP/SAT Solving

The ATMOSTSEQCARD constraint

Explaining ATMOSTSEQCARD

Experimental results

Conclusion & Future research

## Context

**SAT & CP** : Can we get the best from both approaches?

→ A key concept : explaining constraints

An explanation is a set of assignments/prunings triggering a failure/pruning.

### example

Cardinality Constraint :  $\sum_{i=1}^n x_i \leq k$  ;  $D(x_i) = \{0, 1\}$ .

$x_i \leftarrow 1$  is pruned if we already have  $k$  appearances of the value 1.

$$\{x_j \leftarrow 1 \mid D(x_j) = \{1\}\} \rightarrow x_i \leftarrow 1 .$$

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$\text{ATMOSTSEQCARD}(u, q, d, [x_1, \dots, x_n]) \Leftrightarrow$

$$\bigwedge_{i=0}^{n-q} \left( \sum_{l=1}^q x_{i+l} \leq u \right) \wedge \left( \sum_{i=1}^n x_i = d \right)$$

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Example  $\text{ATMOSTSEQCARD}(2, 4, 4, [x_1, \dots, x_7])$

<u>0</u>	<u>1</u>	<u>1</u>	<u>0</u>	<u>1</u>	<u>1</u>	<u>0</u>
	—	—	—			
		—	—	—	—	
		—	—	—	—	

<u>1</u>	<u>1</u>	<u>0</u>	<u>0</u>	<u>1</u>	<u>0</u>	<u>1</u>
	—	—	—			
		—	—	—	—	
		—	—	—	—	

## The propagator

- `leftmost`: computes an assignment  $w$  maximizing the cardinality of the sequence with respect to the `ATMOST` constraints.
- Let  $max(i)$  be the maximum cardinality of the  $q$  subsequences involving  $x_i$  when computing `leftmost`[ $i$ ].

$$\begin{array}{cccc|l}
 0 & 1 & 0 & . & 0 & 1 & 0 & \\
 \text{---} & \text{---} & \text{---} & & & & & \textit{cardinality} = 1 \\
 & \text{---} & \text{---} & \text{---} & & & & \textit{cardinality} = 0 \\
 & & \text{---} & \text{---} & \text{---} & & & \textit{cardinality} = 1 \\
 & & & & & & & \textit{max}(4) = 1
 \end{array}$$

- $Left[i] = \sum_{j=1}^{j=i} leftmost[j]$ .
- $Right[i]$  : same as  $Left$  but in the reverse sense, i.e.  $[x_n, \dots, x_1]$ .
- Example : with `ATMOST(2,5)`:

$\mathcal{D}(x_i)$	0	.	.	.	.	1	.	.
$max(i)$	0	1	2	2	2	2	1	2
$leftmost[i]$	0	1	0	0	0	1	1	0
$Left[i]$	0	1	1	1	1	1	2	2



## Domain consistency

- DC on each ATMOST:  $(\sum_{l=1}^q x_{i+l} \leq u)$
- DC on  $\sum_{i=1}^n x_i = d$
- If  $Left[n] < d$  Then *fail*
- If  $Left[n] = d$  and  $Left[i] + Right[n - i + 1] \leq d$  Then  $\mathcal{D}(x_i) \leftarrow \{0\}$
- If  $Left[n] = d$  and  $Left[i - 1] + Right[n - i] < d$  Then  $\mathcal{D}(x_i) \leftarrow \{1\}$

## Explaining ATMOSTSEQCARD: the key idea

### Explaining Failure

- 1 If a failure is triggered by a cardinality constraint (i.e.  $(\sum_{i=1}^q x_{i+l} \leq u)$  or  $\sum_{i=1}^n x_i = d$ ), then it is easy to generate an explanation.
- 2 If a failure triggered by  $Left[n] < d$ , a naive explanation would be the set of all assignments in the sequence.

## Theorem

### Theorem

Let  $S$  be the set of all assignments,

$S^* = S \setminus (\{x_i \leftarrow 0 \mid \max(i) = u\} \cup \{x_i \leftarrow 1 \mid \max(i) \neq u\})$ , then  
 $S^*$  is a valid explanation.

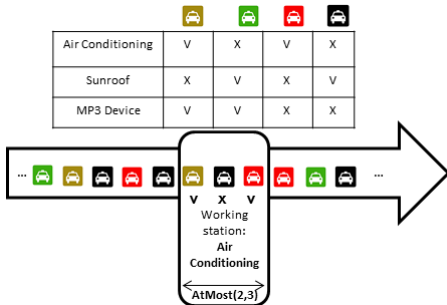
→ runs in  $O(n)$  since we call `leftmost_count` once.

# Example : ATMOSTSEQCARD(2, 5, 8, [x<sub>1</sub>, ..x<sub>22</sub>])

<i>S</i>	1 0 1 0 0 . . 0 0 0 1 1 0 0 0 0 0 1 0 0 0 0 1
leftmost( <i>S</i> (x <sub>i</sub> ))	1 0 1 0 0 1 0 0 0 0 1 1 0 0 0 0 1 0 0 0 0 1
<i>Left</i> [ <i>i</i> ]	1 1 2 2 2 3 3 3 3 3 4 5 5 5 5 5 6 6 6 6 6 7
	<p><i>Left</i>[22] = 7 &lt; 8 : FAILURE</p>
<i>max</i> ( <i>i</i> )	2 2 2 2 2 1 2 2 2 2 2 2 2 2 1 1 1 1 1 1 1
<i>S</i> <sup>*</sup>	1 . 1 . . . . . 1 1 . . . 0 . 0 0 0 0 .

The final explanation size |*S*<sup>\*</sup>| is 9 while the naive one (|*S*|) is 20.

# Car-sequencing



## Constraints

- Each class  $c$  is associated with a demand  $D_c$ .
- For each option  $j$ , each sub-sequence of size  $q_j$  must contain at most  $u_j$  cars requiring the option  $j$ .

## Configuration

- Mistral as a hybrid CP/SAT solver
  - ① *hybrid (VSIDS)* uses VSIDS;
  - ② *hybrid (Slot)* uses a cp heuristic based on the usage rate.
  - ③ *hybrid (Slot → VSIDS)* first uses *hybrid (Slot)* then switches after 100 non-improving restarts to VSIDS.
  - ④ *hybrid (VSIDS → Slot)* uses VSIDS and switches after 100 non-improving restarts to *hybrid (Slot)*.
- *pure-CP*: Mistral without clause learning using the *Slot* branching.

Table: Experimental Evaluation

Method	sat[easy] ( $74 \times 5$ )			sat[hard] ( $7 \times 5$ )			unsat ( $28 \times 5$ )		
	#suc	avg fails	time	#suc	avg fails	time	#suc	avg fails	time
<i>hybrid (VSIDS)</i>	370	903	0.23	16	207211	286.32	35	177806	224.78
<i>hybrid (VSIDS <math>\rightarrow</math> Slot)</i>	370	739	0.23	35	76256	64.52	<b>37</b>	<b>204858</b>	<b>248.24</b>
<i>hybrid (Slot <math>\rightarrow</math> VSIDS)</i>	370	132	0.04	34	4568	2.50	37	234800	287.61
<i>hybrid (Slot)</i>	370	132	0.04	<b>35</b>	<b>6304</b>	<b>3.75</b>	23	174097	299.24
<i>pure-CP</i>	<b>370</b>	<b>43.06</b>	<b>0.03</b>	35	57966	16.25	0	-	-

## Conclusion & Future research

### Contributions & Analysis

- A linear time explanation for the ATMOSTSEQCARD constraint
- The experimental results emphasize the importance of using a hybrid approach instead of pure-CP!

### Future research

- Can we generate optimal explanations?
- Is it worthy to use a 'sophisticated' [explanation + propagator] instead of decomposing to simpler constraints?
- To encode into SAT or to propagate?



# Thank you!

## Questions?