5.82 cumulative

	DESCRIPTION	LINKS	GRAPH	AUTOMATON
Origin	[1]			
Constraint	cumulative(TASKS,LIMI	T)		
Synonym	cumulative_max.			
Arguments	TASKS : collection LIMIT : int	<pre>(origin-dvar, duration-dvar, end-dvar, height-dvar</pre>		
Restrictions	$\begin{array}{l} \textbf{require_at_least}(2, \texttt{TA}\\ \textbf{required}(\texttt{TASKS}, \texttt{heigh}\\ \texttt{TASKS.duration} \geq 0\\ \texttt{TASKS.origin} \leq \texttt{TASKS}\\ \texttt{TASKS.height} \geq 0\\ \texttt{LIMIT} \geq 0 \end{array}$	ASKS,[origin,durat t) .end	ion, end])	
Purpose	Cumulative scheduling conset \mathcal{T} of tasks described by that at each point in time, the does not exceed a given lime than or equal to <i>i</i> , and (2) of \mathcal{T} the constraint origin	instraint or scheduling y the TASKS collection the cumulated height of nit. A task overlaps a p its end is strictly great a + duration = end.	under resource constraints . The cumulative constr f the set of tasks that overl point <i>i</i> if and only if (1) its er than <i>i</i> . It also imposes	c. Consider a aint enforces ap that point, origin is less for each task
Example	$\left(\begin{array}{c} \text{origin}-1 & \text{dr}\\ \text{origin}-2 & \text{dr}\\ \text{origin}-3 & \text{dr}\\ \text{origin}-6 & \text{dr}\\ \text{origin}-7 & \text{dr}\end{array}\right)$	ration - 3 end - ration - 9 end - ration - 10 end - ration - 6 end - ration - 2 end -	$ \begin{array}{c c} -4 & \text{height} -1, \\ -11 & \text{height} -2, \\ -13 & \text{height} -1, \\ -12 & \text{height} -1, \\ -9 & \text{height} -3 \end{array} \right\rangle, 8 $	3
	Figure 5.165 shows the c task of the cumulative c same colour: the sum of t the task, while the height task have the same height) cumulative constraint hol resource consumption strict	cumulated profile ass onstraint corresponds the lengths of the rec of the rectangles (i.e corresponds to the re ds since at each poin ly greater than the upp	ociated with the examp a set of rectangles colo tangles corresponds to th a, all the rectangles asso esource consumption of t t in time we do not have er limit 8 enforced by the	le. To each ured with the he duration of ociated with a the task. The e a cumulated last argument

of the cumulative constraint.

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Typical	<pre> TASKS > 1 range(TASKS.origin) > 1 range(TASKS.duration) > 1 range(TASKS.end) > 1 range(TASKS.height) > 1 TASKS.duration > 0 TASKS.height > 0 LIMIT < sum(TASKS.height)</pre>		
Symmetries	• Items of TASKS are permutable.		
	• TASKS.height can be decreased to any value ≥ 0 .		
	• One and the same constant can be added to the origin and end attributes of all items of TASKS.		
	• LIMIT can be increased.		
Remark	In the original cumulative constraint of CHIP the LIMIT parameter was a domain variable corresponding to the <i>maximum peak of the resource consumption profile</i> . Given a fixed time frame, this variable could be used as a cost in order to directly minimise the maximum resource consumption peak.		
	Some systems like Ilog CP Optimizer also assume that a zero-duration task overlaps a point i if and only if (1) its origin is less than or equal to i , and (2) its end is greater than or equal to i . Under this definition, the height of a zero-duration task is also taken into account in the resource consumption profile.		
	Note that the concept of cumulative is <i>different</i> from the concept of rectangles non-overlapping even if, most of the time, each task of a ground solution of a cumulative constraint is simply drawn as a single rectangle. As illustrated by Figure 5.206, this is in fact not always possible (i.e., some rectangles may need to be broken apart). In fact the cumulative constraint is only a necessary condition for rectangles non-overlapping (see Figure 5.205 and the corresponding explanation in the Algorithm slot of the diffn constraint).		
Algorithm	The first filtering algorithms were related to the notion of compulsory part of a task [223].		

The first filtering algorithms were related to the notion of compulsory part of a task [223]. They compute a cumulated resource profile of all the compulsory parts of the tasks and



Figure 5.165: Resource consumption profile

prune the origins of the tasks with respect to this profile in order to not exceed the resource capacity. These methods are sometimes called *time tabling*. Even if these methods are quite local, i.e., a task has a non-empty compulsory part only when the difference between its latest start and its earliest start is strictly less than its duration, it scales well and is therefore widely used. Later on, more global algorithms⁴ based on the resource consumption of the tasks on specific intervals were introduced [135, 94, 236]. A popular variant, called *edge finding*, considers only specific intervals [254]. An efficient implementation of edge finding in $O(kn \log n)$, where k is the number of distinct task heights and n is the number of tasks, based on a specific data structure, so called a *cumulative* Φ -tree [396], is provided in [395]. A $O(n^2 \log n)$ filtering algorithm based on tasks that can not be the earliest (or not be the latest) is described in [341].

Within the context of linear programming, the reference [191] provides a relaxation of the cumulative constraint.

A necessary condition for the cumulative constraint is obtained by stating a disjunctive constraint on a subset of tasks \mathcal{T} such that, for each pair of tasks of \mathcal{T} , the sum of the two corresponding minimum heights is strictly greater than LIMIT. This can be done by applying the following procedure:

- Let h be the smallest minimum height strictly greater than
 <u>LIMIT</u>

 of the tasks of the cumulative constraint. If no such task exists then the procedure is stopped without stating any disjunctive constraint.
- Let \mathcal{T}_h denote the set of tasks of the cumulative constraint for which the minimum height is greater than or equal to h. By construction, the tasks of \mathcal{T}_h cannot overlap. But we can eventually add one more task as shown by the next step.
- When it exists, we can add one task that does not belong to T_h and such that its minimum height is strictly greater than LIMIT -h. Again, by construction, this task cannot overlap all the tasks of T_h .

When the tasks are involved in several cumulative constraints more sophisticated methods are available for extracting disjunctive constraints [16, 15].

In the context where, both the duration and height of all the tasks are fixed, [33] provides two kinds of additional filtering algorithms that are specially useful when the slack σ (i.e., the difference between the available space and the sum of the surfaces of the tasks) is very small:

- The first one introduces bounds for the so called *cumulative longest hole problem*. Given an integer ε that does not exceed the resource limit, and a subset of tasks *T'* for which the resource consumption is a most ε, the *cumulative longest hole problem* is to find the largest integer *lmax*^ε_σ(*T'*) such that there is a cumulative placement of maximum height ε involving a subset of tasks of *T'* where, on one interval [*i*, *i* + *lmax*^ε_σ(*T'*) − 1] of the cumulative profile, the area of the empty space does not exceed σ.
- The second one used dynamic programming for filtering so called *balancing knapsack constraints*. When the slack is 0, such constraints express the fact that the total height of tasks ending at instant *i* must equal the total height of tasks starting at instant *i*. Such constraints can be generalized to non-zero slack.

⁴Even if these more global algorithms usually can prune more early in the search tree, these algorithms do not catch all deductions derived from the cumulated resource profile of compulsory parts.

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Systems

See also

cumulativeMax in Choco, cumulative in Gecode, cumulative in JaCoP, cumulative in SICStus.

assignment dimension added: coloured_cumulatives (sum of task heights replaced by number of distinct colours, assignment dimension added), cumulatives (negative heights allowed and assignment dimension added).

common keyword: coloured_cumulative (resource constraint, sum of task heights replaced by number of distinct values), coloured_cumulatives (resource constraint), cumulative_convex (resource constraint, task defined by a set of points), cumulative_product (resource constraint, sum of task heights replaced by product of task heights), cumulative_with_level_of_priority (resource constraint, a cumulative constraint for each set of tasks having a priority less than or equal to a given threshold).

generalisation: cumulative_two_d (task replaced by rectangle with a height).

implied by: diffn(cumulative is a neccessary condition for each dimension of the diffn constraint).

related: lex_chain_less, lex_chain_lesseq(lexicographic ordering on the origins of tasks, rectangles, ...), ordered_global_cardinality (controlling the shape of the cumulative profile for breaking symmetry).

soft variant: soft_cumulative.

specialisation: atmost (task *replaced by* variable), bin_packing (all tasks have a duration of 1 and a fixed height), disjunctive (all tasks have a height of 1).

used in graph description: sum_ctr.

Keywords characteristic of a constraint: core, automaton, automaton with array of counters.

complexity: sequencing with release times and deadlines.

constraint type: scheduling constraint, resource constraint, temporal constraint.

filtering: linear programming, dynamic programming, compulsory part, cumulative longest hole problems, Phi-tree.

modelling: zero-duration task.

problems: producer-consumer.

puzzles: squared squares.

Arc input(s)	TASKS		
Arc generator	$SELF \mapsto \texttt{collection}(\texttt{tasks})$		
Arc arity	1		
Arc constraint(s)	tasks.origin + tasks.duration = tasks.end		
Graph property(ies)	NARC= TASKS		
Arc input(s)	TASKS TASKS		
Arc generator	$PRODUCT \mapsto \texttt{collection}(\texttt{tasks1},\texttt{tasks2})$		
Arc arity	2		
Arc constraint(s)	 tasks1.duration > 0 tasks2.origin ≤ tasks1.origin tasks1.origin < tasks2.end 		
Graph class	• ACYCLIC • BIPARTITE • NO_LOOP		
Sets	$ \left[\begin{array}{c} \text{SUCC} \mapsto \\ \text{source,} \\ \text{variables} - \text{col} \left(\begin{array}{c} \text{VARIABLES} - \text{collection}(\text{var} - \text{dvar}), \\ [\text{item}(\text{var} - \text{TASKS.height})] \end{array} \right) \end{array} \right] $		
Constraint(s) on sets	$\texttt{sum_ctr}(\texttt{variables}, \leq, \texttt{LIMIT})$		
Graph model	The first graph constraint enforces for each task the link between its origin, its duration and its end. The second graph constraint makes sure, for each time point t corresponding to the start of a task, that the cumulated heights of the tasks that overlap t does not exceed the limit of the resource.		
	Parts (A) and (B) of Figure 5.166 respectively show the initial and final graph associated with the second graph constraint of the Example slot. On the one hand, each source vertex of the final graph can be interpreted as a time point. On the other hand the successors of a source vertex correspond to those tasks that overlap that time point. The cumulative constraint holds since for each successor set S of the final graph the sum of the heights of the tasks in S does not exceed the limit LIMIT = 8.		

SignatureSince TASKS is the maximum number of vertices of the final graph of the first graph con-
straint we can rewrite NARC = |TASKS| to $NARC \ge |TASKS|$. This leads to simplify
NARC to NARC.



(A)



(B)

Figure 5.166: Initial and final graph of the cumulative constraint





Figure 5.167 depicts the automaton associated with the cumulative constraint. To each

item of the collection TASKS corresponds a signature variable S_i that is equal to 1.

Automaton