5.420 weighted_partial_alldiff

	DESCRIPTION	LINKS	GRAPH
Origin	[406, page 71]		
Constraint	weighted_partial_alldiff(V	ARIABLES, UNDEFINED	, VALUES, COST)
Synonyms	weighted_partial_alldiffer	ent,weighted_parti	al_alldistinct, wpa.
Arguments	VARIABLES : collection(UNDEFINED : int VALUES : collection(COST : dvar		ıt)
Restrictions	<pre>required(VARIABLES, var) VALUES > 0 required(VALUES, [val, weig in_attr(VARIABLES, var, VAL distinct(VALUES, val)</pre>		
Purpose	All variables of the VARIABLES must have pairwise distinct valu addition COST is the sum of the w the variables of VARIABLES. Wi explicitly defined with a weight of	es from the val attributes assoc thin the VALUES collect	te of the VALUES collection. In iated with the values assigned to
Example	$\left(\begin{array}{c} \langle 4,0,1,2,0,0\rangle,0,\\ \mathrm{val}-0\mathrm{weight}-0\\ \mathrm{val}-1\mathrm{weight}-2\\ \langle \mathrm{val}-2\mathrm{weight}-2\\ \mathrm{val}-4\mathrm{weight}-2\\ \mathrm{val}-5\mathrm{weight}-2\\ \mathrm{val}-6\mathrm{weight}-2\\ \end{array}\right)$	/	::
	• No value, except value UND		
	• COST = 8 is equal to the su assigned to the variables of		-1 and 7 of the values 1, 2 and 4 $2, 0, 0\rangle$.
Typical	$\begin{split} \texttt{VARIABLES} &> 0 \\ \texttt{atleast}(1,\texttt{VARIABLES},\texttt{UNDEF} \\ \texttt{VARIABLES} &\leq \texttt{VALUES} + 2 \end{split}$	FINED)	

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Symmetries	 Items of VARIABLES are permutable. Items of VALUES are permutable. 	
	• All occurrences of two distinct values in VARIABLES.var or VALUES.val that are both different from UNDEFINED can be swapped; all occurrences of a value in VARIABLES.var or VALUES.val that is different from UNDEFINED can be renamed to any unused value that is also different from UNDEFINED.	
Arg. properties	Functional dependency: COST determined by VARIABLES and VALUES.	
Usage	In his PhD thesis [406, pages 71–72], Sven Thiel describes the following three potential scenarios of the weighted_partial_alldiff constraint:	
	• Given a set of tasks (i.e., the items of the VARIABLES collection), assign to each task a resource (i.e., an item of the VALUES collection). Except for the resource associated with value UNDEFINED, every resource can be used at most once. The cost of a resource is independent from the task to which the resource is assigned. The cost of value UNDEFINED is equal to 0. The total cost COST of an assignment corresponds to the sum of the costs of the resources effectively assigned to the tasks. Finally we impose an upper bound on the total cost.	
	• Given a set of persons (i.e., the items of the VARIABLES collection), select for each person an offer (i.e., an item of the VALUES collection). Except for the offer associated with value UNDEFINED, every offer should be selected at most once. The profit associated with an offer is independent from the person that selects the offer. The profit of value UNDEFINED is equal to 0. The total benefit COST is equal to the sum of the profits of the offers effectively selected. In addition we impose a lower bound on the total benefit.	
	• The last scenario deals with an application to an over-constraint problem involving the alldifferent constraint. Allowing some variables to take an "undefined" value is done by setting all weights of all the values different from UNDEFINED to 1. As a consequence all variables assigned to a value different from UNDEFINED will have to take distinct values. The COST variable allows to control the number of such variables.	
Remark	It was shown in [406, page 104] that, finding out whether the weighted_partial_alldiff constraint has a solution or not is NP-hard. This was achieved by reduction from subset sum.	
Algorithm	A filtering algorithm is given in [406, pages 73–104]. After showing that, deciding wheth the weighted_partial_alldiff has a solution is NP-complete, [406, pages 105–10 gives the following results of his filtering algorithm with respect to consistency under t 3 scenarios previously described:	
	• For scenario 1, if there is no restriction of the lower bound of the COST variable, the filtering algorithm achieves arc-consistency for all variables of the VARIABLES collection (but not for the COST variable itself).	
	• For scenario 2, if there is no restriction of the upper bound of the COST variable, the filtering algorithm achieves arc-consistency for all variables of the VARIABLES collection (but not for the COST variable itself).	

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		the filtering algorithm achieves arc-consistency for all vari- 5 collection as well as for the COST variable.	
See also	attached to cost variant: al	ldifferent,alldifferent_except_0.	
	<pre>soft_alldifferent_var(s</pre>	global_cardinality_with_costs (weighted assignment), rent (cost filtering constraint, weighted assignment), oft constraint), nct_values (weighted assignment).	
Keywords	application area: assignmen	t.	
	characteristic of a constrain	it: all different, joker value.	
	complexity: subset sum.		
	constraint type: soft constra	int, relaxation.	
	filtering: cost filtering constr	aint.	
	modelling: functional depen	lency.	
	problems: weighted assignm	ent.	

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Arc input(s)	VARIABLES VALUES	
Arc generator	$PRODUCT \mapsto \texttt{collection}(\texttt{variables}, \texttt{values})$	
Arc arity	2	
Arc constraint(s)	 variables.var ≠ UNDEFINED variables.var = values.val 	
Graph property(ies)	 MAX_ID≤ 1 SUM(VALUES, weight) = COST 	

Graph model

Parts (A) and (B) of Figure 5.804 respectively show the initial and final graph associated with the **Example** slot. Since we also use the **SUM** graph property we show the vertices of the final graph from which we compute the total cost in a box.

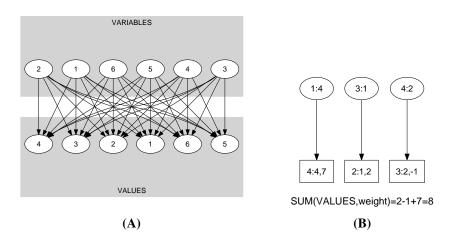


Figure 5.804: Initial and final graph of the weighted_partial_alldiff constraint