

**5.167 global\_cardinality\_with\_costs**

	DESCRIPTION	LINKS	GRAPH
<b>Origin</b>	[344]		
<b>Constraint</b>	global_cardinality_with_costs(VARIABLES, VALUES, MATRIX, COST)		
<b>Synonyms</b>	gcc, cost_gcc.		
<b>Arguments</b>	VARIABLES : collection(var-dvar) VALUES : collection(val-int, noccurrence-dvar) MATRIX : collection(i-int, j-int, c-int) COST : dvar		
<b>Restrictions</b>	<pre> required(VARIABLES, var)  VALUES  &gt; 0 required(VALUES, [val, noccurrence]) distinct(VALUES, val) VALUES.noccurrence ≥ 0 VALUES.noccurrence ≤  VARIABLES  required(MATRIX, [i, j, c]) increasing_seq(MATRIX, [i, j]) MATRIX.i ≥ 1 MATRIX.i ≤  VARIABLES  MATRIX.j ≥ 1 MATRIX.j ≤  VALUES   MATRIX  =  VARIABLES  *  VALUES  </pre>		
<b>Purpose</b>	<p>Each value <code>VALUES[i].val</code> should be taken by exactly <code>VALUES[i].noccurrence</code> variables of the <code>VARIABLES</code> collection. In addition the <code>COST</code> of an <a href="#">assignment</a> is equal to the sum of the elementary costs associated with the fact that we assign variable <math>i</math> of the <code>VARIABLES</code> collection to the <math>j^{th}</math> value of the <code>VALUES</code> collection. These elementary costs are given by the <code>MATRIX</code> collection.</p>		

**Example**

$$\left( \begin{array}{l} \langle 3, 3, 3, 6 \rangle, \\ \left\langle \begin{array}{ll} \text{val} - 3 & \text{noccurrence} - 3, \\ \text{val} - 5 & \text{noccurrence} - 0, \\ \text{val} - 6 & \text{noccurrence} - 1 \end{array} \right\rangle, \\ i - 1 \quad j - 1 \quad c - 4, \\ i - 1 \quad j - 2 \quad c - 1, \\ i - 1 \quad j - 3 \quad c - 7, \\ i - 2 \quad j - 1 \quad c - 1, \\ i - 2 \quad j - 2 \quad c - 0, \\ \left\langle \begin{array}{ll} i - 2 \quad j - 3 \quad c - 8, \\ i - 3 \quad j - 1 \quad c - 3, \\ i - 3 \quad j - 2 \quad c - 2, \\ i - 3 \quad j - 3 \quad c - 1, \\ i - 4 \quad j - 1 \quad c - 0, \\ i - 4 \quad j - 2 \quad c - 0, \\ i - 4 \quad j - 3 \quad c - 6 \end{array} \right\rangle, 14 \end{array} \right)$$

The `global_cardinality_with_costs` constraint holds since:

- Values 3, 5 and 6 respectively occur 3, 0 and 1 times within the collection  $\langle 3, 3, 3, 6 \rangle$ .
- The `COST` argument corresponds to the sum of the costs respectively associated with the first, second, third and fourth items of  $\langle 3, 3, 3, 6 \rangle$ , namely 4, 1, 3 and 6.

**All solutions**

Figure 5.358 gives all solutions to the following non ground instance of the `global_cardinality_with_costs` constraint:

$V_1 \in [3, 4]$ ,  $V_2 \in [2, 3]$ ,  $V_3 \in [1, 2]$ ,  $V_4 \in [2, 4]$ ,  $V_5 \in [2, 3]$ ,  $V_6 \in [1, 2]$ ,  
 $O_1 \in [1, 1]$ ,  $O_2 \in [2, 3]$ ,  $O_3 \in [0, 1]$ ,  $O_4 \in [2, 3]$ ,  
 $C \in [0, 16]$ ,

`global_cardinality_with_costs`( $\langle V_1, V_2, V_3, V_4, V_5, V_6 \rangle$ ,  
 $\langle 1 O_1, 2 O_2, 3 O_3, 4 O_4 \rangle$ ,  
 $\langle 1 1 5, 1 2 0, 1 3 1, 1 4 1,$   
 $2 1 2, 2 2 7, 2 3 0, 2 4 2,$   
 $3 1 3, 3 2 3, 3 3 6, 3 4 6,$   
 $4 1 4, 4 2 3, 4 3 0, 4 4 0,$   
 $5 1 2, 5 2 0, 5 3 6, 5 4 3,$   
 $6 1 5, 6 2 4, 6 3 5, 6 4 4 \rangle, C$ ).

**Typical**

```
|VARIABLES| > 1
range(VARIABLES.var) > 1
|VALUES| > 1
range(VALUES.noccurrence) > 1
range(MATRIX.c) > 1
|VARIABLES| > |VALUES|
```

**Arg. properties**

- **Functional dependency:** `VALUES.noccurrence` determined by `VARIABLES`.
- **Functional dependency:** `COST` determined by `VARIABLES`, `VALUES` and `MATRIX`.

**Usage**

A classical utilisation of the `global_cardinality_with_costs` constraint corresponds to the following [assignment](#) problem. We have a set of persons  $\mathcal{P}$  as well as a set of jobs

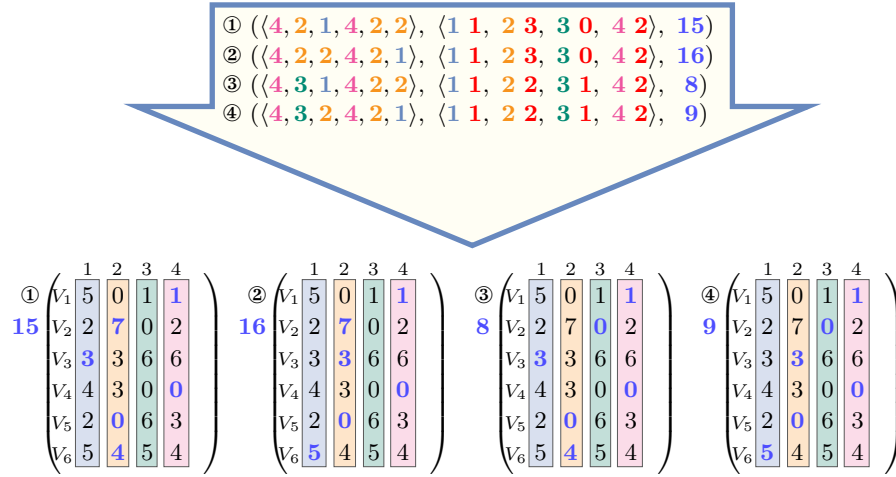


Figure 5.358: All solutions corresponding to the non ground example of the `global_cardinality_with_costs` constraint of the **All solutions** slot

$\mathcal{J}$  to perform. Each job requires a number of persons restricted to a specified interval. In addition each person  $p$  has to be assigned to one specific job taken from a subset  $\mathcal{J}_p$  of  $\mathcal{J}$ . There is a cost  $C_{pj}$  associated with the fact that person  $p$  is assigned to job  $j$ . The previous problem is modelled with a single `global_cardinality_with_costs` constraint where the persons and the jobs respectively correspond to the `VARIABLES` and `VALUES` collection.

The `global_cardinality_with_costs` constraint can also be used for modelling a conjunction `alldifferent`( $X_1, X_2, \dots, X_n$ ) and  $\alpha_1 \cdot X_1 + \alpha_2 \cdot X_2 + \dots + \alpha_n \cdot X_n = \text{COST}$ . For this purpose we set the domain of the `nooccurrence` variables to  $\{0, 1\}$  and the cost attribute `c` of a variable  $X_i$  and one of its potential value  $j$  to  $\alpha_i \cdot j$ . In practice this can be used for the *magic squares* and the *magic hexagon* problems where all the  $\alpha_i$  are set to 1.

#### Algorithm

A filtering algorithm achieving `arc-consistency` independently on each side (i.e., the *greater than or equal to* side and the *less than or equal to* side) of the `global_cardinality_with_costs` constraint is described in [344, 346]. This algorithm assumes for each value a fixed minimum and maximum number of occurrences. If we rather have occurrence variables, the **Reformulation slot** explains how to also obtain some propagation from the cost variable back to the occurrence variables.

#### Reformulation

Let  $n$  and  $m$  respectively denote the number of items of the `VARIABLES` and of the `VALUES` collections. Let  $v_1, v_2, \dots, v_m$  denote the values `VALUES[1].val, VALUES[2].val, \dots, VALUES[m].val`. In addition let  $LINE_i$  (with  $i \in [1, n]$ ) denote the values `MATRIX[m \cdot (i - 1) + 1].c, MATRIX[m \cdot (i - 1) + 2].c, \dots, MATRIX[m \cdot i].c`, i.e., line  $i$  of the matrix `MATRIX`.

By introducing  $2 \cdot n$  auxiliary variables  $U_1, U_2, \dots, U_n$  and  $C_1, C_2, \dots, C_n$ , the `global_cardinality_with_costs`(`VARIABLES`, `VALUES`, `MATRIX`, `COST`) constraint can be expressed in term of the conjunction of one

`global_cardinality`(`VARIABLES`, `VALUES`) constraint,  $2 \cdot n$  `element` constraints and one arithmetic constraint `sum_ctr`.

For each variable  $V_i$  (with  $i \in [1, |\text{VARIABLES}|]$ ) of the `VARIABLES` collection a first `element`( $U_i, \langle v_1, v_2, \dots, v_m \rangle, V_i$ ) constraint provides the correspondence between the variable  $V_i$  and the index of the value  $U_i$  to which it is assigned. A second `element`( $U_i, \text{LINE}_i, C_i$ ) links the previous index  $U_i$  to the cost  $C_i$  variable associated with variable  $V_i$ . Finally the total cost `COST` is equal to the sum  $C_1 + C_2 + \dots + C_n$ .

In the context of the **Example** slot we get the following conjunction of constraints:

```
global_cardinality(⟨3, 3, 6⟩,
  ⟨val - 3 noccurrence - 3,
   val - 5 noccurrence - 0,
   val - 6 noccurrence - 1⟩),
element(1, ⟨3, 5, 6⟩, 3),
element(1, ⟨3, 5, 6⟩, 3),
element(1, ⟨3, 5, 6⟩, 3),
element(3, ⟨3, 5, 6⟩, 6),
element(1, ⟨4, 1, 7⟩, 4),
element(1, ⟨1, 0, 8⟩, 1),
element(1, ⟨3, 2, 1⟩, 3),
element(3, ⟨0, 0, 6⟩, 6),
14 = 4 + 1 + 3 + 6.
```

We now show how to add implied constraints that can also propagate from the cost variable back to the occurrence variables. Let  $O_1, O_2, \dots, O_m$  respectively denote the variables `VALUES[1].noccurrence, VALUES[2].noccurrence, \dots, VALUES[m].noccurrence`.

The idea is to get for each value  $v_i$  (with  $i \in [1, m]$ ) an idea of its minimum and maximum contribution in the total cost `COST` that is linked to the number of times it is assigned to a variables of `VARIABLES`. E.g., if value  $v_i$  (with  $i \in [1, m]$ ) is used twice, then the corresponding minimum (respectively maximum) contribution in the total cost `COST` will be at least equal to the sum of the two smallest (respectively largest) costs attached to row  $i$ . Let  $D_i$  (with  $i \in [1, m]$ ) denotes the contribution that stems from the variables of `VARIABLES` that are assigned value  $v_i$ . For each value  $v_i$  (with  $i \in [1, m]$ ) we create one `element` constraint for linking  $O_i + 1$  to the corresponding minimum contribution  $LOW_i$ . The table of that `element` constraint has  $n + 1$  entries, where entry  $j$  (with  $j \in [0, n]$ ) corresponds to the sum of the  $j^{\text{th}}$  smallest entries of row  $i$  of the cost matrix `MATRIX`. Similarly we create for each value  $v_i$  (with  $i \in [1, m]$ ) one `element` constraint for linking  $O_i + 1$  to the corresponding maximum contribution  $UP_i$ . The table of that `element` constraint also has  $n + 1$  entries, where entry  $j$  (with  $j \in [0, n]$ ) corresponds to the sum of the  $j^{\text{th}}$  largest entries of row  $i$  of the cost matrix `MATRIX`.

In the context of the cost matrix of the **Example** slot we get the following conjunction of implied constraints:

```
COST = D1 + D2 + D3,
n = O1 + O2 + O3,
P1 = O1 + 1,
P2 = O2 + 1,
P3 = O3 + 1,
element(P1, ⟨0, 0, 1, 4, 8⟩, LOW1),
element(P2, ⟨0, 0, 0, 1, 3⟩, LOW2),
```

`element`( $P_3$ ,  $\langle 0, 1, 7, 14, 22 \rangle$ ,  $LOW_3$ ),  
`element`( $P_1$ ,  $\langle 0, 4, 7, 8, 8 \rangle$ ,  $UP_1$ ),  
`element`( $P_2$ ,  $\langle 0, 2, 3, 3, 3 \rangle$ ,  $UP_2$ ),  
`element`( $P_3$ ,  $\langle 0, 8, 15, 21, 22 \rangle$ ,  $UP_3$ ),  
 $LOW_1 \leq D_1, D_1 \leq UP_1$ ,  
 $LOW_2 \leq D_2, D_2 \leq UP_2$ ,  
 $LOW_3 \leq D_3, D_3 \leq UP_3$ .

**Systems** `global_cardinality` in **SICStus**.

**See also** **attached to cost variant:** `global_cardinality` (cost associated with each variable, value pair removed).

**common keyword:** `minimum_weight_alldifferent` (cost filtering constraint, weighted assignment), `sum_of_weights_of_distinct_values`, `weighted_partial_alldiff` (weighted assignment).

**implies:** `global_cardinality`.

**Keywords** **application area:** assignment.

**constraint arguments:** pure functional dependency.

**filtering:** cost filtering constraint.

**heuristics:** regret based heuristics, regret based heuristics in matrix problems.

**modelling:** cost matrix, scalar product, functional dependency.

**problems:** weighted assignment.

**puzzles:** magic square, magic hexagon.

For all items of VALUES:

<b>Arc input(s)</b>	VARIABLES
<b>Arc generator</b>	<code>SELF</code> $\mapsto$ <code>collection</code> (variables)
<b>Arc arity</b>	1
<b>Arc constraint(s)</b>	variables.var = VALUES.val
<b>Graph property(ies)</b>	<code>NVERTEX</code> = VALUES.noccurrence
<hr/>	
<b>Arc input(s)</b>	VARIABLES VALUES
<b>Arc generator</b>	<code>PRODUCT</code> $\mapsto$ <code>collection</code> (variables, values)
<b>Arc arity</b>	2
<b>Arc constraint(s)</b>	variables.var = values.val
<b>Graph property(ies)</b>	<code>SUM_WEIGHT_ARC</code> $\left( \text{MATRIX} \left[ \sum \left( \begin{array}{c} (\text{variables.key} - 1) *  \text{VALUES} , \\ \text{values.key} \end{array} \right) \right].c \right) = \text{COST}$

### Graph model

The first graph constraint forces each value of the VALUES collection to be taken by a specific number of variables of the VARIABLES collection. It is identical to the graph constraint used in the `global_cardinality` constraint. The second graph constraint expresses that the COST variable is equal to the sum of the elementary costs associated with each variable-value assignment. All these elementary costs are recorded in the MATRIX collection. More precisely, the cost  $c_{ij}$  is recorded in the attribute c of the  $((i - 1) \cdot |\text{VALUES}| + j)^{th}$  entry of the MATRIX collection. This is ensured by the `increasing` restriction that enforces the fact that the items of the MATRIX collection are sorted in lexicographically increasing order according to attributes i and j.

Parts (A) and (B) of Figure 5.359 respectively show the initial and final graph associated with the second graph constraint of the **Example** slot.

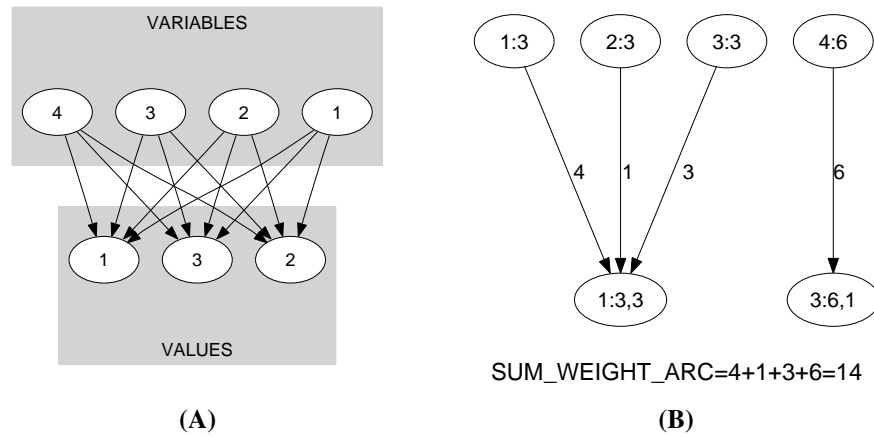


Figure 5.359: Initial and final graph of the `global_cardinality_with_costs` constraint

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