

5.167 **global_cardinality_with_costs**

	DESCRIPTION	LINKS	GRAPH
Origin	[344]		
Constraint	global_cardinality_with_costs(VARIABLES, VALUES, MATRIX, COST)		
Synonyms	gccc, cost_gcc.		
Arguments	VARIABLES : collection(var-dvar) VALUES : collection(val-int, noccurrence-dvar) MATRIX : collection(i-int, j-int, c-int) COST : dvar		
Restrictions	required(VARIABLES, var) VALUES > 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		
Purpose	Each value VALUES[i].val should be taken by exactly VALUES[i].noccurrence variables of the VARIABLES collection. In addition the COST of an assignment is equal to the sum of the elementary costs associated with the fact that we assign variable <i>i</i> of the VARIABLES collection to the <i>jth</i> value of the VALUES collection. These elementary costs are given by the MATRIX collection.		

Example

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

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 5.375 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



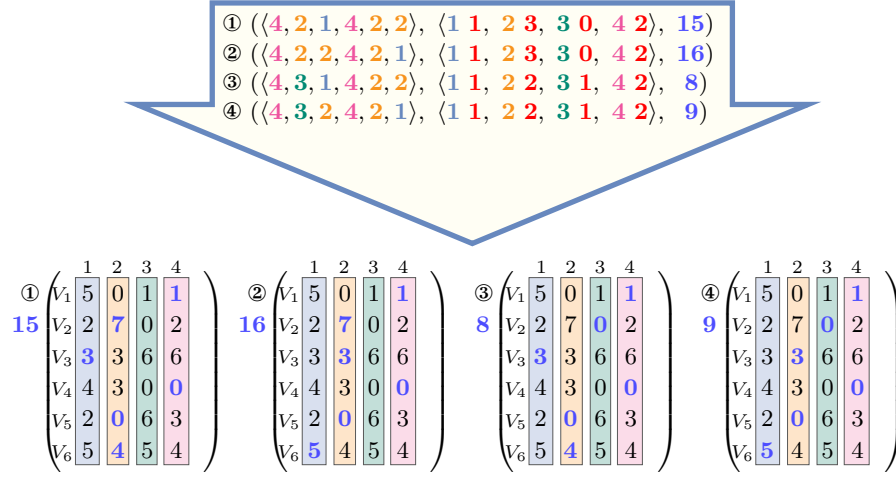


Figure 5.375: All solutions corresponding to the non ground example of the `global_cardinality_with_costs` constraint of the **All solutions** slot

\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 items of 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 `noccurrence` 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 α_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, ..., VALUES[m].val`. In addition let $LINE_i$ (with $i \in [1, n]$) denote the values `MATRIX[m · (i - 1) + 1].c, MATRIX[m · (i - 1) + 2].c, ..., MATRIX[m · 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, 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, 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^{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^{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:

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

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 [5.376](#) respectively show the initial and final graph associated with the second graph constraint of the **Example** slot.

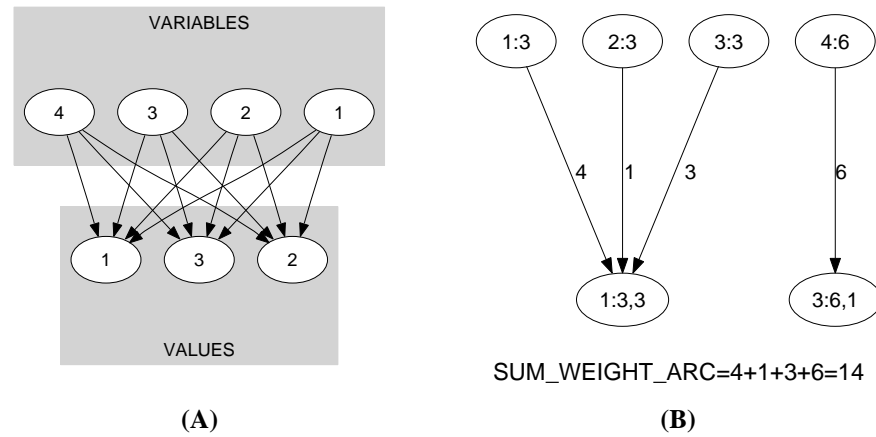


Figure 5.376: Initial and final graph of the `global_cardinality_with_costs` constraint

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