

## 5.221 minimum\_weight\_alldifferent

	DESCRIPTION	LINKS	GRAPH
<b>Origin</b>	[151]		
<b>Constraint</b>	minimum_weight_alldifferent(VARIABLES, MATRIX, COST)		
<b>Synonyms</b>	minimum_weight_alldiff, minimum_weight_alldistinct, min_weight_alldiff, min_weight_alldifferent, min_weight_alldistinct.		
<b>Arguments</b>	VARIABLES : collection(var-dvar) MATRIX : collection(i-int, j-int, c-int) COST : dvar		
<b>Restrictions</b>	VARIABLES  > 0 required(VARIABLES, var) VARIABLES.var ≥ 1 VARIABLES.var <  VARIABLES  required(MATRIX, [i, j, c]) increasing_seq(MATRIX, [i, j]) MATRIX.i ≥ 1 MATRIX.i ≤  VARIABLES  MATRIX.j ≥ 1 MATRIX.j ≤  VARIABLES   MATRIX  =  VARIABLES  *  VARIABLES		
<b>Purpose</b>	All variables of the VARIABLES collection should take a distinct value located within interval [1,  VARIABLES ]. In addition COST is equal to the sum of the costs associated with the fact that we assign value $i$ to variable $j$ . These costs are given by the matrix MATRIX.		

### Example

$$\left( \langle 2, 3, 1, 4 \rangle, \begin{array}{ccc} i-1 & j-1 & c-4, \\ i-1 & j-2 & c-1, \\ i-1 & j-3 & c-7, \\ i-1 & j-4 & c-0, \\ i-2 & j-1 & c-1, \\ i-2 & j-2 & c-0, \\ i-2 & j-3 & c-8, \\ i-2 & j-4 & c-2, \\ i-3 & j-1 & c-3, \\ i-3 & j-2 & c-2, \\ i-3 & j-3 & c-1, \\ i-3 & j-4 & c-6, \\ i-4 & j-1 & c-0, \\ i-4 & j-2 & c-0, \\ i-4 & j-3 & c-6, \\ i-4 & j-4 & c-5 \end{array} \right), 17$$

The `minimum_weight_alldifferent` constraint holds since the cost 17 corresponds to the sum  $\text{MATRIX}[(1-1) \cdot 4 + 2].c + \text{MATRIX}[(2-1) \cdot 4 + 3].c + \text{MATRIX}[(3-1) \cdot 4 + 1].c + \text{MATRIX}[(4-1) \cdot 4 + 4].c = \text{MATRIX}[2].c + \text{MATRIX}[7].c + \text{MATRIX}[9].c + \text{MATRIX}[16].c = 1 + 8 + 3 + 5$ .

**Algorithm**

The [Hungarian method for the assignment problem](#) [216] can be used for evaluating the bounds of the `COST` variable. A filtering algorithm is described in [343]. It can be used for handling both side of the `minimum_weight_alldifferent` constraint:

- Evaluating a lower bound of the `COST` variable and pruning the variables of the `VARIABLES` collection in order to not exceed the maximum value of `COST`.
- Evaluating an upper bound of the `COST` variable and pruning the variables of the `VARIABLES` collection in order to not be under the minimum value of `COST`.

**Systems**

`assignment` in **SICStus**.

**See also**

**attached to cost variant:** `alldifferent`.

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

**Keywords**

**application area:** `assignment`.

**characteristic of a constraint:** `core`.

**filtering:** `cost filtering constraint`, `Hungarian method for the assignment problem`.

**final graph structure:** `one_succ`.

**modelling:** `cost matrix`.

**problems:** `weighted assignment`.

<b>Arc input(s)</b>	VARIABLES
<b>Arc generator</b>	<code>CLIQUE</code> $\mapsto$ <code>collection</code> (variables1, variables2)
<b>Arc arity</b>	2
<b>Arc constraint(s)</b>	variables1.var = variables2.key
<b>Graph property(ies)</b>	<ul style="list-style-type: none"> <li>• <b>NTREE</b> = 0</li> <li>• <b>SUM_WEIGHT_ARC</b> <math>\left( \text{MATRIX} \left[ \sum \left( \begin{array}{l} (\text{variables1.key} - 1) *  \text{VARIABLES}  \\ \text{variables1.var} \end{array} \right) \right] .c \right) = \text{COST}</math></li> </ul>

**Graph model**

Since each variable takes one value, and because of the arc constraint `variables1 = variables.key`, each vertex of the initial graph belongs to the final graph and has exactly one successor. Therefore the sum of the out-degrees of the vertices of the final graph is equal to the number of vertices of the final graph. Since the sum of the in-degrees is equal to the sum of the out-degrees, it is also equal to the number of vertices of the final graph. Since **NTREE** = 0, each vertex of the final graph belongs to a circuit. Therefore each vertex of the final graph has at least one predecessor. Since we saw that the sum of the in-degrees is equal to the number of vertices of the final graph, each vertex of the final graph has exactly one predecessor. We conclude that the final graph consists of a set of vertex-disjoint elementary circuits.

Finally the graph constraint expresses the fact 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  $((i - 1) \cdot |\text{VARIABLES}| + j)^{\text{th}}$  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*.

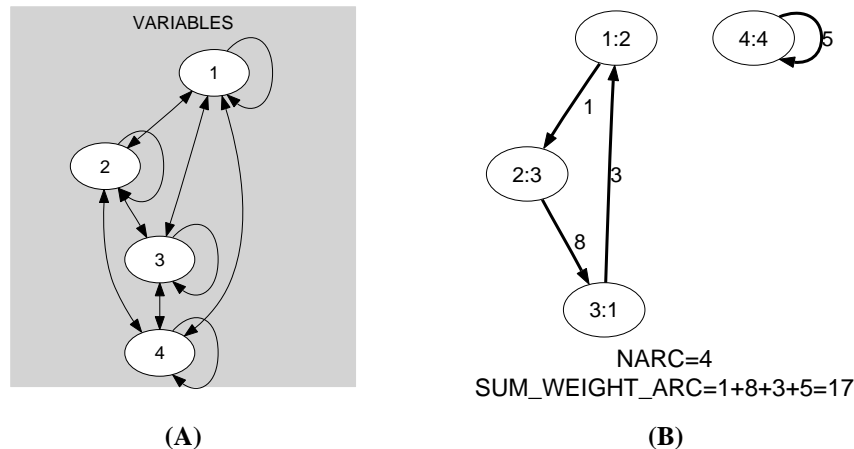


Figure 5.426: Initial and final graph of the `minimum_weight_alldifferent` constraint

Parts (A) and (B) of Figure 5.426 respectively show the initial and final graph associated with the **Example** slot. Since we use the **SUM\_WEIGHT\_ARC** graph property, the

arcs of the final graph are stressed in bold. We also indicate their corresponding weight.