

AUTOMATED REASONING

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- Constraints on **sets** of variables (with some precise meaning) are called **global constraints**
- Often global constraints can be rewritten as combination of binary constraints. However, propagation on these binary constraints is rather poor wrt the constraint viewed as a whole.
- Therefore, global constraints are often studied independently.
- The most famous is `all_different`.

ALL DIFFERENT CONSTRAINT

- Let X_1, \dots, X_k be variables with domains $\mathcal{D}_1, \dots, \mathcal{D}_k$.
- The (k -ary) constraint $\text{all_different}(X_1, \dots, X_k)$ is defined as follows:

$$\text{all_different}(X_1, \dots, X_k) = (\mathcal{D}_1 \times \dots \times \mathcal{D}_k) \setminus \{(a_1, \dots, a_k) \in \mathcal{D}_1 \times \dots \times \mathcal{D}_k : \exists i \exists j 1 \leq i < j \leq k (a_i = a_j)\}$$

- A CSP is said *diff-arc consistent* iff every all_different -constraint in it is *hyper arc consistent*.
- Namely, for every $i \in \{1, \dots, k\}$ and every $a_i \in \mathcal{D}_i$ there are $a_1, \dots, a_{i-1}, a_{i+1}, \dots, a_k$ s.t. $\langle a_1, \dots, a_k \rangle \in \text{all_different}(X_1, \dots, X_k)$

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ALL DIFFERENT CONSTRAINT

- Let us observe that the two CSP $\langle \text{all_different}(X_1, \dots, X_k); \mathcal{D}_\epsilon \rangle$ and $\langle X_1 \neq X_2, X_1 \neq X_3, \dots, X_1 \neq X_k, X_2 \neq X_3, \dots, X_{k-1} \neq X_k; \mathcal{D}_\epsilon \rangle$ are equivalent.
- hyper-arc-consistency of $\text{all_different}(X_1, \dots, X_k)$ implies (binary) arc consistency in the second CSP.
- The converse does not hold:
 $\langle \text{all_different}(X_1, X_2, X_3); \mathcal{D}_1 = \mathcal{D}_2 = \mathcal{D}_3 = \{0, 1\} \rangle$ vs
 $\langle X_1 \neq X_2, X_1 \neq X_3, X_2 \neq X_3; \mathcal{D}_1 = \mathcal{D}_2 = \mathcal{D}_3 = \{0, 1\} \rangle$.
- Let $d_i = |\mathcal{D}_i|$ for $i \in \{1, \dots, k\}$ and $d = \max_{i=1}^k \{d_i\}$.
- A propagation algorithm for hyper-arc-consistency based on the definition has cost $O(\dots d^{k+1})$.
- Not applicable with large k . But is it an intrinsic problem or just a too naive algorithm?

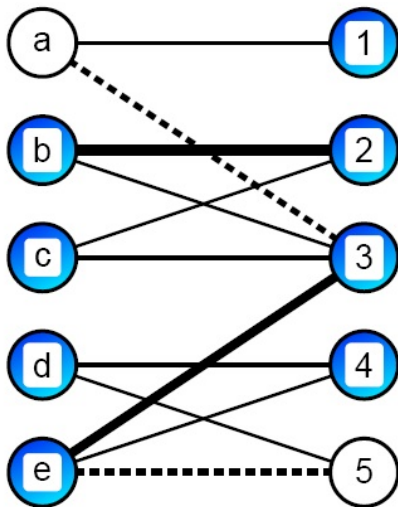
ALL DIFFERENT CONSTRAINT

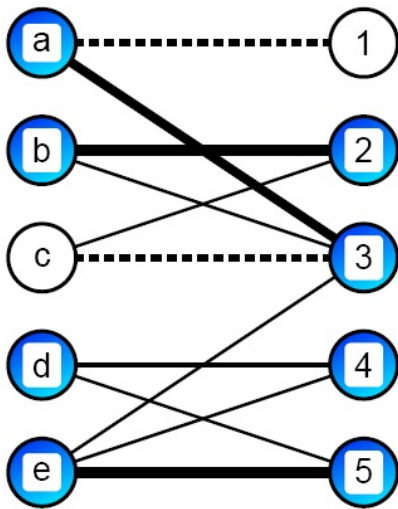
- Let us observe that the two CSP $\langle \text{all_different}(X_1, \dots, X_k); \mathcal{D}_\infty \rangle$ and $\langle X_1 \neq X_2, X_1 \neq X_3, \dots, X_1 \neq X_k, X_2 \neq X_3, \dots, X_{k-1} \neq X_k; \mathcal{D}_\infty \rangle$ are equivalent.
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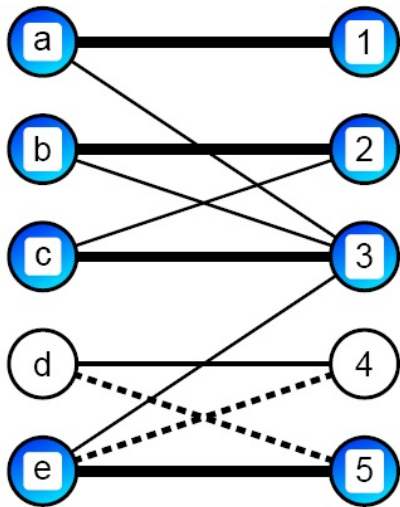
Propagation of all_different

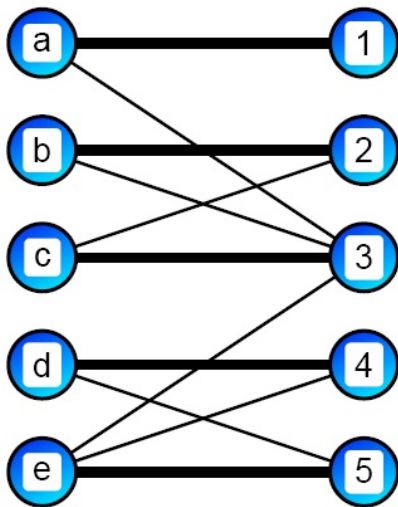
- An *bipartite graph* is a triple $G = \langle X, Y, E \rangle$ where X and Y are disjoint sets of nodes and $E \subseteq X \times Y$ is a set of edges
- Edges are treated as not directed
- A *matching* $M \subseteq E$ is a set of edges such that there are no pairs of edges that share the same node.
- Given G and M a node is said *matched* if it is in some edge in M ; otherwise it is *free*.
- A *path* is a sequence of edges $(x_1, y_1), (y_1, x_2), (x_2, y_2), \dots$

- Given $G = \langle X, Y, E \rangle$ bipartite graph and $M \subseteq E$ matching, a path in G is *alternating* for M if edges of the path are alternatively in M and not in M .
- An alternating path is *augmenting* for M if it is *acyclic* and **starts** and **ends** in free nodes.
- Observe that every augmenting path starting in a node in X ends in Y (or vice versa).
- If $M = \emptyset$, any (set containing a single) edge is an augmenting path for M .









BIPARTITE GRAPHS

RESULTS

PROP. Given a matching M and an augmenting path P for M , we have that: $M' = M \oplus P = (M \setminus P) \cup (P \setminus M)$ is a matching such that $|M'| = |M| + 1$.

Proof.

Let $M = \{(x_1, y_1), \dots, (x_k, y_k),$ The part of M in the path
 $(x_{k+1}, y_{k+1}), \dots, (x_n, y_n)\}$

Suppose P is the augmenting path: $P =$

$\{(y_0, x_1), (x_1, y_1), (y_1, x_2), (x_2, y_2), \dots, (y_{k-1}, x_k), (x_k, y_k), (y_k, x_0)\}$

(wlog, $y_0 \in Y$ and $x_0 \in X$ and the path uses the 'first' k edges of the set M).

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M' is a matching: $x_1, \dots, x_n, y_1, \dots, y_n$ are all different (M is a matching and x_0 and y_0 are different since free by hypothesis).

Its length is $|M| + 1$.

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MAXIMUM MATCHING

- We are interested in matching of maximum size *maximum bipartite matching*.
- Of course the size of the maximum matching is $\leq \min\{|X|, |Y|\}$.
- **Theorem** (Berge–1957). Given a bipartite graph $G = \langle X, Y, E \rangle$, (M is a maximum matching) if and only if (there are not augmenting paths for M).
- (\rightarrow) To show $A \rightarrow B$ we prove $\neg B \rightarrow \neg A$.
Suppose there is an augmenting path for M .
Then (previous Prop) we can define a matching M' of size $|M| + 1$, thus M is not of maximum size.

MAXIMUM MATCHING

BERGE'S THEOREM

- (\leftarrow) To show $A \leftarrow B$ we prove that $\neg A \rightarrow \neg B$. Namely, if M is a matching **not** of maximum size, then there is an augmenting path for M . Let M be a not maximal matching. Then there is a matching M' of bigger size ($|M'| > |M|$). Now let $U = \langle X, Y, M \oplus M' \rangle$.
 - 1 Being M and M' matchings, at most one edge of M and one of M' might incide on a node in U . Thus the degree of a node in U is ≤ 2 .
 - 2 The graph U can have cycles. In this case half of the edges are from M and the other from M' (and the length of the cycle is even).
- Now, removing the cycles, the paths use alternatively edges from M and from M' . Since $|M'| > |M|$ **there must be at least one path with more edges from M' than from M .**
- That path **starts** and **ends** with edges in M' starting and ending with two nodes free for M : There is an augmenting path for M . □

MAXIMUM MATCHING

NAIVE ALGO

Let $n = \min\{|X|, |Y|\}$ and $m = |E|$.

Max_Matching_Naive($\langle X, Y, E \rangle$)

- 1 $M \leftarrow \emptyset$;
- 2 **while** (there is an augmenting path P for M)
- 3 **do**
- 4 $M \leftarrow M \oplus P$;
- 5 **return** M ;

The algorithm terminates in $\leq n$ iterations.

Let us analyze briefly the cost of each iteration.

MAXIMUM MATCHING

NAIVE ALGO

Find_Augmenting_Path($\langle X, Y, E \rangle, M$)

```
1   $S \leftarrow X; A \leftarrow E;$ 
2  trovato  $\leftarrow$  false;
3  while ( $S$  contains a free node  $\wedge \neg$ trovato)
4  do
5      choose a free node  $x$  in  $S$ ;
6      depth first search of an augmenting path for  $M$  in  $\langle S, Y, A \rangle$ ;
7      let  $E(x)$  be the set of edges visited starting from  $x$ ;
8      if (a path is found)
9          then
10             trovato  $\leftarrow$  true
11         else  $S \leftarrow S \setminus \{x\}; A \leftarrow A \setminus E(x);$ 
12 return trovato/path
```

$O(|E|)$: globally we have $O(nm)$.

Hopcroft-Karp: $O(m\sqrt{n}) = O(n^2\sqrt{n})$.

Applying these results to the propagation of the all_different constraint

all_different CONSTRAINT: BIPARTITE GRAPH

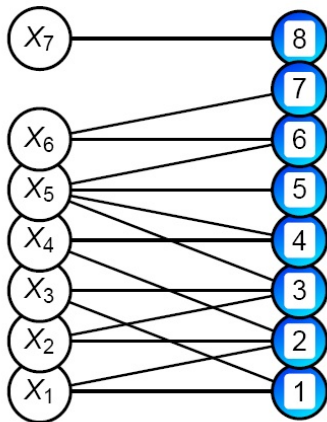
Given $\text{all_different}(X_1, \dots, X_k)$, with domains $\mathcal{D}_1, \dots, \mathcal{D}_k$, let us define the bipartite graph $GV(C) = \langle X_C, Y_C, E_C \rangle$ as follows:

- $X_C = \{X_1, \dots, X_k\}$
- $Y_C = \bigcup_{i=1}^k \mathcal{D}_i$
- $E_C = \{(X_i, a) : a \in \mathcal{D}_i\}$

ALL DIFFERENT CONSTRAINT

EXAMPLE: $\text{all_different}(X_1, \dots, X_7)$

$X_1 \in 1..2, X_2 \in 2..3, X_3 \in \{1, 3\}, X_4 \in \{2, 4\},$
 $X_5 \in 3..6, X_6 \in 6..7, X_7 \in \{8\}$



ALL DIFFERENT CONSTRAINT

REGIN

Theorem: A CSP $\mathcal{P} = (\mathcal{C}; \mathcal{D}_{\in})$ is diff-arc consistent if and only if for all all_different-constraint C in \mathcal{C} any edge in $GV(C)$ belongs to (at least) one matching of the same size of the set of variables of C .

Proof:

(\rightarrow) Let C in \mathcal{C} a all_different-constraint. Let X_1, \dots, X_k be its variables. Let (X_i, a_i) in $GV(C)$ (this implies that $a_i \in \mathcal{D}_i$). Since \mathcal{P} is diff-arc consistent, C is hyper-arc consistent. Thus, there are $a_1, \dots, a_{i-1}, a_{i+1}, \dots, a_k$ such that $X_1 = a_1, \dots, X_k = a_k$ is a solution of C . This solution is a matching of the desired size to which the edge (X_i, a_i) belongs.

(\leftarrow) For any all_different-constraint C (assume it has k vars) consider an edge (X_i, a_i) belonging to a matching of size k . From that matching we obtain the values for the other variables to guarantee the hyper-arc consistency property. \square

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FILTERING

- For building $GV(C)$, we introduce k nodes (for the variables), $|\mathcal{D}_1 \cup \dots \cup \mathcal{D}_k|$ for the domain objects, and $e = d_1 + \dots + d_k \leq kd$ edges.
- A constraint $C = \text{all_different}(X_1, \dots, X_k)$ is hyper-arc consistent if and only if any edge in $GV(C)$ belongs to a matching of size k (maximum).
- We know that we can find **ONE** maximum matching in time $O(\sqrt{ke}) = O(k^{3/2}d)$.
- We are interested in edges that belong to **at least one** maximum matching.
- We need a new idea here!

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BERGE 2

Teorema: [Berge–1970] Let $G = \langle X, Y, E \rangle$ be a bipartite graph. An edge e belongs to **some but not to all** maximum matchings if and only if **for an arbitrary** maximum matching M , e belongs to:

- an **acyclic** even length alternating path that starts in a free vertex
OR
- an alternating cycle (of course of even length).

Proof. (\Leftarrow) Let M be a maximum matching.

- 1 Let P an alternating even length path with a free extreme (the other is not free). Let $M' = M \oplus P$ another matching with the same size. Half of the edges of the path P are in M , the other half in M' .
- 2 Let P be an alternating cycle. Similar.

ALL DIFFERENT CONSTRAINT

An edge e belongs to some but not to all maximum matchings \Rightarrow for an arbitrary maximum matching M , e belongs to: an acyclic even length alternating path that starts in a free vertex OR an alternating cycle (of course of even length).

Let (x_0, x_1) be an edge that belongs to some but not to all maximum matchings.

Let M a generic max matching s.t. $(x_0, x_1) \in M$ and M' a generic max matching s.t. $(x_0, x_1) \notin M$.

By hypothesis, at least one of such M and M' exists.

Now, let $M'' = M \oplus M'$. We have that $(x_0, x_1) \in M''$.

We already know that the degree of each node in M'' is ≤ 2 .

Iteratively choose nodes $x_2, x_3, x_4, x_5, \dots, x_m$ such that $(x_i, x_{i+1}) \in M''$ and $x_{i+2} \neq x_i$.

Stop when there are no longer nodes to be chosen or when x_m has been previously chosen.

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Two cases:

- A cycle has been pointed out (i.e., $x_m = x_1$)
- The sequence is not a cycle. In this case, go back to x_0 and choose a new sequence (that, in a sense, lead to x_0) introducing nodes $x_{-1}, x_{-2}, \dots, x_{-t}$ such that $(x_{i-1}, x_i) \in M''$. Also in this case:
 - A cycle has been generated (but this is not possible . . . why?)
 - No cycle has been generated.

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An edge e belongs to some but not to all maximum matchings \Rightarrow for an arbitrary maximum matching M , e belongs to: an acyclic even length alternating path that starts in a free vertex OR an alternating cycle (of course of even length).

If there are not cycles, then we have an alternating path both for M and for M' . If the path would be of odd length then it would be augmenting for one of them that would contradict the maximality of M or M' . Then the path is even.

In the case there is a cycle, there are two cases. Either (x_0, x_1) is in the path or not. Assume it is in the path. then there is a node of degree 3 in M'' : this is absurdum.

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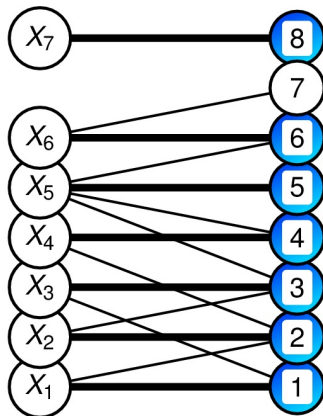
ALL DIFFERENT CONSTRAINT

EFFICIENT FILTERING

- Find a maximal matching M (time $O(k^{3/2}d)$). If the size is less than k stop with unsat.
- For every free node, find the even alternating paths. All nodes reached and edges visited are retained.
- All cycles are detected (algorithm for strongly connected components)
- All edges in M outside these visits should be in all matchings.
- All edges outside M outside these visits are removed.
- Global cost: $O(k^{3/2}d)$
- Data structure is retained to speed up further visits after some labelings.

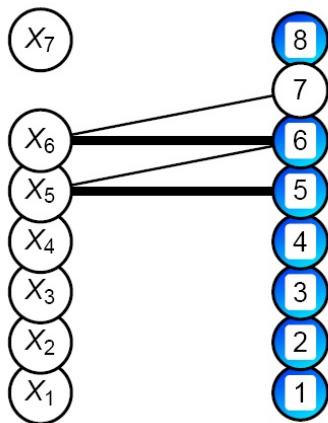
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