

SMT solver for theory of all-different

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Workshop on Formal and Automated Theorem Proving and
Applications.
Belgrade 2009.

Outline

- 1 Introduction to all-different constraint
- 2 Representation in first-order logic
- 3 Our all-different *SMT* solver
- 4 Future work and conclusions

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Definition of all-different constraint

Definition

Given a set of variables x_1, x_2, \dots, x_n , where each variable x_i takes values from its corresponding finite domain $D(x_i)$, then $\text{alldiff}(x_1, x_2, \dots, x_n)$ means that every two different variables must take different values ($i \neq j \Rightarrow x_i \neq x_j$).

Applications

Broad variety of all-different based problems can be reduced to the *SAT* problem, using the *SMT* approach:

- **Puzzle solving** (Sudoku, Latin Square, Eight Queens).
- **Scheduling** and **timetabling**.

Example of all-different based problem

Example (Latin square 5×5)

	3	4		
3	4	5		
4	5			
5				

Notes

- Each cell should be **filled** with a value from 1 to 5.
- Each **row** and each **column** is constrained by **all-different** constraint.
- Some values are **imposed**.

Example of all-different based problem

Example (Latin square 5×5)

1	2	3	4	5
2	3	4	5	1
3	4	5	1	2
4	5	1	2	3
5	1	2	3	4

Notes

- Each cell should be **filled** with a value from 1 to 5.
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- Some values are **imposed**.

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Representation of problem

$$\begin{aligned}
 (x_{11} = 1 \vee x_{11} = 2 \vee \dots \vee x_{11} = 5) & \wedge \\
 (x_{12} = 1 \vee x_{12} = 2 \vee \dots \vee x_{12} = 5) & \wedge \\
 \dots & \wedge \\
 (x_{55} = 1 \vee x_{55} = 2 \vee \dots \vee x_{55} = 5) & \wedge \\
 alldiff(x_{11}, x_{12}, x_{13}, x_{14}, x_{15}) & \wedge \\
 alldiff(x_{21}, x_{22}, x_{23}, x_{24}, x_{25}) & \wedge \\
 \dots & \wedge \\
 alldiff(x_{15}, x_{25}, x_{35}, x_{45}, x_{55}) & \wedge \\
 x_{51} = 5 & \wedge \\
 x_{41} = 4 & \wedge \\
 \dots & \wedge \\
 x_{23} = 4 &
 \end{aligned}$$

Notes

- For each **cell** one **variable** is introduced.

Representation of problem

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Notes

- For each **cell** one **variable** is introduced.
- For each **variable** one clause defines its **domain**.

Representation of problem

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Notes

- For each **cell** one **variable** is introduced.
- For each **variable** one clause defines its **domain**.
- For each **row** and each **column** – unit clause with **all-different**.

Representation of problem

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 & \dots \wedge \\
 x_{23} = 4 & \wedge
 \end{aligned}$$

Notes

- For each **cell** one **variable** is introduced.
- For each **variable** one clause defines its **domain**.
- For each **row** and each **column** – **unit clause** with **all-different**.
- For each **imposed** value – **unit clause** with **equality**.

Representation of solution

$$\begin{array}{l} x_{11} = 1 \wedge x_{12} = 2 \wedge \dots \wedge x_{15} = 5 \quad \wedge \\ x_{21} = 2 \wedge x_{22} = 3 \wedge \dots \wedge x_{25} = 1 \quad \wedge \\ \dots \quad \wedge \\ x_{51} = 5 \wedge \dots \wedge x_{52} = 1 \wedge x_{55} = 4 \end{array}$$

Notes

- **Solution** of the problem can be represented as a **model** of the corresponding formula.

Representation of solution

$$\begin{aligned}x_{11} = 1 \wedge x_{12} = 2 \wedge \dots \wedge x_{15} = 5 & \wedge \\x_{21} = 2 \wedge x_{22} = 3 \wedge \dots \wedge x_{25} = 1 & \wedge \\& \dots \wedge \\x_{51} = 5 \wedge \dots \wedge x_{52} = 1 \wedge x_{55} = 4 & \end{aligned}$$

Notes

- **Solution** of the problem can be represented as a **model** of the corresponding formula.
- **Model** of the formula is an **assignment** of values to variables which **satisfies** all the formula's clauses.

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About *SMT* solvers

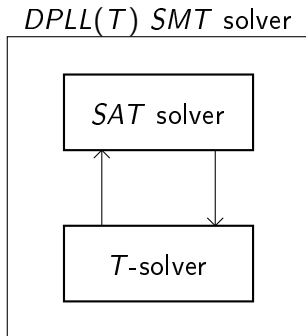
What is an *SMT* solver?

SMT solver checks for satisfiability of **quantifier-free first-order formula** with respect to some **background theory**.

“Lazy” approach

- Each atom in the formula is considered as a **propositional** symbol.
- *SAT* solver checks for **propositional model** of the formula.
- **Propositional model** is checked for satisfiability in the **theory**.

DPLL-based SMT solver structure



Notes

- *SAT* solver:
 - incrementally **builds** partial model
 - checks for **propositional** satisfiability
 - **backtracks** if conflict happens
- *T-solver* (or *Theory* solver):
 - **Detects conflicts** with theory
 - Handles **theory propagations**
 - **Explains** conflicts and/or propagations

Matching theory and all-different

Implementation of all-different T -solver

- T -solver implementation is based on **matching theory** in **bipartite graphs**.
- One **bipartite graph** is assigned to each **all-different** atom appearing in the formula.

Matching theory and all-different

x_1

x_2

x_3

Notes

- Each vertex at the **left** side corresponds to one **variable**.

Matching theory and all-different

a

x_1

b

x_2

c

x_3

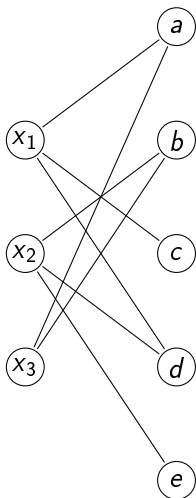
d

e

Notes

- Each vertex at the **left** side corresponds to one **variable**.
- Each vertex at the **right** side corresponds to one **value**.

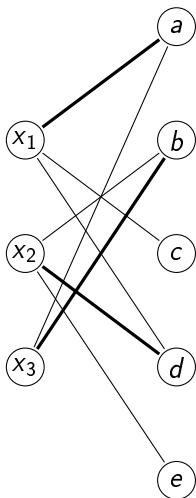
Matching theory and all-different



Notes

- Each vertex at the **left** side corresponds to one **variable**.
- Each vertex at the **right** side corresponds to one **value**.
- Each **variable** is connected to **values** from **its domain**.

Matching theory and all-different



Notes

- Each vertex at the **left** side corresponds to one **variable**.
- Each vertex at the **right** side corresponds to one **value**.
- Each **variable** is connected to **values** from **its domain**.
- **Solution** corresponds to **matching** that **covers left side** vertices.

Conflict detection

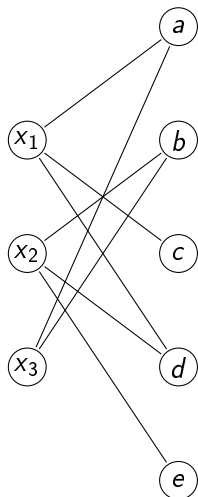
Definition

Given a partial model $\Delta = A_1, A_2, \dots, A_n$, we say that model Δ is **inconsistent** with the theory T if $\Delta \models_T \perp$. We also say that Δ is **in conflict** with the theory T .

Conflict detection in all-different

- **Optimal matching** – matching with **maximal** cardinality.
- all-different is **satisfiable** if and only if optimal matching **covers left side** vertices.

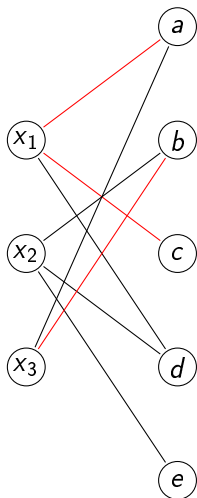
Conflict detection



Notes

- Partial model: $x_1 = d, x_3 \neq b$.

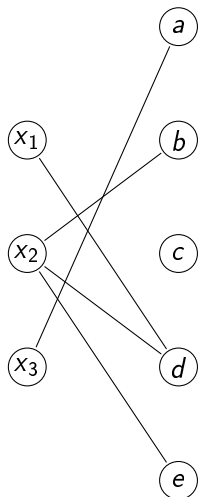
Conflict detection



Notes

- Partial model: $x_1 = d, x_3 \neq b$.
- Edges to remove: $x_1 = a, x_1 = c, x_3 = b$.

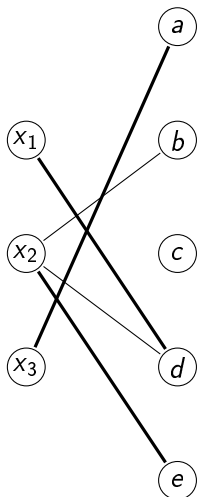
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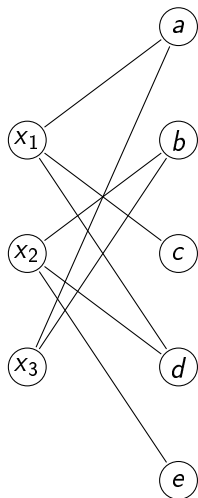
Conflict detection



Notes

- Partial model: $x_1 = d, x_3 \neq b$.
- Edges to remove: $x_1 = a, x_1 = c, x_3 = b$.
- Optimal matching **covers** set of **left vertices**, so all-different constraint is **satisfiable**.

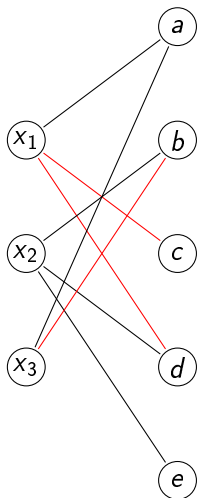
Conflict detection



Notes

- Partial model: $x_1 = a, x_3 \neq b$.

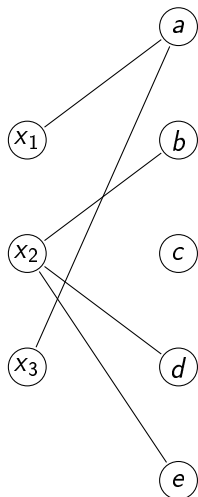
Conflict detection



Notes

- Partial model: $x_1 = a, x_3 \neq b$.
- Edges to remove: $x_1 = c, x_1 = d, x_3 = b$.

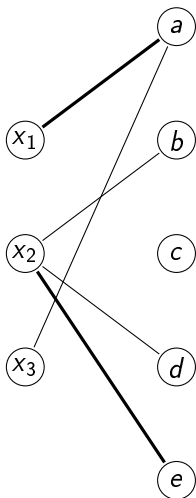
Conflict detection



Notes

- Partial model: $x_1 = a, x_3 \neq b$.
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Conflict detection



Notes

- Partial model: $x_1 = a, x_3 \neq b$.
- Edges to remove: $x_1 = c, x_1 = d, x_3 = b$.
- Optimal matching **doesn't cover** set of **left vertices**, so all-different constraint is **not satisfiable**. **Conflict** is reported!

Conflict detection

Optimal matching construction

- **Hopcroft and Karp's** algorithm (**1973.**).
- The algorithm incrementally augments current matching until optimal matching is constructed.
- The algorithm executes in \sqrt{k} **graph traversals**, where k is cardinality of constructed matching.

Theory propagations

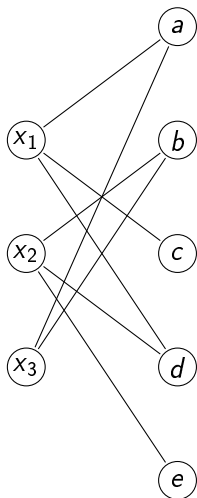
Definition

Given a partial model $\Delta = A_1, A_2, \dots, A_n$, if $\Delta \models_{\mathcal{T}} A$ holds for some atom $A \notin \Delta$, then atom A can be **added** to the partial model. This process is called **theory propagation**.

Theory propagations in all-different

- **Vital edge** – belongs to **all** optimal matchings (**equality propagated**).
- **Inconsistent edge** – doesn't belong to **any** optimal matching (**disequality propagated**).

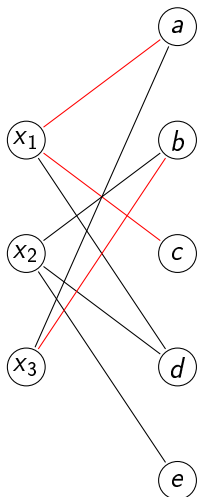
Theory propagations



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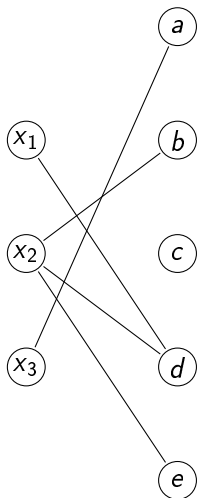
Theory propagations



Notes

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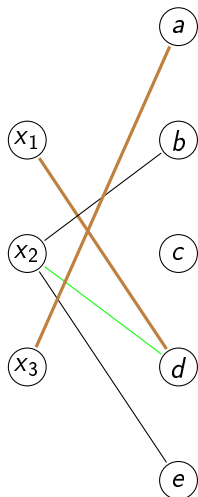
Theory propagations



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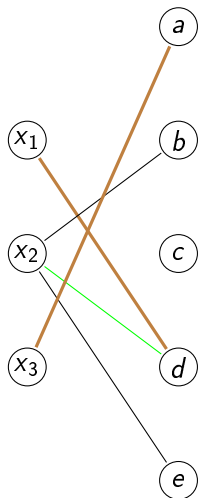
Theory propagations



Notes

- Partial model: $x_1 = d, x_3 \neq b$.
- Edges to remove: $x_1 = a, x_1 = c, x_3 = b$.
- Edges $x_3 = a, x_1 = d$ are vital. Edge $x_2 = d$ is inconsistent.

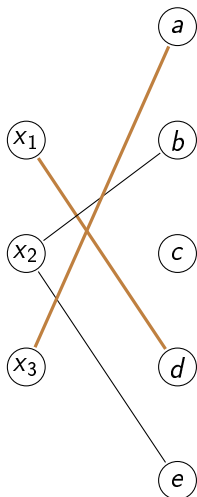
Theory propagations



Notes

- Partial model: $x_1 = d, x_3 \neq b$.
- Edges to remove: $x_1 = a, x_1 = c, x_3 = b$.
- Edges $x_3 = a, x_1 = d$ are vital. Edge $x_2 = d$ is inconsistent.
- Propagation of atoms: $x_3 = a, x_2 \neq d$.

Theory propagations



Notes

- Partial model: $x_1 = a, x_3 \neq b$.
- Edges to remove: $x_1 = a, x_1 = c, x_3 = b$.
- Edges $x_3 = a, x_1 = d$ are vital. Edge $x_2 = d$ is inconsistent.
- Propagation of atoms: $x_3 = a, x_2 \neq d$.
- Inconsistent edges can be removed from the graph.

Theory propagations

Finding vital and inconsistent edges

- **Regin's** filtering algorithm for all-different constraint (**1994.**).
- The algorithm executes in **two graph traversals**.

Theory propagations explaining

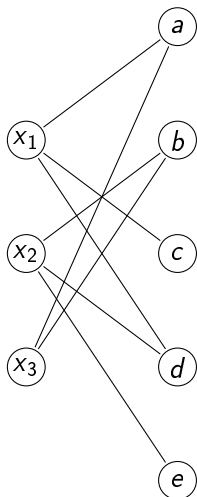
Definition

Given a partial model $\Delta = A_1, A_2, \dots, A_k, \dots, A_n$, if atom A_k is in model as a result of the **theory propagation**, then explanation of atom A_k is any **subset** Σ of A_1, A_2, \dots, A_{k-1} for which $\Sigma \models_T A_k$ holds.

Theory propagations explaining in all-different

Atom A (**equality** or **disequality**) is implied (in all-different theory) by set of atoms Σ **if and only if** the edge corresponding to the atom A is **vital** or **inconsistent** in the graph state corresponding to the model $\Delta = \Sigma$.

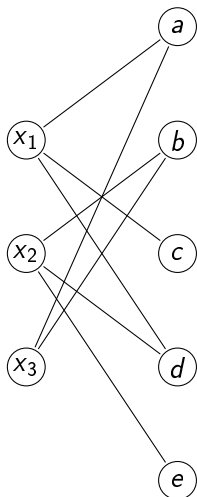
Theory propagations explaining



Notes

- Partial model: $x_1 = d, x_3 \neq b, x_2 \neq e, x_3 = a$.
- Atom $x_3 = a$ has been added to the model by the theory propagation.

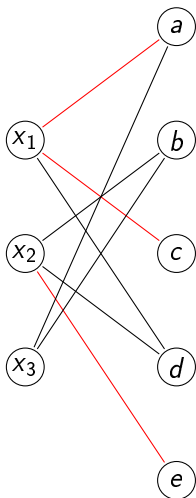
Theory propagations explaining



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- Atom $x_3 = a$ has been added to the model by the theory propagation.
- $x_1 = d, x_2 \neq e \models_T x_3 = a$, because...

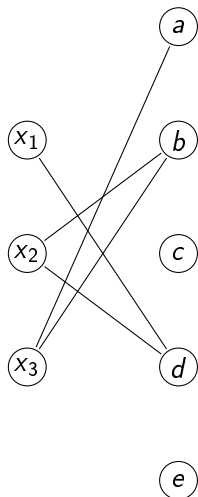
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- Atom $x_3 = a$ has been added to the model by the theory propagation.
- $x_1 = d, x_2 \neq e \models_T x_3 = a$, because... after removal of edges $x_1 = a, x_1 = c, x_2 = e$...

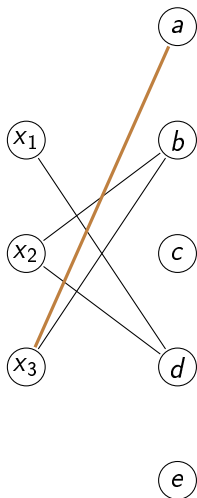
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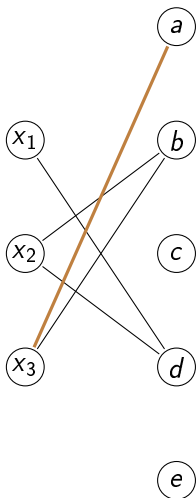
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Notes

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- Atom $x_3 = a$ has been added to the model by the theory propagation.
- $x_1 = d, x_2 \neq e \models_T x_3 = a$, because... after removal of edges $x_1 = a, x_1 = c, x_2 = e$... edge $x_3 = a$ is vital.

Theory propagations explaining



Notes

- Partial model: $x_1 = d, x_3 \neq b, x_2 \neq e, x_3 = a$.
- Atom $x_3 = a$ has been added to the model by the theory propagation.
- $x_1 = d, x_2 \neq e \models_T x_3 = a$, because... after removal of edges $x_1 = a, x_1 = c, x_2 = e$... edge $x_3 = a$ is vital.
- Possible explanation: $x_1 = d, x_2 \neq e$.

Theory propagations explaining

Algorithm for finding minimal explanation

- The algorithm we propose can find explanation which is **minimal in sense of inclusion**.
- The algorithm is based on **Regin's** algorithm.
- Proposed algorithm is very **efficient**, because it executes in **two graph traversals**.

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Future work

Future work

- Our all-different T -solver is planned to be integrated into *ArgoSMT*, which is generic *SMT* platform in early stage of development.
- Possible application: **timetabling** (teaching timetable for Faculty of Mathematics).

Conclusions

Conclusions

- **All-different constraint**: broad application area.
- ***SMT*-approach**: reducing all-different based problems to *SAT*.
- Efficient **decision procedures** based on **matching theory**.
- Theory propagations explaining: **new efficient algorithm is proposed**.