Satisfiability Solving and Tools [original created by Chun-Nan Chou and Ko-Lung Yuan]

Chiao Hsieh

Graduate Institute of Electronics Engineering National Taiwan University

Spring 2015

Outline

- Fundamental Concepts
- 2 Core algorithms of satisfiability problems
- 3 Heuristics
- SAT competitions
- 6 Applications

Outline

- Fundamental Concepts
- Core algorithms of satisfiability problems
- Heuristics
- SAT competitions
- 6 Applications

Boolean Satisfiability Problem(SAT Problem)

- Given a Boolean formula (propositional logic formula), find a variable assignment such that the function evaluates to 1, or prove that no such assignment exists.
 - * EX. $F = (a \lor b) \land (\bar{a} \lor \bar{b} \lor c)$ This function is SAT when a = 1, b = 1, c = 1
- \odot For *n* variables, there are 2^n possible truth assignments to be checked.



- First proofed NP-Complete problem.
 - S. A. Cook, The complexity of theorem proving procedures, Proceedings, Third Annual ACM Symp. on the Theory of Computing, 1971.

Boolean Formula

- There are many ways for representing Boolean function like truth table, Boolean formula, BDD...etc.
- We use Boolean formula when solve SAT problems.
- Boolean variable
 - Boolean variable has two possible value: 0 and 1.
 - 🌞 If a is a Boolean variable, a is also a Boolean formula.
- Boolean formula is constructed by several Boolean formulae with logic connective symbol \lor , \land , and negation. If g and h are Boolean formulae, then so are:
 - \bullet $(g \lor h)$
 - $(g \wedge h)$
 - 🏓 👨

Satisfiable and Unsatisfiable

- Given a Boolean formula F
 - * Unsatisfiable (UNSAT): All assignments let F = 0.
 - * Satisfiable (SAT): there exits one assignment such that F = 1.
 - \bullet Ex1: F = a is satisfiable when a = 1.
 - Ex2: $F = a \wedge b \wedge (\bar{a} \vee \bar{b})$ is unsatisfiable.

Boolean Satisfiability Solvers

- Boolean SAT solvers have been very successful recent years in the verification area.
 - Cooperate with BDDs
 - Applications: equivalence checking and model checking
 - Applicable even for million-gate designs in EDA
- Popular SAT Solvers
 - MiniSat (2008 winner, the most popular one)
 - CryptoMiniSat (2011 winner)

Types of Boolean Satisfiability Solvers

- Conjunctive Normal Form (CNF) Based
 - * A Boolean formula is represented as a CNF (i.e. Product of Sum).
 - * For example: $(a \lor b \lor c) \land (\bar{a} \lor \bar{b} \lor c) \land (\bar{a} \lor b \lor \bar{c})$
 - To be satisfied, all the clauses should be 1.
- Circuit-Based
 - A Boolean formula is represented as a circuit netlist.
 - The SAT algorithm is directly operated on the netlist.

CNF (Conjunction Normal Form)

- Literal is a variable or its negation.
- CNF formula is a conjunction of clauses, where a clause is a disjunction of literals.
- lacktriangle For example, a CNF formula: $(a ee b ee c) \wedge (ar{a} ee ar{b} ee c)$
 - Variable: a, b, c in this CNF formula.
 - Literals: a, \underline{b}, c are literals in $(a \lor \underline{b} \lor c)$.
 - \bullet Literals: \bar{a}, \bar{b}, c are literals in $(\bar{a} \lor \bar{b} \lor c)$.
 - * Clauses: $(a \lor b \lor c)$, $(\bar{a} \lor \bar{b} \lor c)$ are clauses in this CNF formula.

Outline

- Fundamental Concepts
- Core algorithms of satisfiability problems
 - Davis-Putnam Algorithm
 - DPLL Algorithm
 - GRASP Algorithm
 - zChaff Algorithm
- Heuristics
- 4 SAT competitions
- 6 Applications

CNF-Based SAT Algorithms

- 😚 Davis-Putnam (DP), 1960.
 - Explicit resolution based
 - May explode in memory
- 📀 Davis-Putnam-Logemann-Loveland (DPLL), 1962.
 - Search based
 - Most successful, basis for almost all modern SAT solvers
- GRASP, 1996
 - Conflict driven learning and non-chronological backtracking
- zChaff, 2001.
 - Efficient Boolean constraint propagation (BCP) algorithm (two watched literals)

Outline

- Fundamental Concepts
- 2 Core algorithms of satisfiability problems
 - Davis-Putnam Algorithm
 - DPLL Algorithm
 - GRASP Algorithm
 - zChaff Algorithm
- Heuristics
- SAT competitions
- 5 Applications



Davis-Putnam Algorithm

- M. Davis, H. Putnam, "A computing procedure for quantification theory", J. of ACM, 1960. (New York Univ.)
- lacktriangle Three satisfiability-preserving (pprox) transformations in DP:
 - Unit propagation rule
 - Pure literal rule
 - Resolution rule
 - By repeatedly applying these rules, eventually obtain:
 - a formula containing an empty clause indicates unsatisfiability
 - a formula with no clauses indicates satisfiability.
 - No rule can be used and no empty clause existing indicates satisfiability.

Unit Propagation Rule

- \odot Suppose (a) is a unit clause, i.e. a clause contains only one literal.
 - Remove any instances of ā from the formula.
 - Remove all clauses containing a.
- Example:
 - * $(a) \wedge (\bar{a} \vee b \vee c) \wedge (a \vee \bar{b} \vee c) \wedge (\bar{a} \vee \bar{c} \vee d)$ $\approx (b \vee c) \wedge (\bar{c} \vee d)$
 - * (a) \land (a \lor b) \approx satisfiable
 - * (a) \wedge (\bar{a}) \approx () unsatisfiable

Pure Literal Rule

- If a literal appears only positively or only negatively, delete all clauses containing that literal.
- Example: $(\bar{a} \lor b \lor c) \land (\bar{a} \lor \bar{b} \lor c) \land (\bar{b} \lor c \lor d) \land (\bar{a} \lor \bar{c} \lor \bar{d})$ $\approx (\bar{b} \lor c \lor d)$

Resolution Rule

- For a single pair of clauses, $(a \lor l_1 \lor \cdots \lor l_m)$ and $(\bar{a} \lor k_1 \lor \cdots \lor k_n)$, resolution on a forms the new clause $(l_1 \lor \cdots \lor l_m \lor k_1 \lor \cdots \lor k_n)$.
- Example: $(a \lor b) \land (\bar{a} \lor c) \approx (b \lor c)$
 - If a is True, then for the formula to be True, c must be True.
 - If a is False, then for the formula to be True, b must be True.
 - \red So regardless of a, for the formula to be True, $b \lor c$ must be True.

Resolution Rule (cont.)

- Choose a propositional variable p which occurs positively in at least one clause and negatively in at least one other clause.
- ullet Let P be the set of all clauses in which p occurs positively.
- \odot Let N be the set of all clauses in which p occurs negatively.
- Replace the clauses in *P* and *N* with those obtained by resolving each clause in *P* with each clause in *N*.

Example 1

$$(a \lor b) \land (a \lor \bar{b}) \land (\bar{a} \lor c) \land (c \lor d) \land (\bar{a} \lor \bar{c}) \land (d)$$

$$\downarrow Unit \ Propagation \ Rule$$

$$(a \lor b) \land (a \lor \bar{b}) \land (\bar{a} \lor c) \land (\bar{a} \lor \bar{c})$$

$$Resolution \ Rule$$

$$(a) \land (\bar{a} \lor c) \land (\bar{a} \lor \bar{c})$$

$$\downarrow Unit \ Propagation \ Rule$$

$$(c) \land (\bar{c})$$

$$Resolution \ Rule$$

$$(b) \ Unsatisfiable$$

Potential memory explosion problem because of resolution rule

Example 2

- Solve $(a \lor b) \land (a \lor \bar{b}) \land (\bar{a} \lor c) \land (\bar{a} \lor \bar{c})$
- Wrong resolution:

$$(a \lor b) \land (a \lor \overline{b}) \land (\overline{a} \lor c) \land (\overline{a} \lor \overline{c})$$
 Use resolution rule $\approx (b \lor c) \land (\overline{b} \lor \overline{c})$ Use resolution rule $\approx (c \lor \overline{c})$ No rule can be used and no clause is empty! $\approx \mathsf{SAT} \to \mathsf{Wrong}$ result!

- We have to resolve each clause in P with each clause in N.
- Correct resolution:
 - Choose a to do resolution
 - $P = \{(a \lor b), (a \lor \bar{b})\}$
 - $N = \{(\bar{a} \vee c), (\bar{a} \vee \bar{c})\}$
 - $R = \{(b \lor c), (b \lor \bar{c}), (\bar{b} \lor c), (\bar{b} \lor \bar{c})\}$
 - * $(a \lor b) \land (a \lor \bar{b}) \land (\bar{a} \lor c) \land (\bar{a} \lor \bar{c})$ $\approx (b \lor c) \land (b \lor \bar{c}) \land (\bar{b} \lor c) \land (\bar{b} \lor \bar{c})$ Replace P, N with R! $\approx ...$

Outline

- Fundamental Concepts
- Core algorithms of satisfiability problems
 - Davis-Putnam Algorithm
 - DPLL Algorithm
 - GRASP Algorithm
 - zChaff Algorithm
- Heuristics
- SAT competitions
- 6 Applications



DPLL Algorithm

- M. Davis, G. Logemann and D. Loveland, "A Machine Program for Theorem-Proving", Communications of ACM, 1962. (New York Univ.)
- The basic framework for many modern SAT solvers.
- Main strategy
 - Decision Making
 - Unit Clause Rule
 - Implication
 - Conflict Detection
 - Backtracking

DPLL Algorithm

```
DPLL Pseudo Code
Function DPLL(\Phi, A)
    A \leftarrow Unit - Propagation(\Phi, A);
    if A is inconsistent then
         return UNSAT;
    if A assigns a value to every variable then
         return SAT;
    v \leftarrow a variable not assigned a value by A;
    if DPLL(\Phi, A \cup \{ v = \text{False } \}) = SAT
         return SAT;
    else
         return DPLL(\Phi, A \cup \{ v = \text{True } \});
```

- $(\bar{a} \lor b \lor c)$
- $(a \lor c \lor d)$
- $(a \lor c \lor \bar{d})$
- $(a \lor \bar{c} \lor d)$
- $(a \lor \bar{c} \lor \bar{d})$
- $(\bar{b} \vee \bar{c} \vee d)$
- $(\bar{a} \lor b \lor \bar{c})$
- $(\bar{a} \lor \bar{b} \lor c)$

(a)

$$(\bar{a} \lor b \lor c)$$

$$(a \lor c \lor d)$$

$$(a \lor c \lor \bar{d})$$

$$(a \lor \bar{c} \lor d)$$

$$(a \lor \bar{c} \lor \bar{d})$$

$$(\bar{b} \vee \bar{c} \vee d)$$

$$(\bar{a} \lor b \lor \bar{c})$$

$$(\bar{a} \vee \bar{b} \vee c)$$



$$(\bar{a} \lor b \lor c)$$

$$(a \lor c \lor d)$$

$$(a \lor c \lor \bar{d})$$

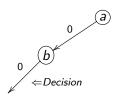
$$(a \lor \bar{c} \lor d)$$

$$(a \lor \bar{c} \lor \bar{d})$$

$$(\bar{b} \vee \bar{c} \vee d)$$

$$(\bar{a} \lor b \lor \bar{c})$$

$$(\bar{a} \vee \bar{b} \vee c)$$



$$(\bar{a} \lor b \lor c)$$
$$(a \lor c \lor d)$$

$$(a \lor c \lor \bar{d})$$

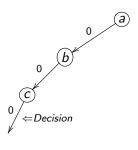
$$(a \lor \bar{c} \lor d)$$

$$(a \lor \bar{c} \lor \bar{d})$$

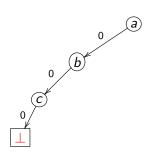
$$(\bar{b} \lor \bar{c} \lor d)$$

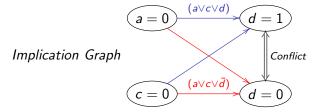
$$(\bar{a} \lor b \lor \bar{c})$$

$$(\bar{a} \vee \bar{b} \vee c)$$

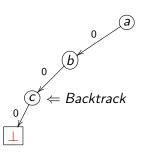


 $(\bar{a} \lor b \lor c)$ $(a \lor c \lor d)$ $(a \lor c \lor \bar{d})$ $(a \lor \bar{c} \lor d)$ $(a \lor \bar{c} \lor \bar{d})$ $(\bar{b} \lor \bar{c} \lor d)$ $(\bar{a} \lor b \lor \bar{c})$ $(\bar{a} \lor \bar{b} \lor c)$





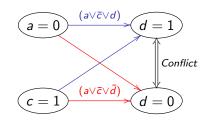
$$\begin{array}{l} (\bar{a}\vee b\vee c)\\ (a\vee c\vee d)\\ (a\vee c\vee \bar{d})\\ (a\vee \bar{c}\vee d)\\ (a\vee \bar{c}\vee \bar{d})\\ (\bar{b}\vee \bar{c}\vee d)\\ (\bar{a}\vee b\vee \bar{c})\\ (\bar{a}\vee \bar{b}\vee c)\\ \end{array}$$



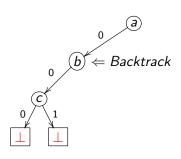
 $(\overline{a} \lor b \lor c)$ $(a \lor c \lor d)$ $(a \lor c \lor \overline{d})$ $(a \lor \overline{c} \lor d)$ $(a \lor \overline{c} \lor \overline{d})$ $(\overline{b} \lor \overline{c} \lor d)$

 $(\bar{a} \lor b \lor \bar{c})$ $(\bar{a} \lor \bar{b} \lor c)$

0
0
0
1=Forced Decision

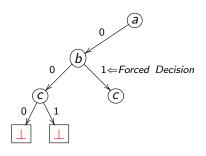


$$\begin{array}{l}
(\bar{a} \lor b \lor c) \\
(a \lor c \lor d) \\
(a \lor c \lor \bar{d}) \\
(a \lor \bar{c} \lor d) \\
(\bar{a} \lor \bar{c} \lor \bar{d}) \\
(\bar{b} \lor \bar{c} \lor d) \\
(\bar{a} \lor b \lor \bar{c}) \\
(\bar{a} \lor \bar{b} \lor c)
\end{array}$$

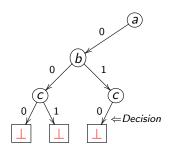


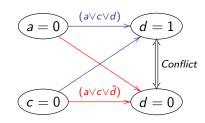
$$\begin{array}{l}
(\bar{a} \lor b \lor c) \\
(a \lor c \lor d) \\
(a \lor \bar{c} \lor \bar{d}) \\
(a \lor \bar{c} \lor \bar{d}) \\
(\bar{b} \lor \bar{c} \lor d) \\
(\bar{a} \lor b \lor \bar{c})
\end{array}$$

 $(\bar{a} \vee \bar{b} \vee c)$

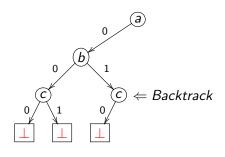


```
\begin{array}{l}
(\bar{a} \lor b \lor c) \\
(a \lor c \lor d) \\
(a \lor c \lor \bar{d}) \\
(a \lor \bar{c} \lor d) \\
(\bar{a} \lor \bar{c} \lor \bar{d}) \\
(\bar{b} \lor \bar{c} \lor d) \\
(\bar{a} \lor b \lor \bar{c}) \\
(\bar{a} \lor \bar{b} \lor c)
\end{array}
```

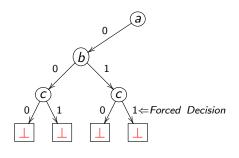


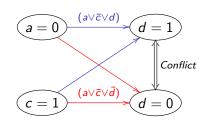


$$\begin{array}{l} (\bar{a}\vee b\vee c)\\ (a\vee c\vee d)\\ (a\vee c\vee \bar{d})\\ (a\vee \bar{c}\vee d)\\ (a\vee \bar{c}\vee \bar{d})\\ (\bar{b}\vee \bar{c}\vee d)\\ (\bar{b}\vee \bar{c}\vee d)\\ (\bar{a}\vee b\vee \bar{c})\\ (\bar{a}\vee \bar{b}\vee c)\\ \end{array}$$

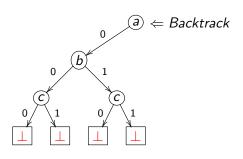


- $(\bar{a} \lor b \lor c)$ $(a \lor c \lor d)$ $(a \lor c \lor \bar{d})$ $(a \lor \bar{c} \lor d)$ $(a \lor \bar{c} \lor \bar{d})$ $(\bar{b} \vee \bar{c} \vee d)$
- $(\bar{a} \lor b \lor \bar{c})$
- $(\bar{a} \vee \bar{b} \vee c)$





$$\begin{array}{l}
(\bar{a} \lor b \lor c) \\
(a \lor c \lor d) \\
(a \lor \bar{c} \lor d) \\
(a \lor \bar{c} \lor d) \\
(\bar{b} \lor \bar{c} \lor d) \\
(\bar{a} \lor b \lor \bar{c}) \\
(\bar{a} \lor \bar{b} \lor c)
\end{array}$$



$$(\bar{a} \lor b \lor c)$$
$$(a \lor c \lor d)$$
$$(a \lor c \lor \bar{d})$$

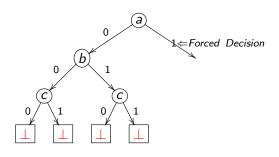
$$(a \lor \bar{c} \lor d)$$

$$(a \lor \bar{c} \lor \bar{d})$$

$$(\bar{b} \lor \bar{c} \lor d)$$

$$(\bar{a} \lor b \lor \bar{c})$$

$$(\bar{a} \lor \bar{b} \lor c)$$



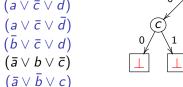
$$(\bar{a} \lor b \lor c)$$

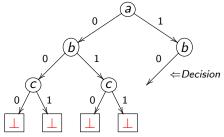
$$(a \lor c \lor d)$$

$$(a \lor c \lor \bar{d})$$

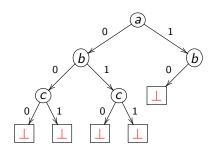
$$(a \lor \bar{c} \lor d)$$

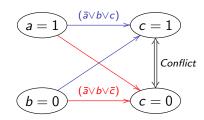
$$(a \lor \bar{c} \lor \bar{d})$$



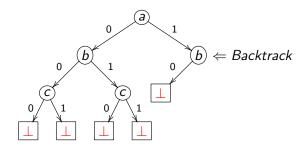


```
\begin{array}{l}
(\bar{a} \lor b \lor c) \\
(a \lor c \lor d) \\
(a \lor \bar{c} \lor d) \\
(a \lor \bar{c} \lor d) \\
(\bar{b} \lor \bar{c} \lor d) \\
(\bar{b} \lor \bar{c} \lor d) \\
(\bar{a} \lor b \lor \bar{c}) \\
(\bar{a} \lor \bar{b} \lor c)
\end{array}
```





$$\begin{array}{l}
(\bar{a} \lor b \lor c) \\
(a \lor c \lor d) \\
(a \lor c \lor d) \\
(a \lor \bar{c} \lor d) \\
(\bar{a} \lor \bar{c} \lor d) \\
(\bar{b} \lor \bar{c} \lor d) \\
(\bar{a} \lor b \lor \bar{c}) \\
(\bar{a} \lor \bar{b} \lor c)
\end{array}$$



$$(\bar{a} \lor b \lor c)$$

$$(a \lor c \lor d)$$

$$(a \lor c \lor d)$$

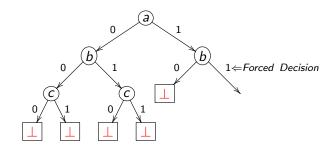
$$(a \lor \bar{c} \lor d)$$

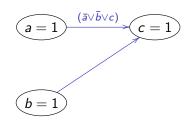
$$(a \lor \bar{c} \lor d)$$

$$(\bar{b} \lor \bar{c} \lor d)$$

$$(\bar{a} \lor b \lor \bar{c})$$

$$(\bar{a} \lor \bar{b} \lor c)$$





$$(\bar{a} \lor b \lor c)$$

$$(a \lor c \lor d)$$

$$(a \lor c \lor d)$$

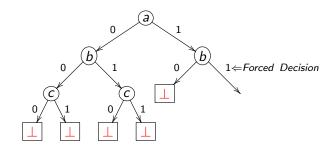
$$(a \lor \bar{c} \lor d)$$

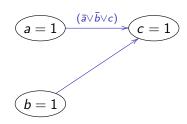
$$(a \lor \bar{c} \lor d)$$

$$(\bar{b} \lor \bar{c} \lor d)$$

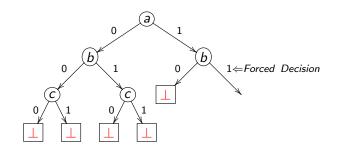
$$(\bar{a} \lor b \lor \bar{c})$$

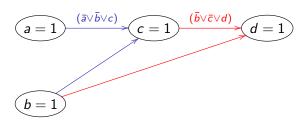
$$(\bar{a} \lor \bar{b} \lor c)$$



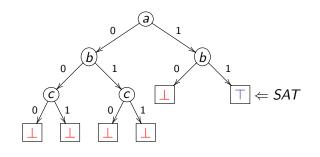


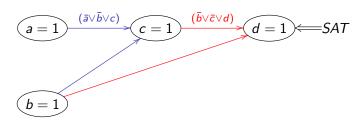
 $\begin{array}{l}
(\bar{a} \lor b \lor c) \\
(a \lor c \lor d) \\
(a \lor c \lor d) \\
(a \lor \bar{c} \lor d) \\
(\bar{a} \lor \bar{c} \lor d) \\
(\bar{b} \lor \bar{c} \lor d) \\
(\bar{a} \lor b \lor \bar{c}) \\
(\bar{a} \lor \bar{b} \lor c)
\end{array}$





 $\begin{array}{l} (\bar{a} \vee b \vee c) \\ (a \vee c \vee d) \\ (a \vee c \vee \bar{d}) \\ (a \vee \bar{c} \vee d) \\ (a \vee \bar{c} \vee \bar{d}) \\ (\bar{b} \vee \bar{c} \vee \bar{d}) \\ (\bar{b} \vee \bar{c} \vee d) \\ (\bar{a} \vee b \vee \bar{c}) \\ (\bar{a} \vee \bar{b} \vee c) \end{array}$





Implications and Unit Clause Rule

- Implication
 - A variable is forced to be True or False based on previous assignments.
- Unit clause rule
 - A rule for elimination of one-literal clauses
 - An unsatisfied clause is a unit clause if it has exactly one unassigned literal.
 - st The only unassigned literal, e.g. \bar{c} , is implied.

$$(a \lor \overline{b} \lor c) \land (b \lor \overline{c}) \land (\overline{a} \lor \overline{c})$$

 $a = T, b = T, c$ is unassigned
Satisfied Literal, Unsatisfied Literal,
Unassigned Literal

Boolean Constraint Propagation

- 😚 Boolean Constraint Propagation (BCP)
 - Iteratively apply the unit clause rule until there is no unit clause available.
 - a.k.a. Unit Propagation
- Workhorse of DPLL based algorithms.

Features of DPLL

- Eliminate the exponential memory requirements of DP
- Exponential time is still a problem
- Limited practical applicability largest use seen in automatic theorem proving
- Very limited size of problems are allowed
 - 32K word memory
 - Problem size limited by total size of clauses (about 1300 clauses)

Outline

- Fundamental Concepts
- Core algorithms of satisfiability problems
 - Davis-Putnam Algorithm
 - DPLL Algorithm
 - GRASP Algorithm
 - zChaff Algorithm
- 3 Heuristics
- SAT competitions
- 6 Applications

GRASP

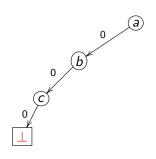
- Marques-Silva and Sakallah [SS96,SS99] (Univ. of Michigan)
 - J. P. Marques-Silva and K. A. Sakallah, "GRASP A New Search Algorithm for Satisfiability", Proc.ICCAD, 1996.
 - J. P. Marques-Silva and Karem A. Sakallah, "GRASP: A Search Algorithm for Propositional Satisfiability", IEEE Trans. Computers, 1999.
- Incorporate conflict driven learning and non-chronological backtracking.
- Practical SAT problem instances can be solved in reasonable time.

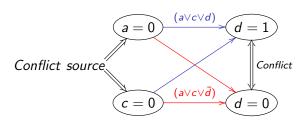
SAT Improvements

- 😚 Conflict driven learning
 - Once we encounter a conflict, figure out the cause(s) of this conflict and prevent to see this conflict again.
 - Add learned clause (conflict clause) which is the negative proposition of the conflict source.
- Non-chronological backtracking
 - After getting a learned clause from the conflict analysis, we backtrack to the "next-to-the-last" variable in the learned clause.
 - Instead of backtracking one decision at a time.

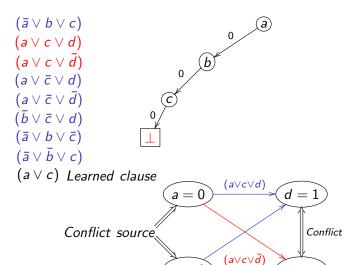
Conflict Driven Learning

 $(\bar{a} \lor b \lor c)$ $(a \lor c \lor d)$ $(a \lor c \lor \bar{d})$ $(a \lor \bar{c} \lor d)$ $(\bar{a} \lor \bar{c} \lor \bar{d})$ $(\bar{b} \lor \bar{c} \lor d)$ $(\bar{a} \lor b \lor \bar{c})$ $(\bar{a} \lor \bar{b} \lor c)$

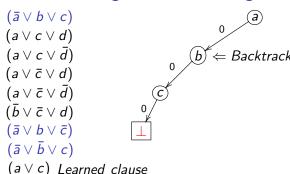




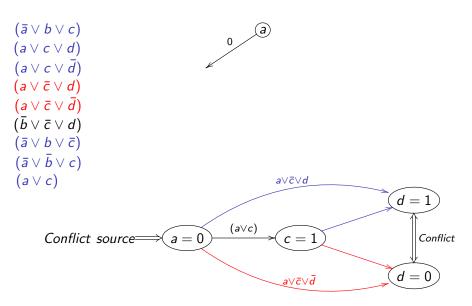
Conflict Driven Learning



c = 0

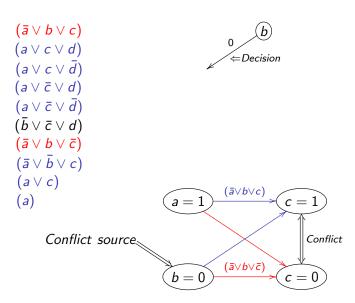


- a v c) Learned Clause
- 😚 'a' is the next-to-the-last variable in the (current) learned clause.
 - c is the last (assigned) variable in this learned clause so a is called the next-to-the-last variable
 - Because of this learned clause, when a is assigned 0 then c will be implied and we don't have to make decision for c
- After doing non-chronological backtracking, we will not forgive the path $a=0,\,b=0...$ if needed.



```
(\bar{a} \lor b \lor c)
(a \lor c \lor d)
(a \lor c \lor \bar{d})
(a \lor \bar{c} \lor d)
(a \lor \bar{c} \lor \bar{d})
(\bar{b} \vee \bar{c} \vee d)
(\bar{a} \lor b \lor \bar{c})
(\bar{a} \vee \bar{b} \vee c)
 (a \lor c)
(a) Learned clause
```

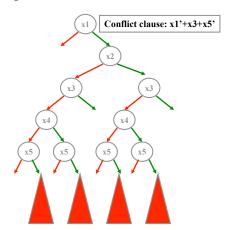
- Since there is only one variable in the learned clause, no one is the next-to-the-last variable.
- Backtrack all decisions



```
(\bar{a} \lor b \lor c)
(a \lor c \lor d)
(a \lor c \lor \bar{d})
(a \lor \bar{c} \lor d)
(a \lor \bar{c} \lor \bar{d})
(\bar{b} \vee \bar{c} \vee d)
(\bar{a} \lor b \lor \bar{c})
(\bar{a} \vee \bar{b} \vee c)
 (a \lor c)
                                                              (\bar{a} \lor \bar{b} \lor c)
                                                                                                       (\bar{b} \vee \bar{c} \vee d)
                                           a = 1
         Learned clause
                                           b = 1
```

What's the big deal?

- Significantly prune the search space because learned clause is useful forever!
- Useful in generating future conflict clauses.



Search Completeness

- With conflict driven learning, SAT search is still guaranteed to be complete.
- SAT search becomes a decision stack instead of a binary decision tree.
- When encountering a conflict, the conflict analysis does the following tasks:
 - Learned clause
 - Indicate where to backtrack
 - Learned implication

SAT Becomes Practical

- Conflict driven learning greatly increases the capacity of SAT solvers (several thousand variables) for structured problems.
- Realistic applications became plausible.
 - Usually thousands and even millions of variables
 - Typical EDA applications can make use of SAT including circuit verification, FPGA routing and many other applications
- Research direction changes towards more efficient implementations.

Outline

- Fundamental Concepts
- Core algorithms of satisfiability problems
 - Davis-Putnam Algorithm
 - DPLL Algorithm
 - GRASP Algorithm
 - zChaff Algorithm
- 3 Heuristics
- SAT competitions
- 6 Applications



zChaff

- M. Moskewicz, C. Madigan, Y. Zhao, L. Zhang, S. Malik," Chaff: Engineering an Efficient SAT Solver" Proc. DAC 2001. (UC Berkeley, MIT and Princeton Univ.)
- Make the core operations fast.
 - After profiling, the most time-consuming parts are Boolean Constraint Propagation (BCP) and Decision.
- As always, pruning search space (i.e. conflict driven learning) is important.

- When can BCP occur?
 - All literals but one are assigned to Falsein a clause.

```
The implied cases of (v1 \lor v2 \lor v3): (0 \lor 0 \lor v3) or (0 \lor v2 \lor 0) or (v1 \lor 0 \lor 0)
```

- For an N-literal clause, this can only occur after N 1 literals have been assigned to False.
- * So, (theoretically) we could completely ignore the first N-2 assignments to this clause.
- Two watched Literals: In reality, we pick two literals in each clause to "watch" and thus can ignore any assignments to the other literals in the clause.

- Heuristically start with watching two unassigned literals in each clause.
- When one of the two watched literals is assigned True, this clause becomes True.
- When one of the two watched literals is assigned False, we send the clause into an Update-Watch queue to do one of the followings:
 - 1. Updating (there exists another unassigned literal)
 - 2. BCP (only one watched literal unassigned)
 - 3. Conflict handling (all literals are False)

- Initially, pick any two literals in each clause as the watched literals.
 - Green: watched literals
- Clauses with only one literal are detected at the mean time.

$$\begin{array}{c} v2 \lor v3 \lor v1 \lor v4 \lor v5 \\ v1 \lor v2 \lor \overline{v3} \\ \hline v1 \lor \overline{v2} \\ \hline \overline{v1} \lor v4 \\ \hline v1 \longleftarrow \end{array} \text{ Detect unit clause}$$

- We begin by processing the assignment v1 = F
 - * Implied by the unit clause $\overline{v1}$

$$v2 \lor v3 \lor \frac{v1}{v1} \lor v4 \lor v5$$

$$v1 \lor v2 \lor \overline{v3}$$

$$\frac{v1}{v1} \lor \overline{v2}$$

$$\overline{v1} \lor v4$$

State: v1 = F

Pending:

- Need not process clauses where watched literals are set to True.
 - Because those clauses are now satisfied.

$$v2 \lor v3 \lor v1 \lor v4 \lor v5$$

$$v1 \lor v2 \lor \overline{v3}$$

$$v1 \lor \overline{v2}$$

$$\Rightarrow \overline{v1} \lor v4$$

State: v1 = F

Pending:

- Need not process clauses where neither watched literal is assigned.
 - Because those clause are definitely not a unit clause.

$$\Rightarrow v2 \lor v3 \lor v1 \lor v4 \lor v5$$

$$v1 \lor v2 \lor \overline{v3}$$

$$v1 \lor \overline{v2}$$

$$\overline{v1} \lor \overline{v4}$$

State: v1 = F

Pending:

Only examine clauses where a watched literal is set to False due to the assignment.

$$v2 \lor v3 \lor v1 \lor v4 \lor v5$$

$$\Rightarrow v1 \lor v2 \lor \overline{v3}$$

$$\Rightarrow v1 \lor \overline{v2}$$

$$\overline{v1} \lor \overline{v2}$$

$$\overline{v1} \lor v4$$

$$State : v1 = F$$

$$Pending :$$

For the second clause, we replace v1 with $\overline{v3}$ as a new watched literal because $\overline{v3}$ is not assigned to False.

State: v1 = F State: v1 = F

Pending: Pending:

- The third clause is a unit clause.
- We record the new implication of $\overline{v2}$, and add it to the queue of assignments to process.

$$v2 \lor v3 \lor v1 \lor v4 \lor v5$$

$$v1 \lor v2 \lor v3 \lor v1 \lor v4 \lor v5$$

$$v1 \lor v2 \lor v3 \lor v1 \lor v4 \lor v5$$

$$v1 \lor v2 \lor v3$$

State:
$$v1 = F$$
 State: $v1 = F$

Pending:
$$\Longrightarrow$$
 Pending: $(v2 = F)$

- \bullet Next, for $\overline{v2}$, only the first two clauses are examined.
 - * For the first clause, replace v2 with v4 as a new watched literal.

State :
$$v1 = F$$
, $v2 = F$ State : $v1 = F$, $v2 = F$

Pending:
$$\Longrightarrow$$
 Pending: $(v3 = F)$

- \bullet Next, for $\overline{v3}$, only the first clause is examined.
 - For the first clause, replace v3 with v5 as a new watched literal.
 - Since there are no pending assignments, and no conflict, BCP terminates and we make a decision. Both v4 and v5 are unassigned. Let's say we assign v4 = True and proceed.

$$\Rightarrow \begin{array}{ccc} v2 \lor v3 \lor v1 \lor v4 \lor v5 & \Longrightarrow & v2 \lor v3 \lor v1 \lor v4 \lor v5 \\ v1 \lor v2 \lor \overline{v3} & v1 \lor v2 \lor \overline{v3} \\ \hline v1 \lor \overline{v2} & v1 \lor \overline{v2} \\ \hline v1 \lor v4 & \overline{v1} \lor v4 \\ \hline \end{array}$$

State :
$$v1 = F$$
, $v2 = F$,
 $v3 = F$

State : $v1 = F$, $v2 = F$,
 $v3 = F$

Pending :

Pending :

BCP Algorithm

- Next, for v4, all clauses are satisfied.
- lacktriangle Depend on implementation, it may continue and assign value to v5.
- The instance is SAT, and we are done.

$$v2 \lor v3 \lor v1 \lor v4 \lor v5$$

$$v1 \lor v2 \lor \overline{v3}$$

$$v1 \lor \overline{v2}$$

$$\overline{v1} \lor v4$$

State:
$$v1 = F, v2 = F,$$

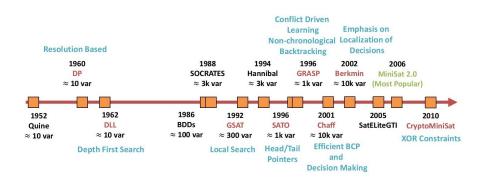
 $v3 = F, v4 = T$

Pending:

BCP Algorithm Summary

- During forward progress: Decisions and Implications
 - Only need to examine clauses where watched literal is set to F
 - Can ignore any assignments of literals to T
 - Can ignore any assignments of non-watched literals
- 😚 During backtrack: Unwind Assignment Stack
 - No action is required at all to unassigned variables
 - But it is computation-intensive part in SATO (SATO: an Efficient Propositional Prover. Hantao Zhang*. Department of Computer Science. The University of Iowa. Iowa City, IA 52242-1419, USA)
- Overall minimize clause access

The Timeline of the SAT Solver



Outline

- Fundamental Concepts
- 2 Core algorithms of satisfiability problems
- 3 Heuristics
 - Decision heuristics
 - Restart mechanism
- SAT competitions
- 6 Applications

Outline

- Fundamental Concepts
- Core algorithms of satisfiability problems
- 3 Heuristics
 - Decision heuristics
 - Restart mechanism
- 4 SAT competitions
- 6 Applications

Make Decision

- Because we want to prove that the target Boolean formula is satisfiable or not, we should start with guessing the state (True or False) of a variable until the proof is done.
- Some strategy:
 - Random
 - Dynamic Largest Individual Sum (DLIS)
 - Variable State Independent Decaying Sum (VSIDS)

RAND and DLIS

- Random
 - Simply select an unassigned variable and a value randomly for the next decision.
- Oynamic Largest Individual Sum (DLIS)
 - * At each decision simply choose the assignment that satisfies the most unsatisfied clauses.
 - Simple and intuitive.
 - However, considerable work is required to maintain the statistics.
 - The total effort required is much more than the effort for the BCP algorithm in zChaff.

VSIDS

- Variable State Independent Decaying Sum (VSIDS)
 - Each variable in each polarity has a counter which is initialized to zero.
 - When a new clause is added to the database, the counter associated with each literal in this clause is incremented.
 - The (unassigned) variable and polarity with the highest counter is chosen at each decision.
 - Ties are broken randomly by default configuration.
 - Periodically, all the counters are divided by a constant.

VSIDS (cont.)

- VSIDS attempts to satisfy the conflict clauses but particularly attempts to satisfy recent learned clauses.
- Difficult problems generate many conflicts (and therefore many conflict clauses), the conflict clauses dominate the problem in terms of literal count.
- Since it is independent of the variable state, it has very low overhead.
- The average rum time overhead in zChaff:
 - BCP: about 80%
 - Decision: about 10%
 - Conflict analysis: about 10%

BerkMin

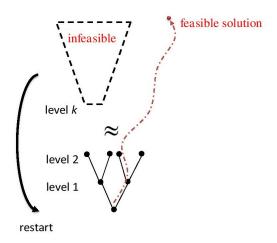
- E. Goldberg, and Y. Novikov, "BerkMin: A Fast and Robust Sat-Solver", Proc. DATE 2002. (Cadence Berkeley Labs and Academy of Sciences in Belarus)
- BerkMin tries to satisfy the most recent clause.
- The clause database is organized as a stack.
- The clauses of the original Boolean formula are located at the bottom of the stack and each new conflict clause is added to the top of the stack.
- The current top clause is the an unsatisfied clause which is the closest to the top of the stack.
- When making decision, choose the most active unassigned variable in the current top clause by using VSIDS.

Outline

- Fundamental Concepts
- Core algorithms of satisfiability problems
- 3 Heuristics
 - Decision heuristics
 - Restart mechanism
- 4 SAT competitions
- 6 Applications

Restart Motivation

Best time to restart: when algorithm spends too much time under a wrong branch



Restart

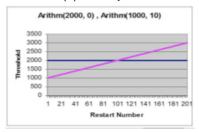
- Motivation: avoid spending too much time in "bad" branches.
 - 🌻 no easy-to-find satisfying assignment
 - no opportunity for fast learning of strong clauses.
- All modern SAT solvers use a restart policy.
 - Following various criteria, the solver is forced to backtrack to level 0.
 - * Abandon the current search tree and reconstruct a new one.
 - The clauses learned prior to the restart are still there after the restart and can help pruning the search space.
- Restarts have crucial impact on performance.
 - Reduce variance increase robustness in the solver.

The Basic Measure for Restarts

- All existing techniques use the number of conflicts learned as of the previous restart.
- The difference is only in the method of calculating the threshold.

Restarts strategies

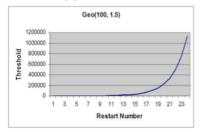
- Arithmetic (or fixed) series.
 - Parameters: x, y
 - t: threshold, when conflict number reaches the threshold, restart!
 - \bullet Init(t) = x
 - \bullet Next(t) = t + y



- Used in (solver name(x, y)):
 - Berkmin (550, 0)
 - Eureka (2000, 0)
 - zChaff 2004 (700, 0)
 - Siege (16000, 0)

Restart Strategies

- Geometric series.
 - Parameters: x, y
 - * t: threshold, when conflict number reaches the threshold, restart!
 - \bullet Init(t) = x
 - Next(t) = t * y



- Used in (solver name(x, y)):
 - Minisat 2007 (100, 1.5)

Restart Strategies

- Inner-Outer Geometric series.
 - * Parameters: x, y, z
 - * t: threshold, when conflict number reaches the threshold, restart!
 - \bullet Init(t) = x

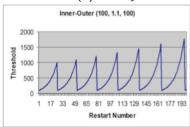
$$\text{if } (t * y < z)$$

$$Next(t) = t * y$$

else

$$Next(t) = x$$

 $Next(z) = z * y$



- Used in (solver name(x, y, z)):
 - Picosat (100, 1.1, 1000)

Other Issues

- Incremental SAT
 - Take apart the clause database.
 - Solve the first part and record the learned information.
 - If it is UNSAT, then stop.
 - If it is SAT, then add the next part to solve.
 - And so on...
- Refutation proof, i.e., proof of UNSAT (Ex.Resolution Proof)
- Parallel computation
- Memory management
- etc...

Outline

- Fundamental Concepts
- Core algorithms of satisfiability problems
- Heuristics
- SAT competitions
- 6 Applications

SAT competitions

- From March to June
- The international SAT Competitions (Starting from 2002) http://www.satcompetition.org/
 - Three main categories of benchmarks: Application(Industrial), Hard Combinatorial(Crafted), Random
 - Three Evaluation in each category: SAT, UNSAT, ALL(SAT + UNSAT)
 - Separate sequential and parallel since 2011
- SAT-Race (2015, 2010, 2008, 2006) http://baldur.iti.kit.edu/sat-race-2015/
- SAT Challenge 2012 http://baldur.iti.kit.edu/SAT-Challenge-2012/

Famous SAT Solvers

- MiniSat, http://minisat.se/
 - Silver in 2005, Gold in 2006 and 2008
 - Well-known for its compact and simple implementation
 - Originally only 600 lines in total but contains most algorithms mentioned in the slide!!
 - A category since 2009 called Minisat Hack
- 📀 SATzilla, http://www.cs.ubc.ca/labs/beta/Projects/SATzilla/
 - Gold in 2007, 2009, and 2012
 - Evaluate the problem instance first
 - Select an appropriate solver to solve

Famous SAT Solvers

- ppfolio, http://www.cril.univ-artois.fr/~roussel/ppfolio/
 - Win a total of 16 medals in 2011
 - Assign cores to the five solvers in use.
 - Winners of recent years
 - glucose, http://www.labri.fr/perso/lsimon/glucose/
 - Lingeling, http://fmv.jku.at/lingeling

Outline

- Fundamental Concepts
- Core algorithms of satisfiability problems
- Heuristics
- 4 SAT competitions
- 6 Applications

The Usage of the MiniSat

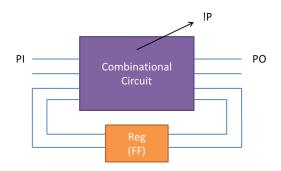
- MiniSat Page: http://minisat.se/
- The newest version: 2.2.0
- Use MiniSat to find a solution of $F = (x_0 \lor x_1 \lor x_2) \land (\overline{x_1} \lor x_2)$.
 - Go to MiniSat Page to download it.
 - Tar the .gz file tar -zxvf minisat-2.2.0.tar.gz
 - Change to directory "core" cd core
 - Modify path export MROOT=../
 - Make and compile in directory "core" make
 - Build DIMACS CNF file for problem you want to solve http://www.satcompetition.org/2009/format-benchmarks2009.html
 - Run the minisat to solve problem ./minisat CnfFileName

DIMACS CNF Format

- It is a standard format for the input files (CNF files) of SAT solvers.
 - Use c to write comments
 - Start with p cnf VarialbeNumber ClauseNumber
 - Write the clause with integer(with/without "-") for representing the literals
 - Use "0" to mark the end of a clause
- Example: $(x_0 \lor x_1 \lor x_2) \land (\overline{x_1} \lor x_2)$ c this is a simple DIMACS cnf, use 1, 2, 3 for x0, x1, x2 respectively p cnf 3 2 1 2 3 0 -2 3 0

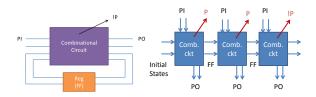
Example 1: Bounded Model Checking

• We want to check property AG(p) for a given sequential circuit. See whether it has bugs!



Timeframe Expansion Model

Iterative timeframe expansion model: sequential SAT becomes a combinational problem.

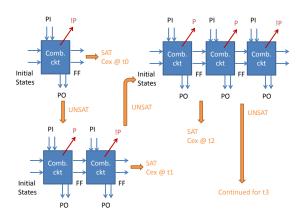


BMC Algorithm

- Let C be the set of constraints on the combinational circuit
- For an iterative model that unfolds the circuit for n times, let C_i correspond to the i-th iteration of the circuit constraint $(0 \le i \le k-1)$
- Let I_0 be the initial state value
- Let P be the property to prove
- Following is the BMC algorithm:
- BMC(P)
 - Let k=1
 - loop:
 - \bullet if $(SAT(I_0 \wedge C_0 \wedge ... \wedge C_{k-1} \wedge !P_{k-1}))$
 - return Find a counter-example at time (k-1)
 - k=k+1
 - go to loop

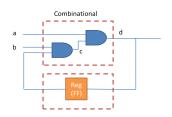
BMC Algorithm

• In other words ...



How to Write CNF for C_i

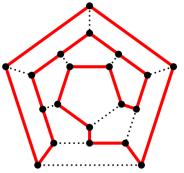
Here is an example:



- We use a_i, b_i, c_i, d_i to represent the signals for timeframe i
- ightharpoonup We use s_{out} to represent FF_{out} for timeframe i
- $C_i = (c_i = b_i \land s_{outi}) \land (d_i = a_i \land c_i) \land (s_{outi} = d_{i-1}) \text{ for } i > 0...(1)$
- $C_0 = (c_0 = b_0 \wedge I_0) \wedge (d_0 = a_0 \wedge c_0)...(2)$
- lacktriangle We can use (3) to rewrite (1) and (2) for CNF

Example 2: Hamiltonian Cycle

Hamiltonian cycle, also called a Hamiltonian circuit, is a graph cycle (i.e., closed loop) through a graph that visits each node exactly once.



(Wiki: http://en.wikipedia.org/wiki/File:Hamiltonian_path.svg)

Encoding

- Encode the Hamiltonian cycle problem into SAT problem by the following way:
 - * Assume that there is a path of length n which is the number of nodes.
 - Let each Boolean variables $x_{i,j}$ represent that i_{th} node is in the j_{th} position of this path.
 - * So there are n^2 Boolean variables in SAT problem by this encoding method.

Add Constraint Clauses

- First constraints: Each node occupies only one position of this path.
- Second constraints: Each position of this path contains only one node.
- Third constraints: Two consecutive nodes are connected by an edge.

First Constraints

- Each node occupies only one position of this path
 - Each node is in the path:

$$(x_{i,0} \lor x_{i,1} \lor \cdots \lor x_{i,n-1})$$
, where $0 \le i \le n-1$

Each node holds only one position (one hot):

$$\begin{split} &\left(\overline{x_{i,0}} \vee \overline{x_{i,1}}\right) \wedge \left(\overline{x_{i,0}} \vee \overline{x_{i,2}}\right) \wedge \dots \\ &\left(\overline{x_{i,0}} \vee \overline{x_{i,n-1}}\right) \wedge \left(\overline{x_{i,1}} \vee \overline{x_{i,2}}\right) \wedge \dots \\ &\left(\overline{x_{i,j}} \vee \overline{x_{i,k}}\right) \wedge \dots \\ &\text{where } 0 \leq i \leq n-1, \ 0 \leq j \leq n-2, \ j+1 \leq k \leq n-1 \end{split}$$

Second Constraints

- Each position of this path contains only one node
 - Each position contains at least a node:

$$(x_{0,i} \lor x_{1,i} \lor \cdots \lor x_{n-1,i}), \text{ where } 0 \le i \le n-1$$

Each position contains only one node (one hot):

$$\begin{split} & \big(\overline{x_{0,i}} \vee \overline{x_{1,i}}\big) \wedge \big(\overline{x_{0,i}} \vee \overline{x_{2,i}}\big) \wedge \dots \\ & \big(\overline{x_{0,i}} \vee \overline{x_{n-1,i}}\big) \wedge \big(\overline{x_{1,i}} \vee \overline{x_{2,i}}\big) \wedge \dots \\ & \big(\overline{x_{j,i}} \vee \overline{x_{k,i}}\big) \wedge \dots \\ & \text{where } 0 \leq i \leq n-1, \ 0 \leq j \leq n-2, \ j+1 \leq k \leq n-1 \end{split}$$

Third Constraints

- Two consecutive nodes are connected by an edge
 - * There is an edge between the i_{th} node and the j_{th} node:

Don't add constraint clauses into solver.

* There is no connection between the i_{th} node and the j_{th} node:

$$\begin{split} &\left(\overline{x_{i,0}} \vee \overline{x_{j,1}}\right) \wedge \left(\overline{x_{i,1}} \vee \overline{x_{j,2}}\right) \wedge \dots \\ &\left(\overline{x_{i,n-2}} \vee \overline{x_{j,n-1}}\right) \vee \left(\overline{x_{i,n-1}} \vee \overline{x_{j,0}}\right) \\ &\textit{where } 0 \leq i \leq n-1, \ 0 \leq j \leq n-1, \textit{and } i \neq j \end{split}$$

Demo

Given following graph, check if there is a Hamiltonian Cycle

