

## 03—Propositional Logic III

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CS 5209: Foundation in Logic and AI

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- 1 Soundness and Completeness
- 2 Conjunctive Normal Form
- 3 SAT Solvers

1 Soundness and Completeness

2 Conjunctive Normal Form

3 SAT Solvers

## Conjunctive Normal Form

### Definition

A literal  $L$  is either an atom  $p$  or the negation of an atom  $\neg p$ . A formula  $C$  is in *conjunctive normal form* (CNF) if it is a conjunction of clauses, where each clause is a disjunction of literals:

$$L ::= p \mid \neg p$$

$$D ::= L \mid L \vee D$$

$$C ::= D \mid D \wedge C$$

## Examples

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$(\neg p \vee q \vee r) \wedge (\neg q \vee r) \wedge (\neg r)$  is in CNF.

$(\neg p \vee q \vee r) \wedge ((p \wedge \neg q) \vee r) \wedge (\neg r)$  is not in CNF.

$(\neg p \vee q \vee r) \wedge \neg(\neg q \vee r) \wedge (\neg r)$  is not in CNF.

## Usefulness of CNF

### Lemma

A disjunction of literals  $L_1 \vee L_2 \vee \dots \vee L_m$  is valid iff there are  $1 \leq i, j \leq m$  such that  $L_i$  is  $\neg L_j$ .

How to disprove

$$\models (\neg q \vee p \vee r) \wedge (\neg p \vee r) \wedge q$$

Disprove any of:

$$\models (\neg q \vee p \vee r) \quad \models (\neg p \vee r) \quad \models q$$

## Usefulness of CNF

### Lemma

A disjunction of literals  $L_1 \vee L_2 \vee \dots \vee L_m$  is valid iff there are  $1 \leq i, j \leq m$  such that  $L_i$  is  $\neg L_j$ .

How to prove

$$\models (\neg q \vee p \vee q) \wedge (p \vee r \neg p) \wedge (r \vee \neg r)$$

Prove all of:

$$\models (\neg q \vee p \vee q) \quad \models (p \vee r \neg p) \quad \models (r \vee \neg r)$$

## Usefulness of CNF

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### Proposition

Let  $\phi$  be a formula of propositional logic. Then  $\phi$  is satisfiable iff  $\neg\phi$  is not valid.

### Satisfiability test

We can test satisfiability of  $\phi$  by transforming  $\neg\phi$  into CNF, and show that some clause is not valid.



## Transformation to CNF

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### Theorem

Every formula in the propositional calculus can be transformed into an equivalent formula in CNF.

## Algorithm for CNF Transformation

- 1 Eliminate implication using:

$$A \rightarrow B \equiv \neg A \vee B$$

- 2 Push all negations inward using De Morgan's laws:

$$\neg(A \wedge B) \equiv (\neg A \vee \neg B)$$

$$\neg(A \vee B) \equiv (\neg A \wedge \neg B)$$

- 3 Eliminate double negations using the equivalence  $\neg\neg A \equiv A$

- 4 The formula now consists of disjunctions and conjunctions of literals. Use the distributive laws

$$A \vee (B \wedge C) \equiv (A \vee B) \wedge (A \vee C)$$

$$(A \wedge B) \vee C \equiv (A \vee C) \wedge (B \vee C)$$

to eliminate conjunctions within disjunctions.

## Example

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$$\begin{aligned}(\neg p \rightarrow \neg q) \rightarrow (p \rightarrow q) &\equiv \neg(\neg\neg p \vee \neg q) \vee (\neg p \vee q) \\ &\equiv (\neg\neg\neg p \wedge q) \vee (\neg p \vee q) \\ &\equiv (\neg p \wedge q) \vee (\neg p \vee q) \\ &\equiv (\neg p \vee \neg p \vee q) \wedge (q \vee \neg p \vee q)\end{aligned}$$

- 1 Soundness and Completeness
- 2 Conjunctive Normal Form**
- 3 SAT Solvers

## WalkSAT: An Incomplete Solver

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Idea: Start with a random truth assignment, and then iteratively improve the assignment until model is found

Details: In each step, choose an unsatisfied clause (clause selection), and “flip” one of its variables (variable selection).

## WalkSAT: Details

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Termination criterion: No unsatisfied clauses are left.

Clause selection: Choose a random unsatisfied clause.

Variable selection:

- If there are variables that when flipped make no currently satisfied clause unsatisfied, flip one which makes the most unsatisfied clauses satisfied.
- Otherwise, make a choice with a certain probability between:
  - picking a random variable, and
  - picking a variable that when flipped minimizes the number of unsatisfied clauses.

## DPLL: Idea

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- Simplify formula based on pure literal elimination and unit propagation
- If not done, pick an atom  $p$  and split:  $\phi \wedge p$  or  $\phi \wedge \neg p$

## A Linear Solver: Idea

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- Transform formula to tree of conjunctions and negations.
- Transform tree into graph.
- Mark the top of the tree as  $\top$ .
- Propagate constraints using obvious rules.
- If all leaves are marked, check that corresponding assignment makes the formula true.



# Transformation

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$$T(p) = p$$

$$T(\phi_1 \wedge \phi_2) = T(\phi_1) \wedge T(\phi_2)$$

$$T(\neg\phi) = \neg\phi(T)$$

$$T(\phi_1 \rightarrow \phi_2) = \neg(T(\phi_1) \wedge \neg T(\phi_2))$$

$$T(\phi_1 \vee \phi_2) = \neg(\neg T(\phi_1) \wedge \neg T(\phi_2))$$

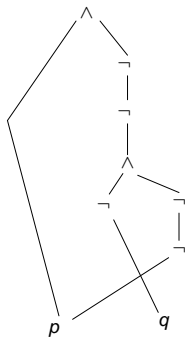
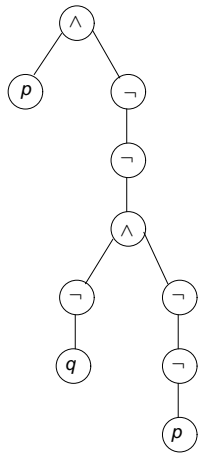
## Example

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$$\phi = p \wedge \neg(q \vee \neg p)$$

$$T(\phi) = p \wedge \neg\neg(\neg q \wedge \neg\neg p)$$

## Binary Decision Tree: Example



# Problem

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What happens to formulas of the kind  $\neg(\phi_1 \wedge \phi_2)$ ?

## A Cubic Solver: Idea

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Improve the linear solver as follows:

- Run linear solver
- For every node  $n$  that is still unmarked:
  - Mark  $n$  with  $\top$  and run linear solver, possibly resulting in temporary marks.
  - Mark  $n$  with  $\perp$  and run linear solver, possibly resulting in temporary marks.
  - Combine temporary marks, resulting in possibly new permanent marks

## The ACC 1997/98 Problem

- “ACC” stands for “Atlantic Coast Conference”, an American college basketball organization
- 9 teams participate in tournament
- dense double round robin: there are  $2 * 9$  dates
- at each date, each team plays either home, away or has a “bye”
- Each team must play each other team once at home and once away.
- there should be at least 7 dates distance between first leg and return match.
- To achieve this, we assume a fixed mirroring between dates: (1,8), (2,9), (3,12), (4,13), (5,14), (6,15) (7,16), (10,17), (11,18)

## The ACC 1997/98 Problem (contd)

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- No team can play away on both last dates
- No team may have more than two away matches in a row.
- No team may have more than two home matches in a row.
- No team may have more than three away matches or byes in a row.
- No team may have more than four home matches or byes in a row.

## The ACC 1997/98 Problem (contd)

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- Of the weekends, each team plays four at home, four away, and one bye.
- Each team must have home matches or byes at least on two of the first five weekends.
- Every team except FSU has a traditional rival. The rival pairs are Clem-GT, Duke-UNC, UMD-UVA and NCSt-Wake. In the last date, every team except FSU plays against its rival, unless it plays against FSU or has a bye.



## The ACC 1997/98 Problem (contd)

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- The following pairings must occur at least once in dates 11 to 18: Duke-GT, Duke-Wake, GT-UNC, UNC-Wake.
- No team plays in two consecutive dates away against Duke and UNC. No team plays in three consecutive dates against Duke UNC and Wake.
- UNC plays Duke in last date and date 11.
- UNC plays Clem in the second date.
- Duke has bye in the first date 16.

## The ACC 1997/98 Problem (contd)

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- Wake does not play home in date 17.
- Wake has a bye in the first date.
- Clem, Duke, UMD and Wake do not play away in the last date.
- Clem, FSU, GT and Wake do not play away in the first date.
- Neither FSU nor NCSt have a bye in the last date.
- UNC does not have a bye in the first date.

## Background

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- Trick and Nemhauser work on the problem from 1995 onwards
- Trick and Nemhauser publish the problem and their approach in “Scheduling a Major Basketball Conference”, Operations Research, 46(1), 1998
- From then onwards, Henz, Walser and Zhang use different techniques to solve the problem

## General Approach

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- Three phases:
  - 1 Generate all possible patterns such as “A H B A H H A H A A H B H A A H H A”
  - 2 Generate all feasible 9-element pattern sets that can be used to construct a schedule
  - 3 Generate schedules from pattern sets
- Output: all feasible solutions, from which the organizers can choose the most suitable one

## Solution Techniques

- Nemhauser and Trick use integer programming for all three steps, leading to a “turn-around time” of 24 hours
- Henz uses constraint programming, turn-around time of less than 1 minute, publishes his approach in “Scheduling a Major Basketball Conference—Revisited”, Operations Research, 49(1), 2001
- Zhang Hantao uses SAT solving, turn-around time of 2 seconds, see “Generating College Conference Basketball Schedules using a SAT Solver”
- Different approach: In 1998, J.P. Walser described a local-search based method for finding some (not all) solutions, without using 3 phases

## How to Encode ACC as a SAT Formula

- Consider Phase 3: Generation of schedule, assigning teams to opponents at every day of the tournament
- For teams  $x, y$ , day  $z$ , introduce atom  $p_{x,y,z} = T$  iff team  $x$  plays a home game against team  $y$  in day  $z$ .
- Example of encoding constraints: “Each team must play each other team once at home and once away.”
- For every pair of distinct teams  $s$  and  $t$ , we have:

$$\begin{aligned}
 & (p_{s,t,1} \wedge \neg p_{s,t,2} \wedge \cdots \wedge \neg p_{s,t,18}) \vee \\
 & (\neg p_{s,t,1} \wedge p_{s,t,2} \wedge \neg p_{s,t,3} \wedge \cdots \wedge \neg p_{s,t,18}) \vee \\
 & \quad \vdots \\
 & (\neg p_{s,t,1} \cdots \wedge \neg p_{s,t,17} \wedge p_{s,t,18})
 \end{aligned}$$

- Convert formula into CNF and use a complete SAT solver

## Some Statistics

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- Zhang Hantao used the DPLL-based SAT solver SATO
- Phase 1:  $18 \cdot 3 = 54$  propositional atoms, 1499 clauses, taking 0.01 seconds, resulting in 38 patterns
- Phase 2:  $38 \cdot 9 \cdot 3 = 1026$  propositional atoms, 569300 clauses, taking 0.60 seconds, resulting in 17 pattern sets
- Phase 3:  $9 \cdot 9 + 9 \cdot 8 \cdot 18 = 1377$  propositional atoms, hundreds of thousands of clauses, taking less than 2 seconds, resulting in 179 solutions

## Conclusion

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- For many discrete constraint satisfaction problems such as the ACC 1997/98 problem, an encoding in SAT and use of a state-of-the-art SAT solver provides an attractive solving technique.
- The approach takes advantage of the effort that the designers of SAT solvers such as SATO spent in order to optimize the solver.
- This works well, because the solver is independent of the application domain; it can be used without modification across application domains.