

Pete Manolios Northeastern

Computer-Aided Reasoning, Lecture 9

Normal Forms

- Minimizing DNF has many applications
 - this is used to analyze the reliability of safety-critical systems
- CNF is the input format of modern SAT solvers
 - this is the so-called DIMACS format
 - modern SAT solvers can solve industrial problems with 1M variables
- There are many other "normal" forms for Boolean formulae
 - decision trees: widely used in machine learning
 - BDDs: very powerful representation used in verification, AI, program analysis, …

Set Theory Connections

- Set Theory forms the foundations of mathematics
- Set Theory provides foundations of ACL2s:
 - ▶ the universe, *U*, is a set
 - recognizers, predicates, etc. in ACL2s are defined in terms of sets
 - atoms in a propositional skeleton are predicates (subsets of U)
- It turns out that there are interesting connections between propositional logic and set theory
- Here is an example

 - ▶ $P \cap (P \cup Q) = P$ is valid (in set theory)
 - there are obvious similarities in the two formulas above
- Let's explore the connections a little bit

Set Theory Connections

Boolean Algebra of (non-empty) X: a non-empty subset of the 2^x closed under union, intersection and complementation (with respect to X)

- Let U be the ACL2s universe
- ▶ Then $B = \{\emptyset, U\}$ is the smallest Boolean algebra of U
- ▶ The largest Boolean algebra of U is 2^U
- ▶ *B* is isomorphic to propositional logic: \emptyset for F and *U* for T

▶ \lor , \land , \neg correspond, respectively, to the (set theoretic) \cup , \cap , \neg

- ▶ In a Boolean algebra, atoms correspond to unary predicates, *e.g.*, in 2^{*U*}:
 - for clarity's sake, we use upper case vars to indicate atoms in 2^U
 - ▶ let P be {x ∈ U : (integerp x)}
 - ▶ let Q be {x ∈ U : (neg-rationalp x)}
 - ▶ so $P \land Q$ (in 2^{*U*}) means $P \cap Q = \{x \in U : (negp x)\}$

A Boolean algebra formula is valid if = U, e.g.: $P \lor \neg P$ (in 2^U) means $P \cup \neg P = U$

▶ In general, a formula in 2^U corresponds to the subset of U for which it holds

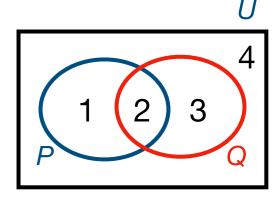
Set Theory Connections

- Boolean Algebra of (non-empty) X: a non-empty subset of the 2^x closed under union, intersection and complementation (with respect to X)
- ${}^{\blacktriangleright}$ V , \wedge , \neg correspond, respectively, to the (set theoretic) $~\cup$, \cap , \neg
- In general, a formula in 2^U corresponds to the subset of U for which it holds
- ^b Can extend Boolean algebra with \Rightarrow , \equiv , etc, using the propositional equalities:

 $P \Rightarrow Q \text{ is } \neg P \lor Q : S = \{x \in U : (\text{implies } (P x) (Q x))\}$

 $P \equiv Q \text{ is } (P \Rightarrow Q) \land (Q \Rightarrow P) : W = \{x \in U : (\text{iff } (P x) (Q x))\}$

- The equalities of propositional logic & Boolean algebra are the same!
 - Propositional logic validity: $p \lor \neg p$
 - Boolean algebra: $P \lor \neg P$ (in 2^U) is valid since = U
 - Check the rest of the equalities in the notes
- The result is useful when analyzing propositional logic formulas, e.g.:
 - $p \land (p \lor q) \equiv p \quad \text{ is valid iff }$
 - ▶ $P \cap (P \cup Q) = P$ is valid (because $P \equiv Q$ is valid iff P = Q holds)



Which regions are in S? 2,3,4

Regions for $Q \Rightarrow P$? 2,1,4

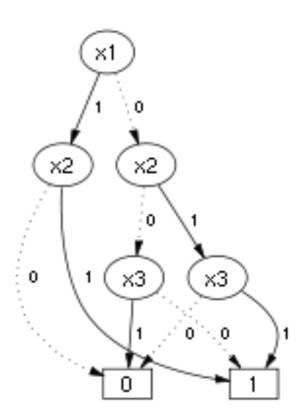
Which regions are in *W*? 2,4 (intersection of above)

BDDs and Decision Trees

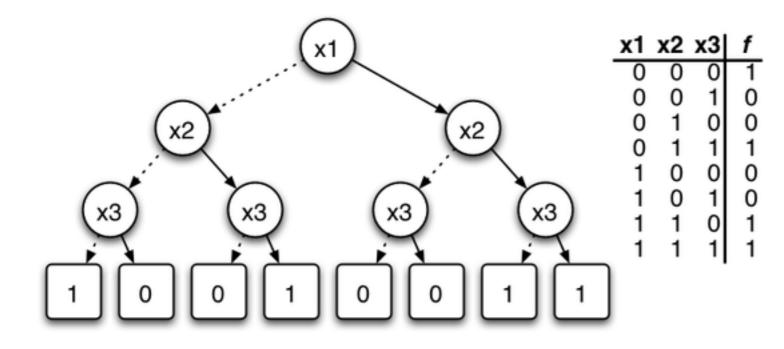
▶ A BDD on x₁, ..., x_n is a DAG G=(V, E) where

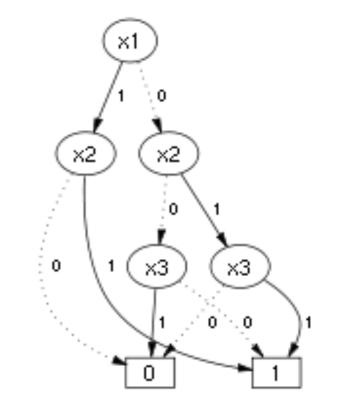
- exactly 1 vertex has indegree 0 (the root)
- all vertices have outdegree 0 (leaves) or 2 (inner nodes)
- ▶ the inner nodes are labeled from $\{x_1, ..., x_n\}$
- the leaves are labeled from {0, 1}
- one of the edges from an inner node is labeled by 0; the other by 1
- The BDD G=(V, E) represents a Boolean function, say f
 - ▶ for any assignment A in Bⁿ, f(A) is computed recursively from root
 - ▶ if we reach a leaf, return the label
 - ▶ for inner nodes, say labeled with x_i, take the edge labeled by A(x_i)
- A decision tree is a BDD whose graph is a tree
- A BDD is an OBDD if there is a permutation on p={1,2, ..., n} s.t. for all edges (u, v) in E, where u, v are labeled by x_i, x_j, we have that p_i < p_j
- An OBDD is an ROBDD if it has no isomorphic subgraphs and all children are distinct

Images from Wikipedia



BDDs and Decision Trees





Decision Tree for f

ROBDD for f

How do we generate DNF from a decision tree? ROBDD?

Images from Wikipedia



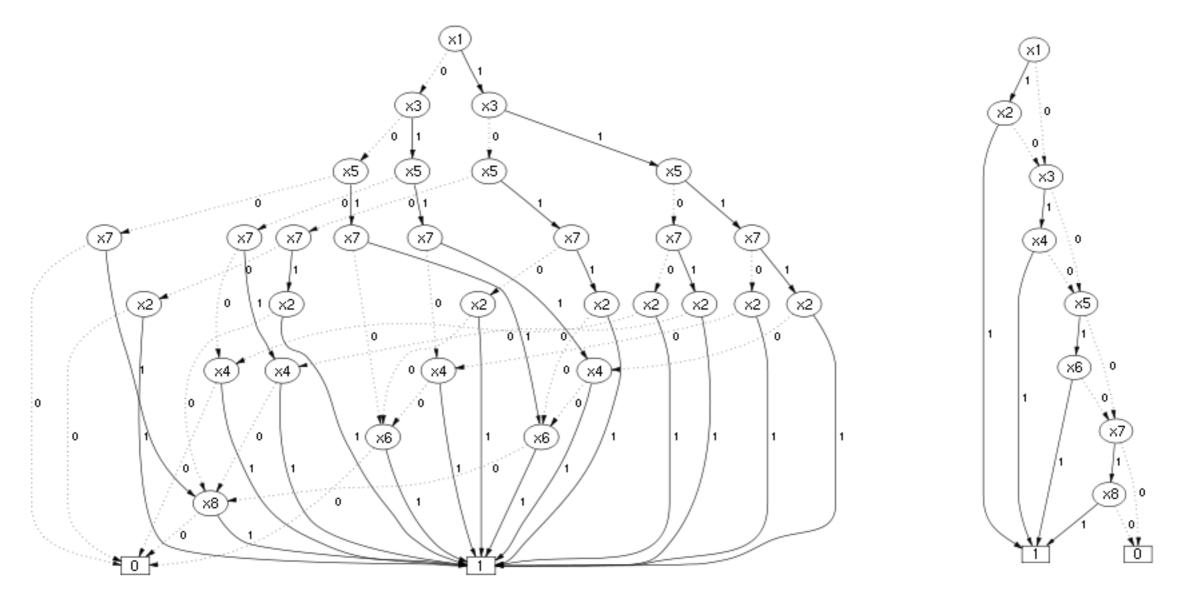
- Decision trees are widely used, e.g., in machine learning (ID3, C4.5, …)
- BDDs are widely used (BDD usually means ROBDD)
 - Popularized by Bryant
 - Very efficient algorithms for constructing, manipulating BDDs
 - Used in verification, synthesis, fault trees, security, AI, model checking, static analysis, …
 - Bryant's paper was the most cited research paper (at some point)
 - Many BDD packages available
- Once a variable ordering is selected, BDDs are canonical!
 - Construct decision tree using Shannon expansion and merge isomorphic nodes, remove nodes who children are equal until you reach a fixpoint
 - To see, this note that BDDs are essentially DFA that recognize strings in {0,1}ⁿ and such automata can be minimized (note nodes with equal children remain)
 - So, checking equality is just pointer equality (with appropriate data structures)
 - Can be used for model checking: represent set of reachable states & transition system with BDDs
 - Bryant, Clarke, Emerson & McMillan got 1998 Paris Kanellakis Award for symbolic model checking

BDD Break

- Made the safety-analysis repo public; see link from slides
- Find someone you haven't spoken to yet
- Come up with an example formula over 4 variables where variable order matters wrt BDD size

Variable Ordering for BDDs

Variable ordering matters: find the best ordering is hard.



Bad Ordering

Good Ordering

What function is this?

Images from Wikipedia

Projects & Presentations

- Talk with me regarding projects
 - Set up 1/2 1 hour slots to go over project ideas
- Some ideas (groups 1-2)
 - Better induction proofs
 - Refinement: verification
 - Distributed system verification: perimeter monitoring example, etc
 - CyC: ontology engineering (Doug Lenat)
 - AI & FM: Reasoning about programs
 - Reproduce interesting result
 - Survey paper on some FM topic
 - Harrison's book in ACL2s

Algorithms for SAT

- Modern SAT solvers accept input in CNF
 - Dimacs format:
 - ▶ 1 -3 4 5 0
 - ≥ -4 7 0
 - ▷...
- Davis & Putnam Procedure (DP)
 - Dates back to the 50's
 - Based on resolution
 - Helps to explain learning

DP SAT Algorithm

- Davis Putnam (1960)
- Input: CNF formula
- Output: SAT/UNSAT
- Idea: apply three rules until
 - Derive the empty clause: UNSAT (identity of \lor is false)
 - No clauses remain: SAT (identity of \land is true)
- Three "rules"
 - Pure literal rule (affirmative-negative rule)
 - Unit resolution rule (unit propagation, BCP, 1-literal rule)
 - Resolution (Called consensus, also used for logic minimization)

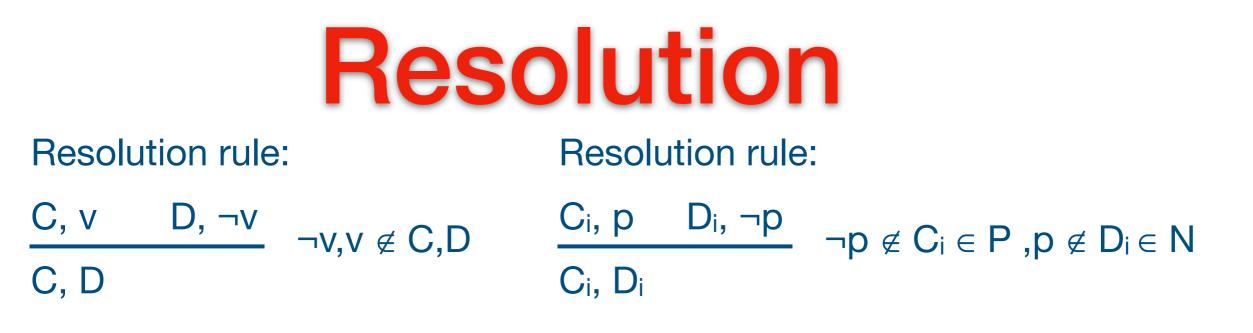
Pure Literal Rule

- Siven F, a set of clauses, and literal ℓ such
 - ▶ ℓ appears in F
 - ▶ ¬ ℓ does not appear in F
 - remove all clauses containing
- Equisatisfiable because we can make l true
- ${}^{\blacktriangleright}$ Notice that this always simplifies F
- Modern SAT solvers tend to not use the rule (efficiency)

Boolean Constraint Propagation

Unit resolution rule:

- ▶ BCP: given a set of clauses including {ℓ}
 - remove all other clauses containing { (subsumption)
 - ▶ remove all occurrences of ¬ℓ in clauses (unit resolution)
 - repeat until a fixpoint is reached



- Soundness of rule: above line implies below line
- If below line is SAT, so is above line (w/ side conditions)
- Given literal p, set of clauses S, let P be the clauses in S that contain p only positively and let N be the clauses that contain p only negatively.
 Let E be the rest of the clauses. Then S is SAT iff S' is SAT, where S'= E
 U the set of all p-resolvents of P and N.
- Proof: If A is an assignment for S, then if A(p)=true, all clauses in N, with ¬p removed are satisfied, so each p-resolvent is satisfied. Similarly if A(p)=false. If A is an assignment for S', then it satisfies all Ci or all Di: suppose it doesn't satisfy Ck, then it must satisfy all Di. If it satisfies all Ci, let A'(p)=false, else A'(p)=true and A'(x)=A(x) otherwise.

Resolution Example

Resolution rule:

$$\frac{C, v \qquad D, \neg v}{C, D} \qquad C, D \text{ are clauses, } \neg v \not\in C \text{ and } v \not\in D$$

Given literal p, set of clauses S, let P be the clauses in S that contain p only positively and let N be the clauses that contain p only negatively. Let E be the rest of the clauses. Then S is SAT iff S' is SAT, where $S' = E \cup$ the set of all p-resolvents of P and N.

$$\{\{\neg p, q, r, s\}, \{p, \neg q, s\}, \{\neg p, \neg q, r, \neg s\}, \{p, \neg r, \neg r\}, \{p, q\}, \{\neg p, \neg q, s\} \}$$
Resolve on q

$$\{\neg p, p, r, s\}, \{\neg p, r, s\}, \{\neg p, r, s\}, \{p, s\} \}$$
Notice that clauses that contain a literal and its negation can be thrown away. Why?

Resolution Example

Resolution rule:

$$\frac{C, v \qquad D, \neg v}{C, D} \qquad C, D \text{ are clauses, } \neg v \notin C \text{ and } v \notin D$$

Given literal p, set of clauses S, let P be the clauses in S that contain p only positively and let N be the clauses that contain p only negatively. Let E be the rest of the clauses. Then S is SAT iff S' is SAT, where S' = E U the set of all p-resolvents of P and N.

$$\{\{\neg p, q, r, s\}, \{p, \neg q, s\}, \{\neg p, \neg q, r, \neg s\}, \{p, \neg r, \neg s\}, \{p, q, \neg r\}, \{p, q\}, \{\neg p, \neg q, s\}\}$$

Resolve on q { $\neg p, p, r, s$ } {{ $p, \neg r, \neg s$ }, { $\neg p, r, s$ }, {p, s}}

Notice that clauses that contain a literal and its negation can be thrown away. Why?

Resolve on r

 $\{\{p,s\}\}$ Sat, resolve on p to get $\{\}$ or use pure literal rule

How do we generate a satisfying assignment? Next homework

DP SAT Algorithm

- Input: CNF formula, Output: SAT/UNSAT
- Base case: empty clause: UNSAT
- Base case: no clauses: SAT
 - Apply these two rules until fixpoint
 - Pure literal rule
 - ▶ BCP
 - Choose var, say x, perform all possible resolutions, remove trivial clauses and clauses containing x
 - Repeat
- Existentially quantify variables, one at a time
- Problem: space blow-up