15-780: Graduate AI Lecture 3. FOL proofs; SAT

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Admin

HW1

- Out today
- Due Tue, Feb. 3 (two weeks)
 - hand in hardcopy at beginning of class
- Covers propositional and FOL
- Don't leave it to the last minute!

Collaboration policy

- OK to discuss general strategies
- What you hand in must be your own work
 - written with no access to notes from joint meetings, websites, etc.
- You must acknowledge all significant discussions, relevant websites, etc., on your HW

Late policy

- You have 3 late days to use on HWs
 - these account for conference travel, holidays, illness, or any other reasons
- After late days, 75% for next day, 50% for next, 0% thereafter (but still must turn in)
- Day = 24 hrs, HWs due at 10:30AM

Office hours

- Office hours start this week (see website for times)
- But, I have a conflict this week due to admissions; let me know by email if there is demand, and if so I can reschedule

Matlab tutorial

• Thu 1/22, 4–5PM, Wean Hall 5409

Review

In propositional logic

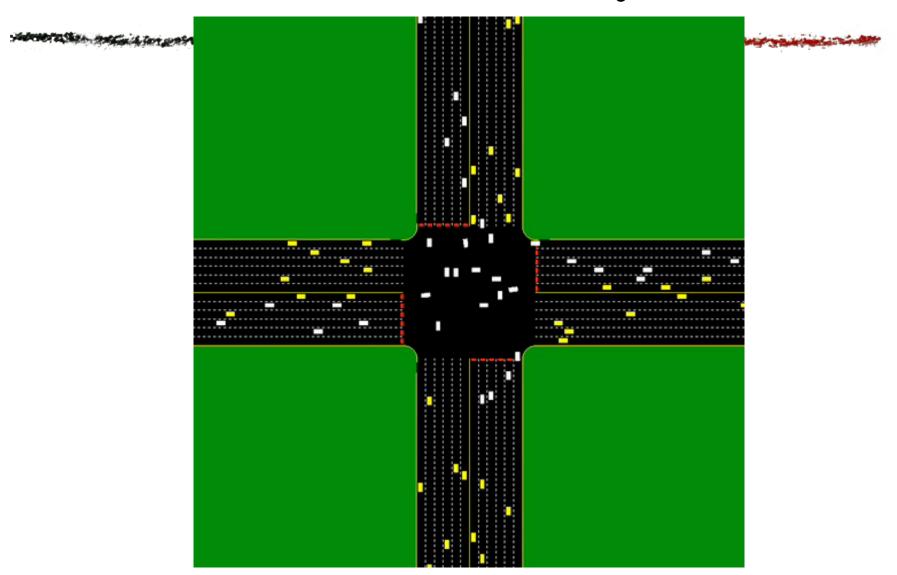
- Compositional semantics, structural induction
- Proof trees, proof by contradiction
- Inference rules (e.g., resolution)
- Soundness, completeness
- Horn clauses
- Nonmonotonic logic

In FOL

- Compositional semantics
 - objects, functions, predicates
 - terms, atoms, literals, sentences
 - quantifiers, free/bound variables
 - models, interpretations
- Generalized de Morgan's law
- Skolemization, CNF

Project Ideas

Traffic insanity



Sensor planning



 Plan a path for this robot so that it gets a good view of an object as fast as possible

Mini-robots





- Do something cool w/ Lego Mindstorms
 - plan footstep placements
 - plan how to grip objects

Poker



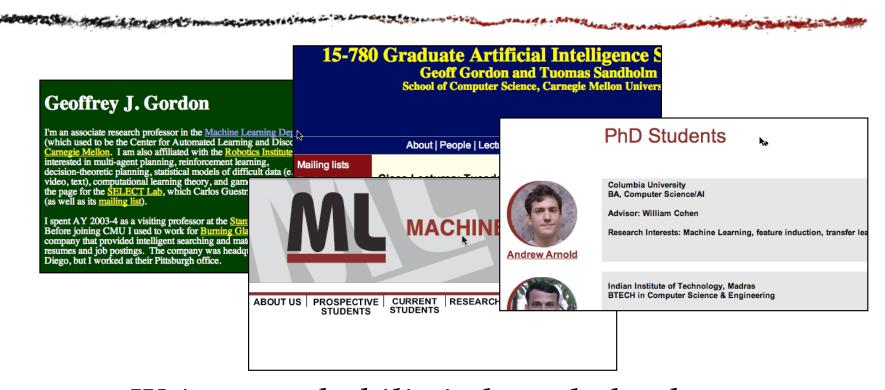
Poker

- Minimax strategy for heads-up poker = solving linear program
- 1-card hands, 13-card deck: 52 vars, <u>instantaneous</u>
- RI Hold'Em: ~1,000,000 vars
 - 2 weeks / 30GB (exact sol, CPLEX)
 - 40 min / 1.5GB (approx sol)
- TX Hold'Em: ??? (up to 10¹⁷ vars or so)

Poker

Learning by repeated play

Understand the web



- Write a probabilistic knowledge base describing a portion of the web
- Learn parameters of the model

Proofs in FOL

FOL is special

- Despite being much more powerful than propositional logic, there is still a sound and complete inference procedure for FOL
- Almost any significant extension breaks this property
- This is why FOL is popular: very powerful language with a sound & complete inference procedure

Proofs

- Proofs by contradiction work as before:
 - \circ add $\neg S$ to KB
 - put in CNF
 - run resolution
 - if we get an empty clause, we've proven
 S by contradiction
- But, CNF and resolution have changed

Generalizing resolution

- *Propositional:* $(\neg a \lor b) \land a \vDash b$
- FOL:

```
(\neg man(x) \lor mortal(x)) \land man(Socrates)
```

- ⊨ (¬man(Socrates) ∨ mortal(Socrates)) ∧ man(Socrates)
- $\models mortal(Socrates)$
- Difference: had to substitute $x \rightarrow Socrates$

Universal instantiation

• What we just did is UI:

```
(\neg man(x) \lor mortal(x))
\vDash (\neg man(Socrates) \lor mortal(Socrates))
```

- Works for x → any ground term
 (¬man(uncle(student(Socrates))) ∨ mortal(uncle(student(Socrates))))
- For proofs, need a good way to find useful instantiations

Substitution lists

- ∘ *List of variable* → *value pairs*
- Values may contain variables (leaving flexibility about final instantiation)
- But, no LHS may be contained in any RHS
 - i.e., applying substitution twice is the same as doing it once
- \circ E.g., $x \to Socrates$, $y \to LCA(Socrates, z)$

 $LCA = last\ common\ advisor$

Unification

- Two FOL terms **unify** with each other if there is a substitution list that makes them syntactically identical
- man(x), man(Socrates) unify using the substitution $x \rightarrow Socrates$
- Importance: purely syntactic criterion for identifying good substitutions

Unification examples

- loves(x, x), loves(John, y) unify using $x \rightarrow John, y \rightarrow John$
- \circ loves(x, x), loves(John, Mary) can't unify
- loves(uncle(x), y), loves(z, aunt(z)):

Unification examples

- loves(x, x), loves(John, y) unify using $x \rightarrow John, y \rightarrow John$
- loves(x, x), loves(John, Mary) can't unify
- loves(uncle(x), y), loves(z, aunt(z)):
 - \circ $z \rightarrow uncle(x), y \rightarrow aunt(uncle(x))$
 - loves(uncle(x), aunt(uncle(x)))

Quiz

Can we unify
 knows(John, x) knows(x, Mary)

What about
 knows(John, x) knows(y, Mary)

Standardizing apart

Can we unify
 knows(John, x) knows(x, Mary)

No!

What about

knows(John, x) knows(y, Mary)

 $x \rightarrow Mary, y \rightarrow John$

Standardize apart

- But knows(x, Mary) is logically equivalent to knows(y, Mary)!
- Moral: standardize apart before unifying

Most general unifier

- May be many substitutions that unify two formulas
- MGU is unique (up to renaming)
- Simple, moderately fast algorithm for finding MGU (see RN); more complex, linear-time algorithm

Linear unification. MS Paterson, MN Wegman. Proceedings of the eighth annual ACM symposium on Theory of Computing, 1976.

First-order resolution

- Given clauses ($a \lor b \lor c$), ($\neg c' \lor d \lor e$), and a substitution list V unifying c and c'
- \circ Conclude (a \lor b \lor d \lor e) : V

Example

rains no outside (x) => wet (x)

wet (x)=> rusty (x) v rust proof (x)

robot (x)=> ~ rust proof (x)

rains

guidebot (Roby)

guidebot (A) => robot (x) no outside (x)

First-order factoring

- When removing redundant literals, we have the option of unifying them first
- \circ Given clause (a \lor b \lor c), substitution V
- If a: V and b: V are the same
- Then we can conclude $(a \lor c) : V$

Completeness

- First-order resolution (together with firstorder factoring) is sound and complete for FOL
- Famous theorem

Completeness

Proof strategy

- We'll show FOL completeness by reducing to propositional completeness
- To prove S, put $KB \land \neg S$ in clause form
- Turn FOL KB into propositional KBs
 - in general, infinitely many
- Check each one in order
- If any one is unsatisfiable, we will have our proof

Propositionalization

- Given a FOL KB in clause form
- And a set of terms U (for universe)
- We can propositionalize KB under U by substituting elements of U for free variables in all combinations

Propositionalization example

- \circ $(\neg man(x) \lor mortal(x))$
- mortal(Socrates)
- favorite_drink(Socrates) = hemlock
- drinks(x, favorite_drink(x))

 \circ $U = \{Socrates, hemlock, Fred\}$

Propositionalization example

- (¬man(Socrates) ∨ mortal(Socrates))
 (¬man(Fred) ∨ mortal(Fred))
 (¬man(hemlock) ∨ mortal(hemlock))
- drinks(Socrates, favorite_drink(Socrates))
 drinks(hemlock, favorite_drink(hemlock))
 drinks(Fred, favorite_drink(Fred))
- mortal(Socrates) \(\) favorite_drink(Socrates) = hemlock

Choosing a universe

- To check a FOL KB, propositionalize it using some universe U
- Which universe?

Herbrand Universe



• **Herbrand universe** H of formula S:

Jacques Herbrand 1908–1931

- start with all objects mentioned in S
- or synthetic object X if none mentioned
- apply all functions mentioned in S to all combinations of objects in H, add to H
- repeat

Herbrand Universe

```
    E.g., loves(uncle(John), Mary) yields
    H = {John, Mary, uncle(John), uncle(Mary), uncle(uncle(John)), uncle(uncle(uncle(Mary)), ...}
```

Herbrand's theorem

- If a FOL KB in clause form is unsatisfiable
- And H is its Herbrand universe
- Then the propositionalized KB is unsatisfiable for some **finite** $U \subseteq H$

Significance

 This is one half of the equivalence we want: unsatisfiable FOL KB ⇒ ∃ finite U.
 unsatisfiable propositional KB

Example

- \circ (¬man(x) ∨ mortal(x)) ∧ man(uncle(Socrates)) ∧ ¬mortal(x)
- $\circ H = \{S, u(S), u(u(S)), \dots\}$
- ∘ If $U = \{u(S)\}$, PKB = $(\neg man(u(S)) \lor mortal(u(S))) \land man(u(S)) \land \neg mortal(u(S))$
- Resolving twice yields F

Converse of Herbrand

- A. J. Robinson proved "lifting lemma"
- Write PKB for a propositionalization of KB (under some universe)
- Any resolution proof in PKB corresponds to a resolution proof in KB
- ...and, if PKB is unsatisfiable, there is a proof of F (by prop. completeness); so, lifting it shows KB unsatisfiable

Example

- $\circ (\neg man(u(S)) \lor mortal(u(S))) \land man(u(S))$ $\land \neg mortal(u(S))$
- We resolved on man(u(S)) yielding mortal(u(S))
- Lifted, resolve $\neg man(x)$ w/ man(u(S)), binding $x \rightarrow u(S)$

Proofs w/ Herbrand & Robinson

- So, FOL KB is unsatisfiable if and only if there is a subset of its Herbrand universe making PKB unsatisfiable
- I.e., if we have a way to find proofs in propositional logic, we have a way to find them in FOL

Proofs w/ Herbrand & Robinson

- To prove S, put $KB \land \neg S$ in CNF: KB'
- Build subsets of Herbrand universe in increasing order of size: $U_1, U_2, ...$
- \circ Propositionalize KB' w/ U_i , look for proof
- If U_i unsatisfiable, use lifting to get a contradiction in KB'
- \circ If U_i satisfiable, move on to U_{i+1}

How long will this take?

- If S is not entailed, we will never find a contradiction
- In this case, if H infinite, we'll never stop
- So, entailment is **semidecidable**
 - equivalently, entailed statements are recursively enumerable

Variation

- Restrict semantics so we only need to check one finite propositional KB
- Unique names: objects with different names are different (John ≠ Mary)
- **Domain closure**: objects without names given in KB don't exist
- Restrictions also make entailment, validity feasible

Who? What? Who where?

Wh-questions

- We've shown how to prove a statement like mortal(Socrates)
- What if we have a question whose answer is not just yes/no, like "who killed JR?" or "where is my robot?"
- Simplest approach: prove $\exists x$. killed(x, JR), hope the proof is constructive

Answer literals

- Simple approach doesn't always work
- Instead of $\neg S(x)$, add $(\neg S(x) \lor answer(x))$
- If there's a contradiction, we can eliminate $\neg S(x)$ by resolution and unification, leaving answer(x) with x bound to a value that causes a contradiction

Example

Wills (Jack, Ca+) VW-lls (Curiosity, Ca+)

FOL Extensions

Equality

- **Paramodulation** is sound and complete for FOL+equality (see RN)
- Or, resolution + axiom schema

Second order logic

- SOL adds quantification over predicates
- E.g., principle of mathematical induction:

$$\bullet \quad \forall P. P(0) \land (\forall x. P(x) \Rightarrow P(S(x)))$$

$$\Rightarrow \forall x. P(x)$$

• There is no sound and complete inference procedure for SOL (Gödel's famous incompleteness theorem)

Others

- Temporal logics ("P(x) will be true at some time in the future")
- Modal logics ("John believes P(x)")
- Nonmonotonic FOL
- First-class functions (lambda operator, application)

0 ...

Using FOL

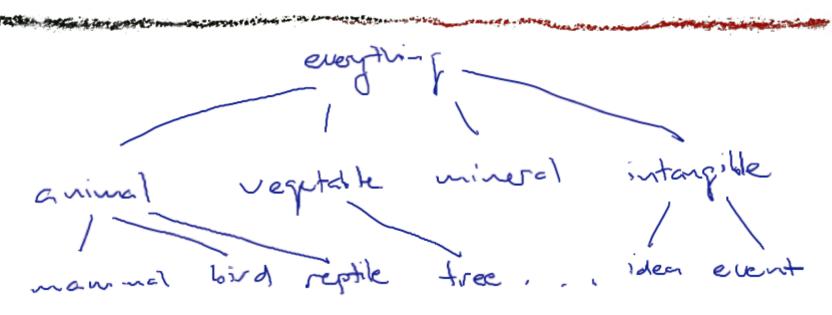
Knowledge engineering

- Identify relevant objects, functions, and predicates
- Encode general background knowledge about domain (reusable)
- Encode specific problem instance
- Pose queries (is P(x) true? Find x such that P(x))

Common themes

- RN identifies many common idioms and problems for knowledge engineering
- Hierarchies, fluents, knowledge, belief, ...
- We'll look at a couple

Taxonomies



- isa(Mammal, Animal)
- disjoint(Animal, Vegetable)
- partition({Animal, Vegetable, Mineral, Intangible}, Everything)

Inheritance

- Transitive: $isa(x, y) \land isa(y, z) \Rightarrow isa(x, z)$
- Attach properties anywhere in hierarchy
 - isa(Pigeon, Bird)
 - $\circ isa(x, Bird) \Rightarrow flies(x)$
 - $\circ isa(x, Pigeon) \Rightarrow gray(x)$
- So, isa(Tweety, Pigeon) tells us Tweety is gray and flies

Physical composition

- partOf(Wean4625, WeanHall)
- partOf(water37, water)
- Note distinction between mass and count nouns: any partOf a mass noun is also an example of that same mass noun

Fluents

- Fluent = property that changes over time
 at(Robot, Wean4623, 11AM)
- Actions change fluents
- Fluents chain together to form possible worlds
- $\circ at(x, p, t) \land adj(p, q) \Rightarrow poss(go(x, p, q)) \land at(x, q, result(go(x, p, q), t))$

Frame problem

- Suppose we execute an unrelated action (e.g., talk(Professor, FOL))
- Robot shouldn't move:
 - if at(Robot, Wean4623, t), want at(Robot, Wean4623, result(talk(Professor, FOL)))
- But we can't prove it using tools described so far!

Frame problem

- The **frame problem** is that it's a pain to list all of the things that don't change when we execute an action
- Naive solution: frame axioms
 - for each fluent, list actions that can't change fluent
 - KB size: O(AF) for A actions, F fluents

Frame problem

- Better solution: successor-state axioms
- For each fluent, list actions that can change it (typically fewer): if go(x, p, q) is possible,

```
at(x, q, result(a, t)) \Leftrightarrow

a = go(x, p, q) \lor (at(x, q, t) \land a \neq go(y, z))
```

• Size O(AE+F) if each action has E effects

Sadly, also necessary...

- Debug knowledge base
 - Severe bug: logical contradictions
 - Less severe: undesired conclusions
 - Least severe: missing conclusions
- First 2: trace back chain of reasoning until reason for failure is revealed
- Last: trace desired proof, find what's missing

SAT

The state of the s

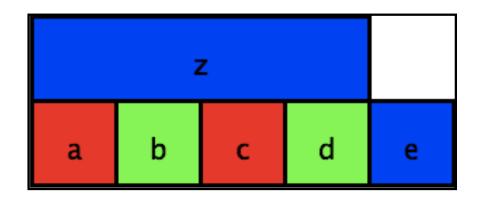
Satisfiability

- SAT is the problem of determining whether a given propositional logic sentence is satisfiable
- A decision problem: given an instance, answer yes or no
- A fundamental problem in CS

SAT is a general problem

- Many other useful decision problems reduce to SAT
- Informally, if we can solve SAT, we can solve these other problems
- So a good SAT solver is a good AI building block

Example decision problem



 k-coloring: can we color a map using only k colors in a way that keeps neighboring regions from being the same color?

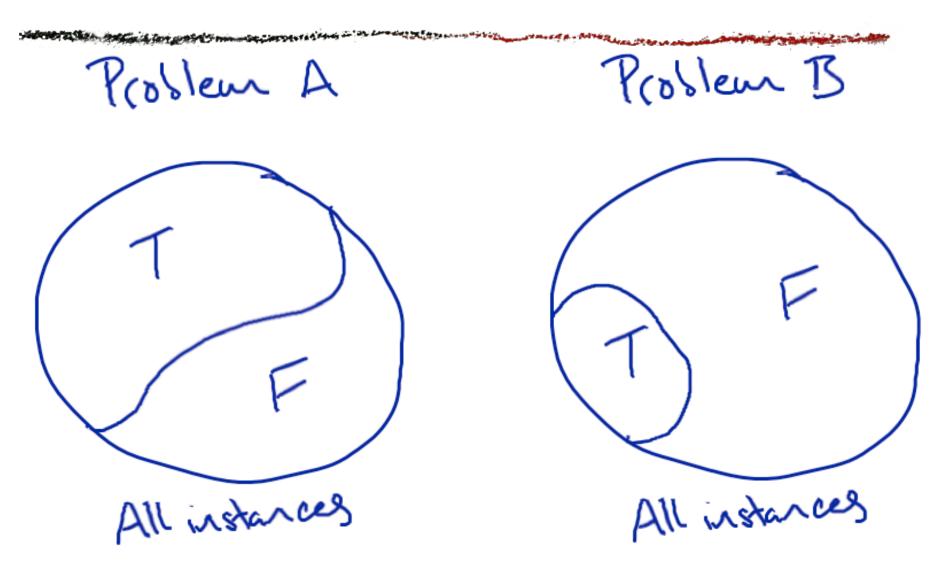
Example decision problem

- Propositional planning: given a list of operators (w/ preconditions, effects), can we apply operators in some order to achieve a desired goal?
- Have cake & eat it too example
- We'll see later how to represent as SAT

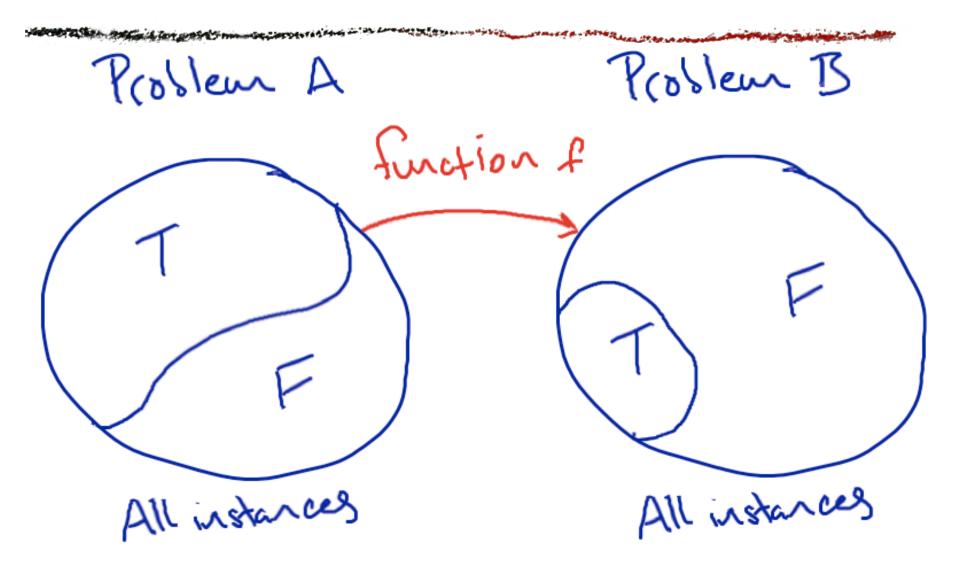
Reduction

- Loosely, "A reduces to B" means that if we can solve B then we can solve A
- More formally, A, B are decision problems (instances → truth values)
- A reduction is a poly-time function f such that, given an instance a of A
 - \circ f(a) is an instance of B, and
 - $\circ A(a) = B(f(a))$

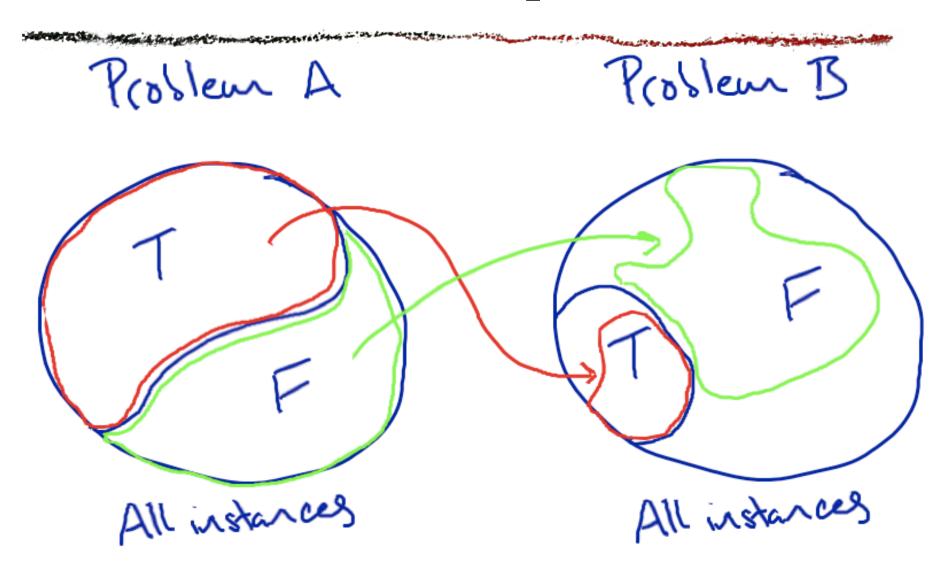
Reduction picture



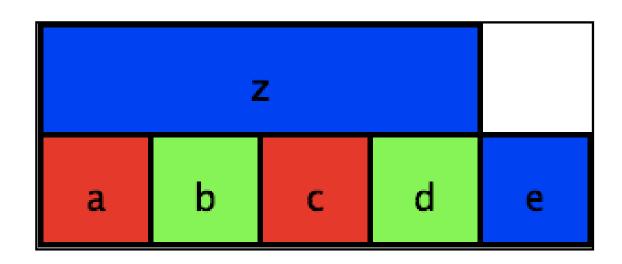
Reduction picture



Reduction picture



Back to example

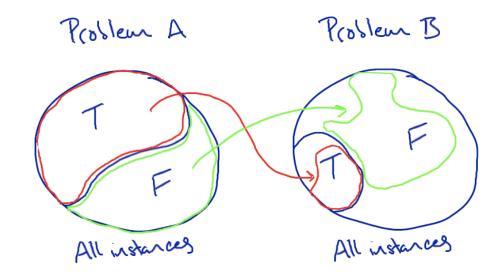


- Each square must be red, green, or blue
- Adjacent squares can't both be red (similarly, green or blue)

Back to example

- $(a_r \vee a_g \vee a_b) \wedge (b_r \vee b_g \vee b_b) \wedge (c_r \vee c_g \vee c_b) \wedge (d_r \vee d_g \vee d_b) \wedge (e_r \vee e_g \vee e_b) \wedge (z_r \vee z_g \vee z_b)$
- $\circ (\neg a_r \vee \neg b_r) \wedge (\neg a_g \vee \neg b_g) \wedge (\neg a_b \vee \neg b_b)$
- $\circ (\neg a_r \lor \neg z_r) \land (\neg a_g \lor \neg z_g) \land (\neg a_b \lor \neg z_b)$
- 0 ...

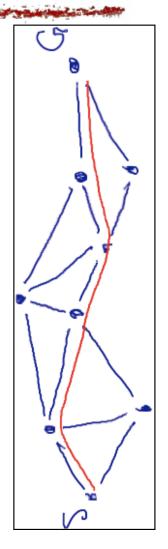
Direction of reduction



- If A reduces to B:
 - if we can solve B, we can solve A
 - so B must be at least as hard as A
- Trivially, can take an easy problem and reduce it to a hard one

Not-so-useful reduction

- Path planning reduces to SAT
- Variables: is edge e in path?
- Constraints:
 - exactly 1 path-edge touches start
 - exactly 1 path-edge touches goal
 - either 0 or 2 touch each other node



More useful: reduction to 3SAT

- We saw that decision problems can be reduced to SAT
 - is CNF formula satisfiable?
- Can reduce even further, to 3SAT
 - is 3CNF formula satisfiable?
- Useful if reducing SAT/3SAT to another problem (to show other problem hard)

Reduction to 3SAT

- Must get rid of long clauses
- \circ *E.g.*, $(a \lor \neg b \lor c \lor d \lor e \lor \neg f)$
- Replace with

$$(a \lor \neg b \lor x) \land (\neg x \lor c \lor y) \land (\neg y \lor d \lor z) \land (\neg z \lor e \lor \neg f)$$

NP-completeness

- S.A. Cook in 1971 proved that many useful decision problems reduce back and forth to SAT
 - showed how to simulate poly-sizememory computer w/ (very complicated, but still poly-size) SAT problem
- Equivalently, SAT is exactly as hard (in theory at least) as these other problems

Cost of reduction

- Complexity theorists often ignore little things like constant factors (or even polynomial factors!)
- So, is it a good idea to reduce your decision problem to SAT?
- Answer: sometimes...

Cost of reduction

- \circ SAT is well studied \Rightarrow fast solvers
- So, if there is an efficient reduction, ability to use fast SAT solvers can be a win
 - e.g., 3-coloring
 - another example later (SATplan)
- Other times, cost of reduction is too high
 - usu. because instance gets bigger
 - will also see example later (MILP)

Choosing a reduction

- May be many reductions from problem A to problem B
- May have wildly different properties
 - e.g., solving transformed instance may take seconds vs. days