

Knowledge-Based Agents

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Agenda

Knowledge-Based Agents

Core Concepts

Propositional Logic

Debugging

Knowledge-Based Agents

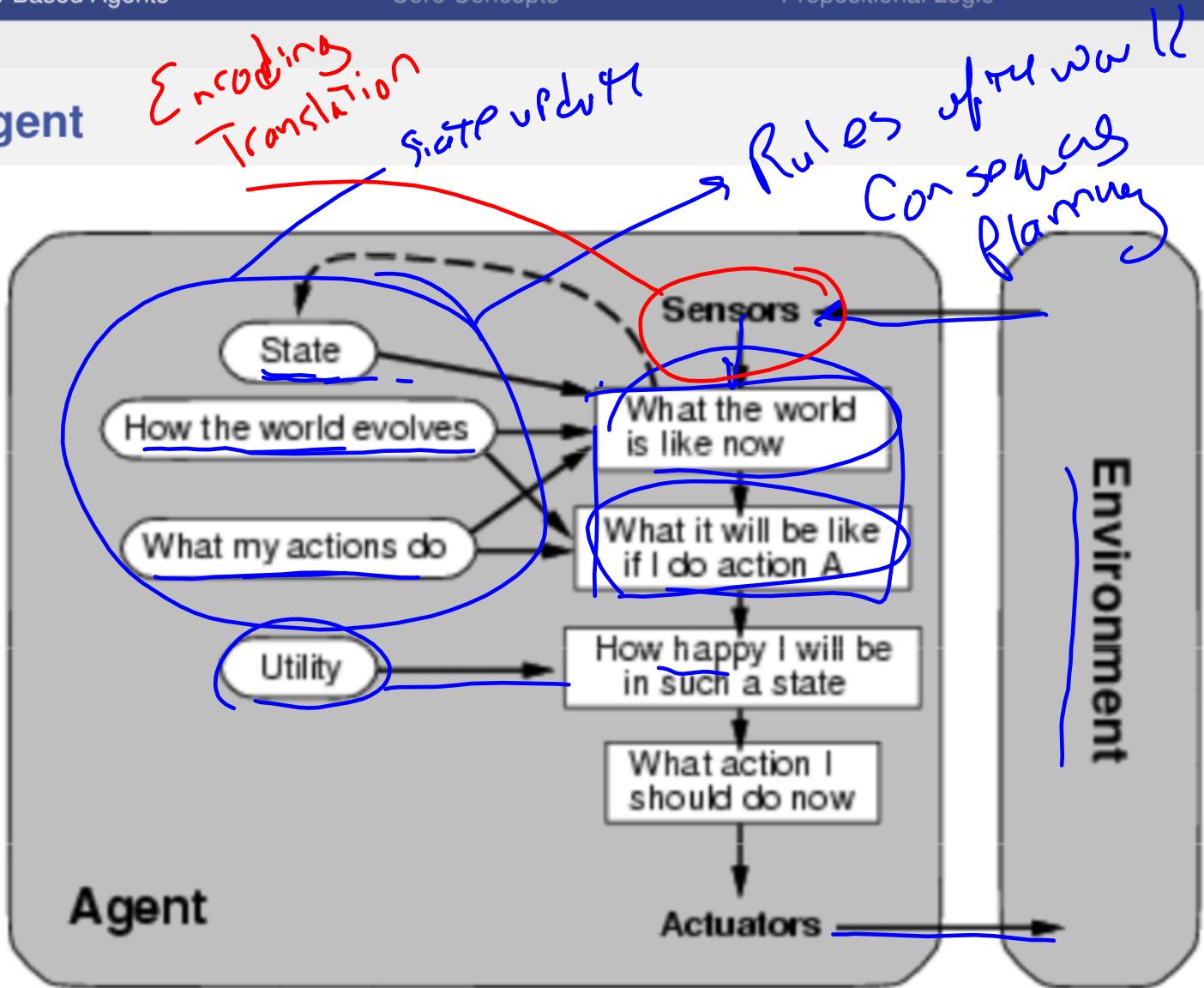
Representations

- ▶ Atomic (Search)

- ▶ Factored (Constraints)
- ▶ **Structured (TBD)**



An Agent

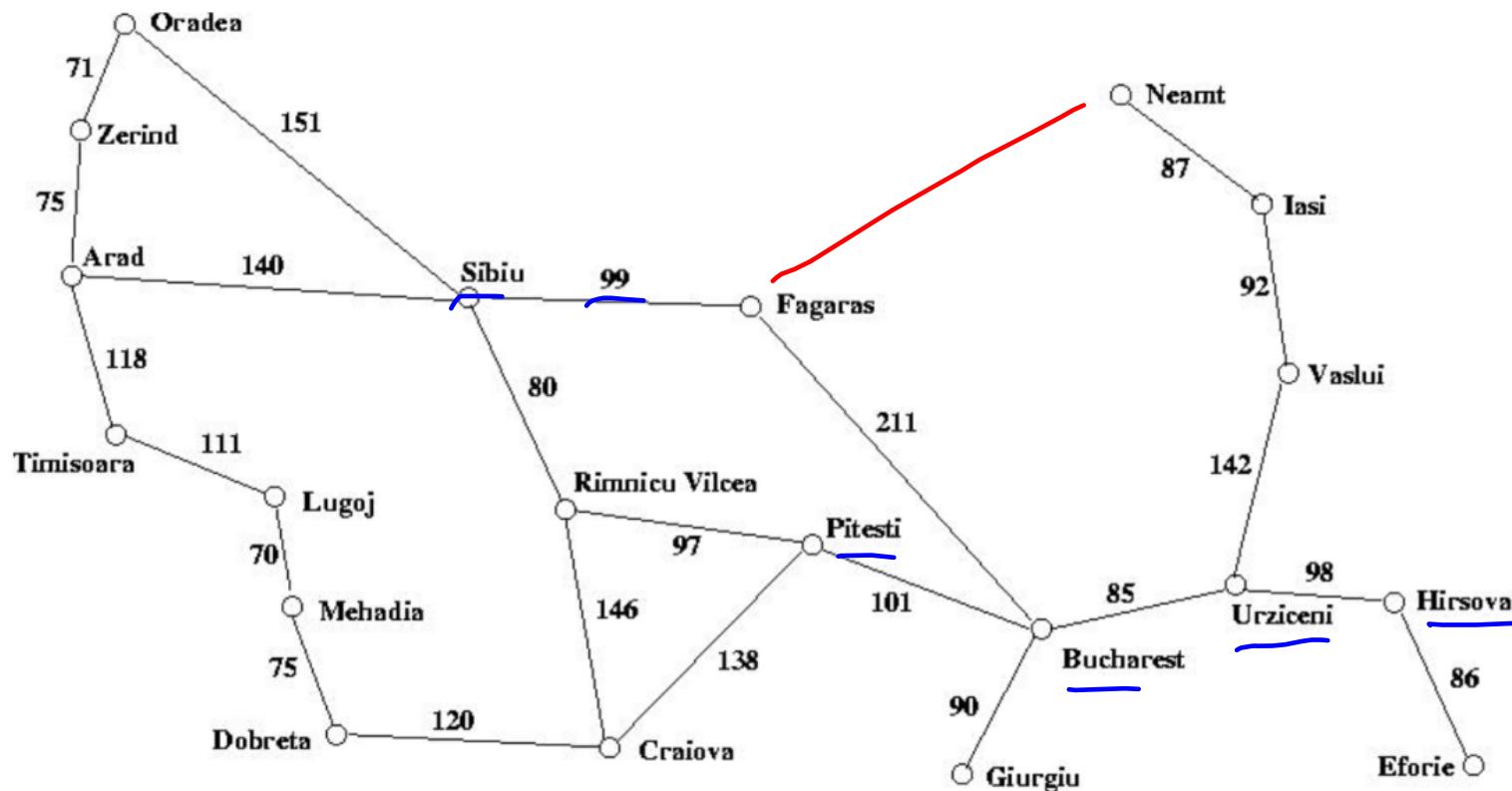


The Goal

- ▶ Domain-General Reasoning,
- ▶ Applied to Domain-Specific Knowledge
- ▶ (In a Domain-General Representation)

Standard Algo's
Problem implmtn.
Standard lang
Domain Survey
Doku

Implicit

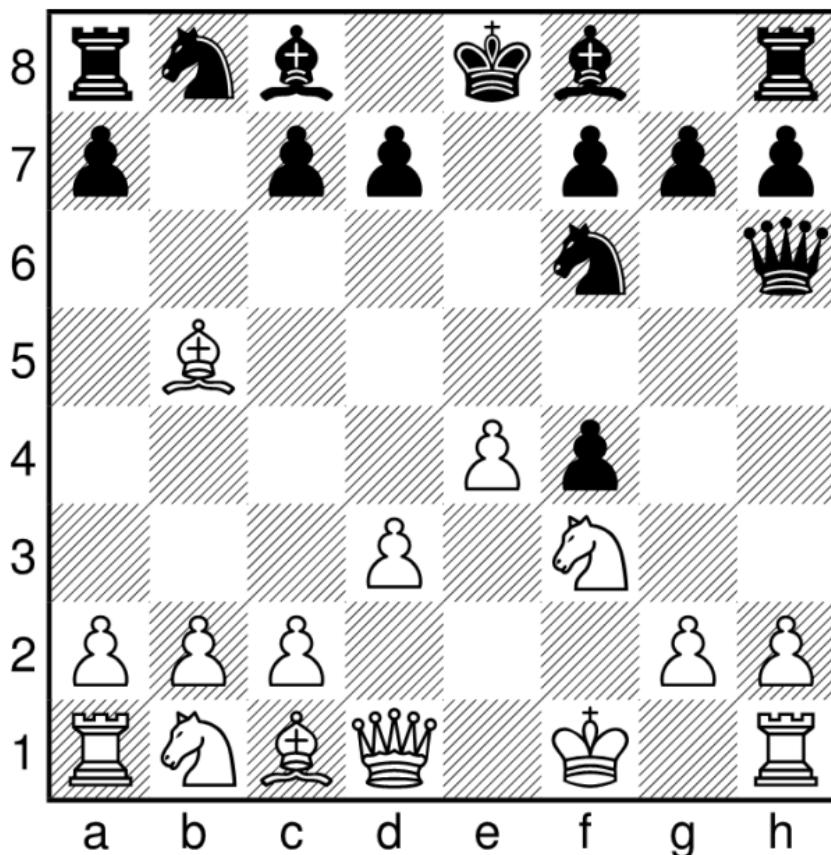


Implicit Explicit

1 road(oradea, zerind, 71).
2 road(zerind, arad, 75).
3
4 city(arad, 46.16667, 21.3).
5 city(bucharest, 44.41667, 26.1).
6
7 twoWayRoad(City1, City2, Distance):-
8 road(City1, City2, Distance).
9 twoWayRoad(City1, City2, Distance):-
10 road(City2, City1, Distance).
11
12 move(City1, City2):- twoWayRoad(City1, City2, _).
13
14 move(City1, City2, Distance):-
15 twoWayRoad(City1, City2, Distance).

Prolog

Chess Examples



1. C3.is clear
2. B3.is clear
3. A3.is clear
4. WKn1 in b1
5. Can move WKn1 to C3
6. \neg can move WQ to C3

Types of Reasoning

- ▶ Knowledge Representations:
 - ▶ Logic: propositional, predicate, temporal, fuzzy, probabilistic, etc. etc.
 - ▶ Networks of all kinds, including so-called Semantic Networks
 - ▶ Frames
 - ▶ Semantic Web: RDF, OWL, XML, etc.
 - ▶ English, Swahili, Urdu, etc. (for humans).
- ▶ Reasoning:
 - ▶ Logical inference
 - ▶ Temporal reasoning
 - ▶ Graph/subgraph matching
 - ▶ Analogical reasoning
 - ▶ Probabilistic inference

Assessment Criteria

- ▶ Assessment Criteria
 - ▶ **Representational Adequacy**: the ability to represent *sufficient* kinds of knowledge in the domain.
 - ▶ **Acquisitional Efficiency**: the ability to *easily* and *quickly* represent new knowledge in the formalism.
- ▶ Reasoning Methods
 - ▶ **Inferential Adequacy**: the ability to manipulate representational structures to derive *sufficient* new ones
 - ▶ **Inferential Efficiency**: the ability to derive new representational structures quickly

KB-Based Agent: Schema

```
1  class KB_Agent:  
2      Knowledge_Base = {...}    DB Infer  
3      Curr_Time = 0           Encoding / Decoding  
4  
5  def get_action(Percept):  
6      Percept_Seq = make_percept_sequence(  
7          Percept, self.Curr_time)  
8      tell(self.Knowledge_Base, Percept_Seq)  
9      Query = make_action_query(self.Curr_Time)  
10     Act = ask(self.Knowledge_Base, Query)  
11     Act_Seq = make_action_sequence(  
12         Act, self.Curr_Time)  
13     tell(self.Knowledge_Base, Act_Seq)  
14     self.Curr_Time += 1  
15     return Act
```

Expert Systems

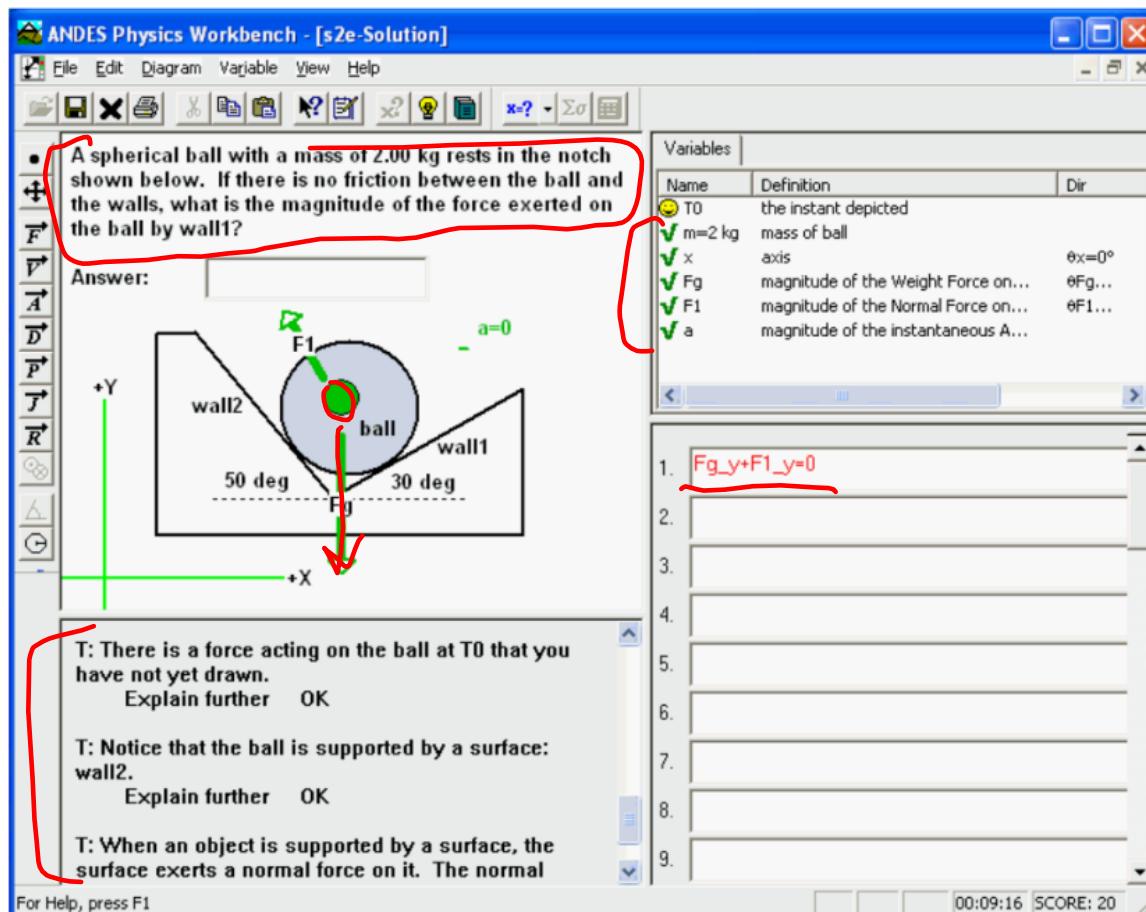


Image Credit: ubiquity.acm.org

Core Concepts

The basics

- ▶ Statements Individual facts, percepts, or information.
- ▶ Knowledge The integration of those statements to support reasoning.
- ▶ Inference The induction of new knowledge or decisionmaking.

Andes

KB \rightarrow JI \rightarrow G

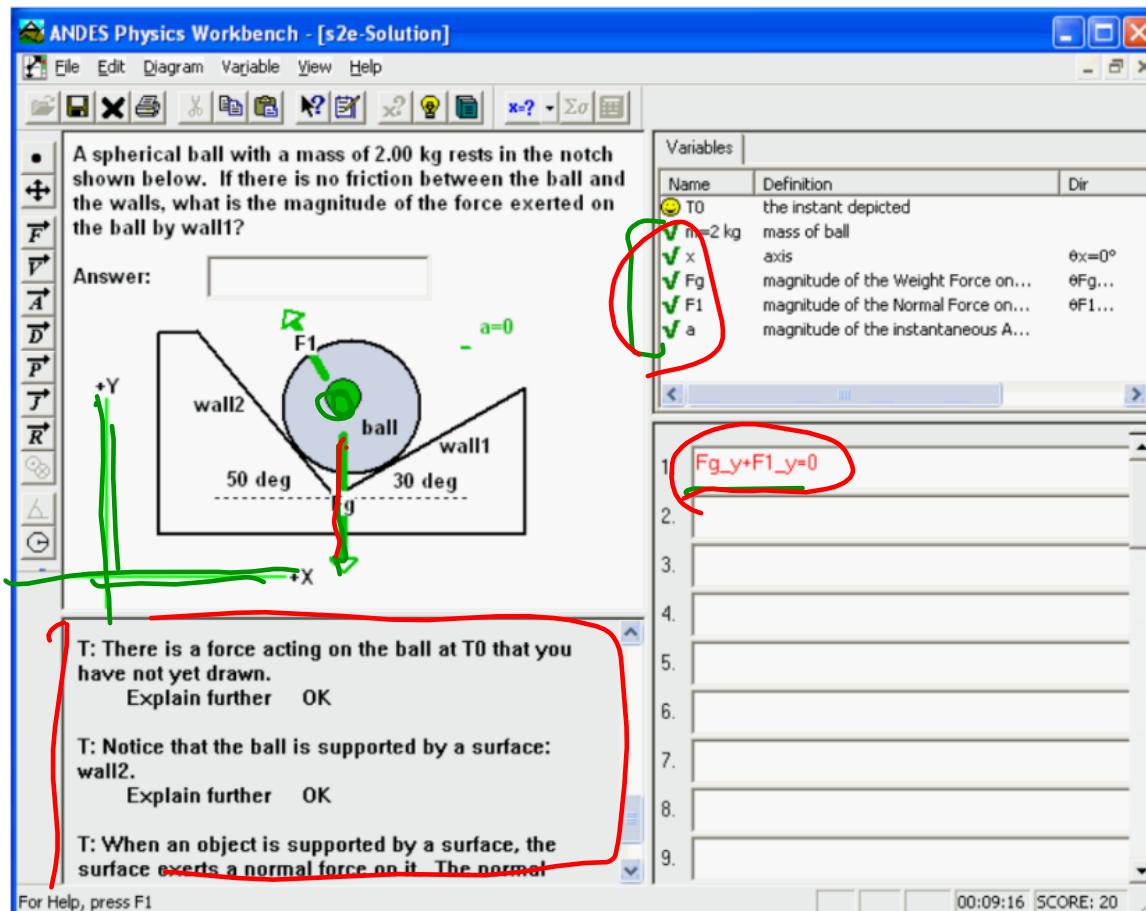


Image Credit: ubiquity.acm.org

Problem

A spherical ball with a mass of 2.00 kg rests in the notch shown below. If there is no friction between the ball and the walls, what is the magnitude of the force exerted on the ball by wall1?

KB

- ▶ (DRAW-AXES 40)
- ▶ (EQN (= |Yc_Fn_BALL_WALL2_1_40| 0))
- ▶ (IMPLICIT-EQN (= |OFn_BALL_WALL2_1| (DNUM 40 |deg|)))
- ▶ (VECTOR (FORCE BALL WALL2 NORMAL :TIME 1) (DNUM 40 |deg|))
- ▶ (DEFINE-VAR (MASS BALL))
- ▶ (EQN (= |Yc_Fw_BALL_EARTH_1_40| (* |Fw_BALL_EARTH_1| (SIN (- (DNUM 270 |deg|) (DNUM 40 |deg|))))))

The basics

- ▶ **Sentences** (*statements*) in the knowledge base.
- ▶ **Syntax**: rules for well formed sentences.
- ▶ **Semantics**: rules for the meaning of the sentence or *truth in each possible world*.
- ▶ Sentences must be true XOR false in each possible world.
- ▶ **Model**: A fixed assignment of values to all possible sentences (possible world).
- ▶ A sentence **S satisfies** a model if it is true in it.
- ▶ **M is a model of S** if S satisfies it.

Propositional Logic

Basic Elements

- ▶ Set P of *propositional symbols*: p, q, r , etc.
- ▶ Two *truth values*: true, false
— —
- ▶ *Logical Symbols*: \neg , \wedge , \vee , \Rightarrow , \Leftrightarrow , \Box
- ▶ *Punctuation*: (), [], { }.

If (Is runs \wedge Not Error).
Continuous

Propositions

- ▶ A symbol: \underline{a}
- ▶ a negated proposition: $\underline{\neg}P$
- ▶ Conjunction or Disjunction: $(P_1 \wedge P_2)$ $(P_1 \vee P_2)$
- ▶ An implication: $P_1 \Rightarrow P_2$ $\neg P_1 \vee P_2$
- ▶ A logical equivalence: $P_1 \Leftrightarrow P_2$. $(P_1 \rightarrow P_2) \wedge (P_2 \rightarrow P_1)$
- ▶ Note: Implication is not causation!

Semantics

- ▶ An interpretation is a specification of values (t—f) for all symbols:

$$m_i = \{s_0 = t, s_1 = f, \dots\} \quad (1)$$

- ▶ Sentence values are constructed recursively from symbols up.
- ▶ Or stored in truth-tables.
- ▶ A model for a proposition (p) is an interpretation in which p is true.
- ▶ A proposition p is valid if for any interpretation I, I $\models p$ i.e., every interpretation is a model.

$$\overline{(a \vee (\neg b \wedge c) \vee d)}$$

Model Checking

KB ⊢ P

- ▶ p is *satisfiable* if \exists an I s.t. $I \models p$.
- ▶ p is *unsatisfiable* no valid model can be found.
- ▶ Theorem proving can be handled via DFS or rule algorithms.
- ▶ What is wrong with that?

Logic
consistency

Complex

Theorem Proving

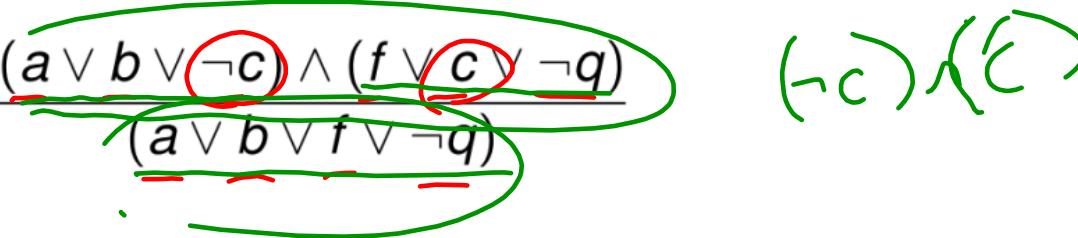
- ▶ Logical Equivalence $a \equiv b$ iff $(a \models b) \wedge (b \models a)$.
- ▶ Valid (Tautology) $\forall m : \underline{m \models p}$
- ▶ Deduction Theorem: $(a \models b)$ iff $(\underline{a} \rightarrow \underline{b})$.
- ▶ Thus: $(a \models b)$ iff $(\underline{a} \wedge \neg \underline{b})$ is unsatisfiable.

Resolution

- ▶ Convert all Knowledge to CNF ($(a \vee b \vee \dots) \wedge (c \vee \dots) \wedge \dots$)
- ▶ *Resolve* conflicting clauses to yield new information

$$\frac{(a \vee b \vee \neg c) \wedge (f \vee c \vee \neg q)}{(a \vee b \vee f \vee \neg q)}$$

$(\neg c) \wedge (\neg c)$

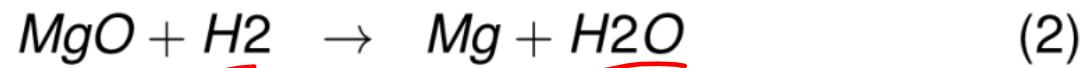


- ▶ This process is sound and complete.

Definite & Horn Clauses

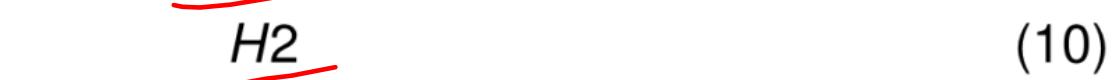
- ▶ Resolution is powerful but unnecessary with *Horn Clauses*
 - ▶ At most one variable is positive: $(x \vee \neg y \vee \dots)$
 - ▶ These can be handled with implications $(p \wedge q) \rightarrow \underline{r}$
- ▶ Can do Forward-chaining *data-driven* reasoning.
 - ▶ And backward-chaining *goal-directed* reasoning.

Example: Chemistry



Assuming sufficient quantities of MgO, H₂, C, and O₂, prove that we can synthesize H₂CO₃.

Example: Chemistry (Propositional)



Conclusion

$\underline{\underline{H_2CO_3}}$?

Example: Chemistry (CNF)

1. $\neg MgO \vee \neg H_2 \vee Mg$
2. $\neg MgO \vee \neg H_2 \vee H_2O$
3. $\neg C \vee \neg O_2 \vee CO_2$
4. $\neg CO_2 \vee \neg H_2O \vee H_2CO_3$

5. MgO
6. H_2
7. C
8. O_2
9. $\neg H_2CO_3$

Negated conclusion

$(a \wedge \neg a) \perp$

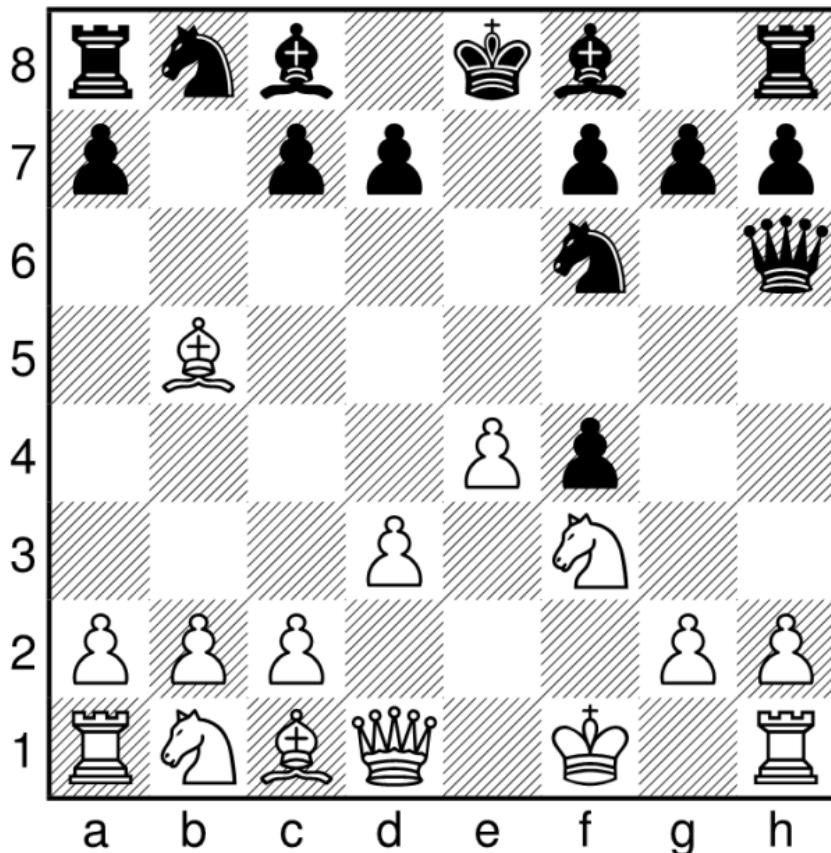
Law of non-Excluded middle

Example: Chemistry (CNF)

- | | | |
|-----|---------------------------------------|---------------|
| 10. | $\neg CO_2 \vee \neg H_2O$ | <u>4 + 9</u> |
| 11. | $\neg C \vee \neg O_2 \vee \neg H_2O$ | <u>3 + 10</u> |
| 12. | $\neg O_2 \vee \neg H_2O$ | <u>7 + 11</u> |
| 13. | $\neg H_2O$ | <u>8 + 12</u> |
| 14. | $\neg MgO \vee \neg H_2$ | <u>2 + 13</u> |
| 15. | $\neg H_2$ | <u>5 + 14</u> |
| 16. | \square | <u>6 + 15</u> |

Planning & Frames

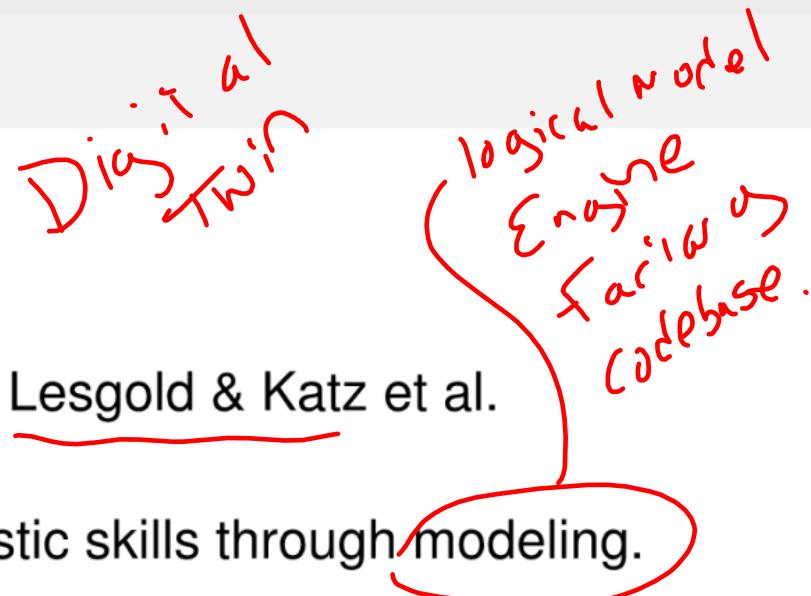
We can even use this to do planning (sort of)



1. clear_{C3}^{t1}
2. clear_{B3}^{t1}
3. clear_{A3}^{t1}
4. $WKn1_{b1}^{t1}$
5. $\text{Move}_{WKn1 \text{ to } c3} \Leftrightarrow (WKn1_{b1}^{t1} \wedge \text{clear}_{C3}^{t1})$

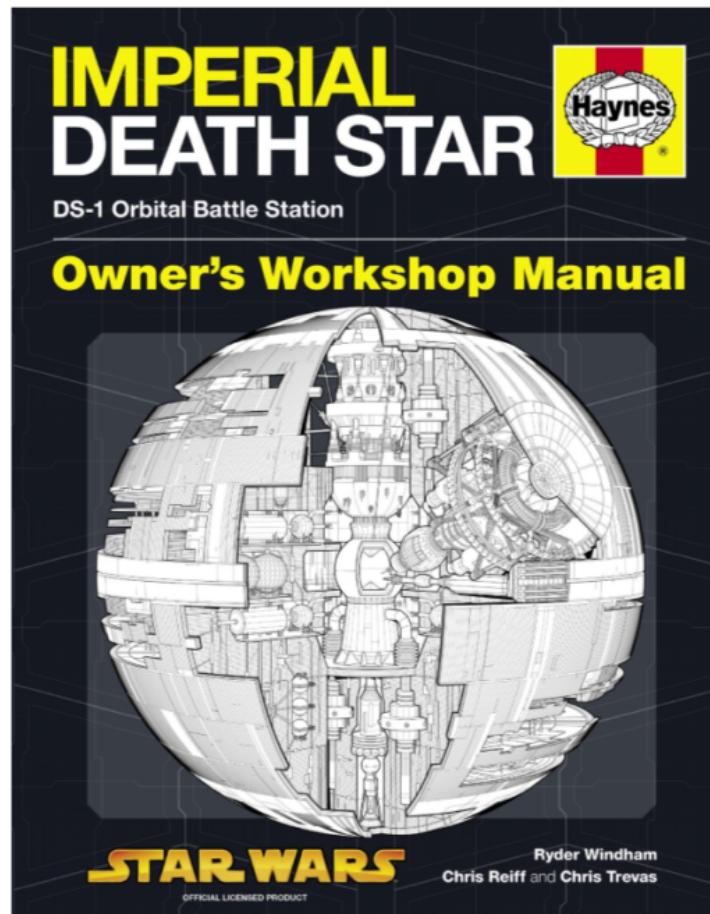
Debugging

Real Problem: Sherlock



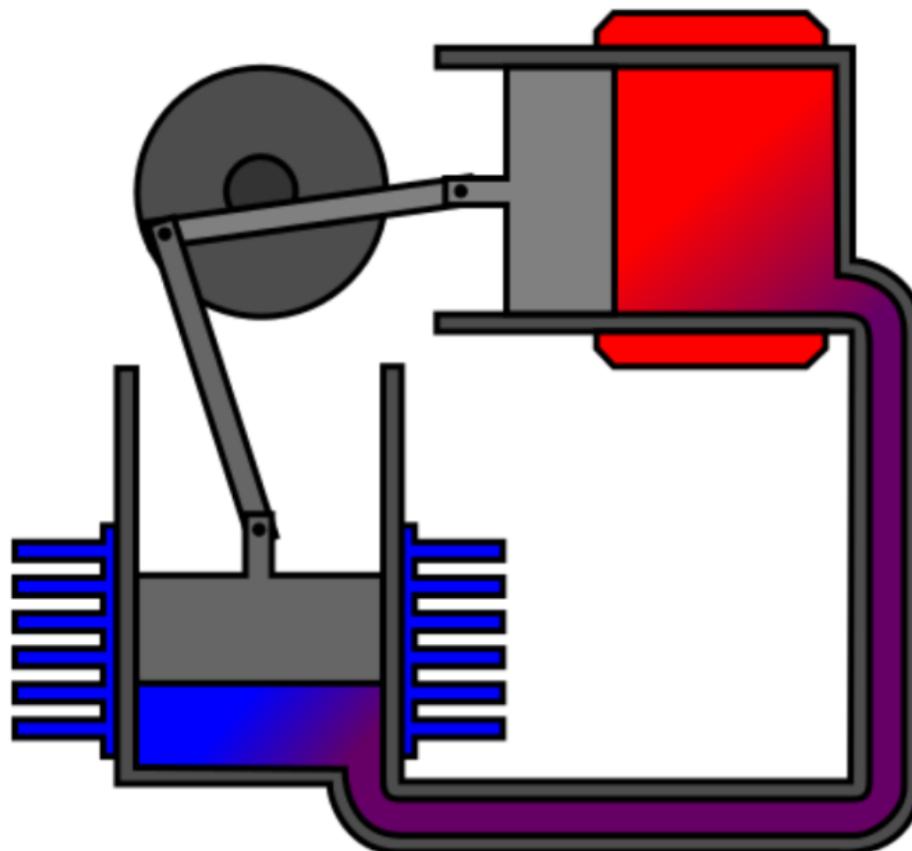
- ▶ The Sherlock system: Lesgold & Katz et al.
- ▶ Teaches basic diagnostic skills through modeling.
- ▶ Uses a logical model of student and system.
- ▶ Saved \$>500k in one instance.

Service Debugging



Source: <http://www.fine1steditions.co.uk>

Basic Piston engine.



Source: https://en.wikipedia.org/wiki/File:Alpha_Stirling_frame_12.svg