Artificial Intelligence

Logical agents Chapter 7, AIMA

This presentation owes some to V. Pavlovic @ Rutgers and D. Byron @ OSU

Motivation for Knowledge Representation

- Search algorithms discussed previously are often called *meta-programming*
 - they are general, but it still is programming
 - the code needs to be specialised for every concrete application taking domain knowledge into account
- We need something more general
 - letting us to only specify the rules of the game
 - and use ,,out-of-the-box" reasoning engine







Goal: Get the gold



Problem 1: Big, hairy, smelly, dangerous Wumpus. Will eat you if you run into it, but you can smell it a block away.



Problem 2: Big, bottomless pits where you fall down. You can feel the breeze when you are near them.

PEAS description

Performance measure:

- +1000 for gold
- -1000 for being eaten or falling down pit
- -1 for each action
- -10 for using the arrow

Environment:

 4×4 grid of "rooms", each "room" can be empty, with gold, occupied by Wumpus, or with a pit.

Acuators:

Move forward, turn left 90°, turn right 90° Grab, shoot

Sensors:



PEAS description

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Actuators:

Move forward, turn left 90°, turn right 90° Grab, shoot

Sensors:

















Grab the gold and get out!



How do we automate this kind of reasoning? How can we make computers figure it out on their own?

Logic

Logic is a formal language for representing information in such a way that conclusions can be drawn

- A logic has
 - Syntax that specifies symbols in the language and how they can be combined to form *sentences*
 - **Semantics** that specifies what facts in the world these sentences refer to and assigns *truth values* to them based on their meaning in the world.
 - Inference procedure, a mechanical method for computing (deriving) new (true) sentences from existing (known) sentences.



Entailment

$\mathsf{A} \vDash \mathsf{B}$

The sentence A entails the sentence B

- If A is true, then B must also be true
- B is a "logical consequence" of A

Let's explore this concept a bit...

Example: Wumpus entailment

Agent's knowledge base (KB) after having visited (1,1) and (1,2):

- 1) The rules of the game (PEAS)
- 2) Nothing in (1,1)
- 3) Breeze in (1,2)

Which models (states of the world) match these observations?

Every possible world state is a *model* but not all are consistent with what we already know!



 $\mathbf{x}_{1,1} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$

Example: Wumpus entailment

We only care about neighboring rooms, i.e. {(2,1),(2,2),(1,3)}. We can't know anything about the other rooms.

We care about pits, because we have detected a breeze. We don't want to fall down a pit.

There are 2³=8 possible arrangements of {pit, no pit} in the three neighboring rooms.



Possible conclusions:

 α_1 : There is no pit in (2,1) α_2 : There is no pit in (2,2)

- i_2 . There is no pit in (2,2)
- α_3 : There is no pit in (1,3)

















The eight possible situations...





PIT

1

PIT

Breeze

2





The eight possible situations...

These are the ones that agree with our Knowledge Base (KB), i.e. the rules of the game and our observations.









PIT

1

1

PIT

Beere

2

PIT





...let's explore this conclusion





α_l : There is no pit in (2,1)

...let's explore this conclusion



- KB = The set of models that agrees with the knowledge base (the observed facts) [The KB is true in these models]
- α_{l} = The set of models that agrees with conclusion α_{l} [conclusion α_{l} is true in these models]



KB = The set of models that agrees with the knowledge base (the observed facts) [The KB is true in these models]

...let's try exploring this conclusion instead...

 α_2 = The set of models that agrees with conclusion α_2 [conclusion α_2 is true in these models]



- KB = The set of models that agrees with the knowledge base (the observed facts) [The KB is true in these models]
- α_3 = The set of models that agrees with conclusion α_3 [conclusion α_3 is true in these models]



- KB = The set of models that agrees with the knowledge base (the observed facts) [The KB is true in these models]
- α_3 = The set of models that agrees with conclusion α_3 [conclusion α_3 is true in these models]

Inference engine

- We need an algorithm that produces the entailed conclusions <u>automatically</u>
 - for any user-defined Knowledge Base
- Entailment is the most important and most commonly used property in logic
 - most of the things we are interested in can be expressed using entailment
- We will call such an algorithm, as well as it's implementation, an *inference engine*

Inference engine



 $\mathsf{KB} \vdash_i \mathsf{A}$

"A is derived from KB by inference engine i"

- **Truth-preserving:** *i* only derives entailed sentences
- **Complete:** *i* derives all entailed sentences

We want inference engines that are both truth-preserving and complete

Propositional (boolean) logic Syntax

Atomic sentence = a single propositional symbol e.g. P, Q, P₁₃, W₃₁, G₃₂, T, F Pit in room (1,3)

Wumpus in room (3,1)

Complex sentence = combinations of simpler

sentences, formed using connectives

 \neg (not) negation

 \wedge (and) conjunction $|P_{13} \wedge W_{31}|$

- v (or) disjunction $W_{31} \vee \neg W_{31}$
- \Rightarrow (implies) implication $W_{31} \Rightarrow S_{32}$
- \Leftrightarrow (iff = if and only if) biconditional/logical equality

Precedence: $\neg, \land, \lor, \Rightarrow, \Leftrightarrow$

Propositional (boolean) logic Semantics

Semantics: The rules for whether any given sentence is true or false

- T (true) is true in every model
- F (false) is false in every model
- The truth values for other propositional symbols are specified in the model

Atomic sentences

 Truth values for complex sentences are specified according to the definitions of connectives

 using a *truth table*

Boolean truth table

 $\neg P \quad P \land Q \quad P \lor Q \quad P \Rightarrow Q \quad P \Leftrightarrow Q$

PQFalseFalseFalseTrueTrueFalseTrueTrue

Please complete this table...
Р	Q	¬Ρ	$P \land Q$	$P\!\vee\!Q$	P⇒Q	P⇔Q
False	False	True				
False	True	True				
True	False	False				
True	True	False				

Not P is the opposite of P

Р	Q	$\neg P$	$P \land Q$	$P\!\vee\!Q$	P⇒Q	P⇔Q
False	False	True	False			
False	True	True	False			
True	False	False	False			
True	True	False	True			

 $P\,\wedge\,Q$ is true only when both P and Q are true

Р	Q	¬Ρ	$P \land Q$	$P{\vee}Q$	P⇒Q	P⇔Q
False	False	True	False	False		
False	True	True	False	True		
True	False	False	False	True		
True	True	False	True	True		

 $\mathsf{P}\,\vee\,\mathsf{Q}$ is true when either P or Q is true

Р	Q	¬Ρ	$P{\wedge}Q$	$P\!\vee\!Q$	P⇒Q	P⇔Q
False	False	True	False	False	True	
False	True	True	False	True	True	
True	False	False	False	True	False	
True	True	False	True	True	True	

 $P \Rightarrow Q$: If P is true then we claim that Q is true, otherwise we make no claim

Р	Q	$\neg P$	$P{\wedge}Q$	$P\!\vee\!Q$	P⇒Q	P⇔Q
False	False	True	False	False	True	True
False	True	True	False	True	True	False
True	False	False	False	True	False	False
True	True	False	True	True	True	True

 $P \Leftrightarrow Q$ is true when the truth values for P and Q are identical



The exlusive or (XOR) is different from the OR

Р	Q	P⊕Q
False	False	False
False	True	True
True	False	True
True	True	False

The exlusive or (XOR) is different from the OR

Example: Wumpus KB

Interesting sentences [tell us what is in <u>neighbour</u> squares]

Knowledge base





Nothing in (1,1)
Breeze in (1,2)

 $\mathsf{KB} = \mathsf{R}_1 \land \mathsf{R}_2 \land \mathsf{R}_3 \land \mathsf{R}_4 \land \mathsf{R}_5 \land \mathsf{R}_6 \land \mathsf{R}_7 \land \mathsf{R}_8 \land \mathsf{R}_9 \land \mathsf{R}_{10}$

Plus the rules of the game

Example: Wumpus KB

Those are the basic rules of the game



What is in squares (1,3), (2,1), and (2,2)?



What is in squares (1,3), (2,1), and (2,2)?



What do we deduce from this?

What is in squares (1,3), (2,1), and (2,2)?



 $\mathsf{KB} \vDash \neg \mathsf{W}_{21} \land \neg \mathsf{W}_{22} \land \neg \mathsf{W}_{13} \land \neg \mathsf{P}_{21}$

- Can be naturally implemented as a depth-first search on a constraint graph
 - with backtracking
- Time complexity ~ $O(2^n)$

where n is the number of relevant symbols

• Space complexity $\sim O(n)$

Not very impressive...

Although computers are really, really good with long sequences of 0s and 1s

Some more definitions

Equivalence:

 $A \equiv B \text{ iff } A \models B \text{ and } B \models A$

Validity: A valid sentence is one that is true in all the models (a tautology)

 $A \models B \text{ iff } (A \Rightarrow B)$ is valid

Satisfiability: A sentence is satisfiable if it is true in *at least one* model

 $A \models B \text{ iff } (A \land \neg B)$ is unsatisfiable

Let's explore *satisfiability* first...



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Some more definitions

Equivalence:

 $A \equiv B \text{ iff } A \models B \text{ and } B \models A$

For example, $A \equiv \neg(\neg A)$

 $A \models B$ means that the set of models where A is true is a subset of the models where B is true: $A \subseteq B$

 $B \vDash A$ means that the set of models where B is true is a subset of the models where A is true: $B \subseteq A$

Therefore, the set of models where A is true must be equal to the set of models where B is true: A \equiv B



Some more definitions

Validity: A valid sentence is true in all models (a tautology)

For example, $A \lor \neg A$ is valid

 $A \models B$ iff (A \Rightarrow B) is valid



А	В	A⇒B
False	False	True
False	True	True
True	False	False
True	True	True

Logical equivalences

(A ∧ B)≡	(B ∧ A)	∧ is commutative
(A ∨ B)≡	(B ∨ A)	\vee is commutative
$((A \land B) \land C) \equiv$	(A ∧ (B ∧ C))	\land is associative
$((A \lor B) \lor C) \equiv$	(A ∨ (B ∨ C))	\vee is associative
¬(¬A)≡	A	Double-negation elimination
$(A\RightarrowB)\equiv$	$(\neg B \Rightarrow \neg A)$	Contraposition
$(A\RightarrowB)\equiv$	(¬A ∨ B)	Implication elimination
(A ⇔ B)≡	$((A \Rightarrow B) \land (B \Rightarrow A))$	Biconditional elimination
¬(A ∧ B)≡	(¬A ∨ ¬B)	"De Morgan"
¬(A ∨ B)≡	(¬A ∧ ¬B)	"De Morgan"
$(A \land (B \lor C)) \equiv$	$((A \land B) \lor (A \land C))$	Distributivity of \land over \lor
$(A \lor (B \land C)) \equiv$	((A ∨ B) ∧ (A ∨ C))	Distributivity of \lor over \land

А	В	$A \wedge B$	¬(A ∧ B)	$\neg A$	¬Β	$\neg A \lor \neg B$
False	False					
False	True					
True	False					
True	True					

А	В	$A \wedge B$	¬(A ∧ B)	$\neg A$	$\neg B$	$\neg A \lor \neg B$
False	False	False				
False	True	False				
True	False	False				
True	True	True				

А	В	$A \wedge B$	¬(A ∧ B)	$\neg A$	$\neg B$	$\neg A \lor \neg B$
False	False	False	True			
False	True	False	True			
True	False	False	True			
True	True	True	False			

А	В	$A \wedge B$	¬(A ∧ B)	$\neg A$	$\neg B$	$\neg A \lor \neg B$
False	False	False	True	True	True	
False	True	False	True	True	False	
True	False	False	True	False	True	
True	True	True	False	False	False	

А	В	$A \wedge B$	¬(A ∧ B)	$\neg A$	$\neg B$	$\neg A \lor \neg B$
False	False	False	True	True	True	True
False	True	False	True	True	False	True
True	False	False	True	False	True	True
True	True	True	False	False	False	False

Logical equivalences

egation elimination

vity of \land over \lor

vity of \lor over \land

Work out some of these on paper for yourself...

Inference

- There are two main approaches towards automating the inference:
 - model enumeration
 - inference rules

Inference rules

• Inference rules are written as



If the KB contains the antecedent, you can add the consequent, because the KB is guaranteed to entail it

Slide adapted from D. Byron

Commonly used inference rules



 $\frac{A \Longrightarrow B, A}{B}$

A	B .	$A \Rightarrow$	В
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- 1 False False True
- 2 False True True
- 3 True False False
- 4 True True True







Proof for Unit Resolution

 $\frac{A \lor B, -B}{A}$

	A	В	$A \lor B$	$\neg A$	$\neg B$
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- 1 False False False True True
- 2 False True True True False
- 3 True False True False True
- 4 True True True False False

Proof for Unit Resolution <u>A\B,-B</u> A Α В $A \lor B \neg A$ $\neg B$ False False False 1 True True False 2 True True True False False True 3 True False True These are the cases when $A \vee B$ is True False False True True True 4



Proof for Unit Resolution $A \lor B, -B$ A Α B $A \lor B \neg A$ $\neg B$ 1 False False False True True 2 False False True True True This is the case True False True False 3 True when both $\neg B$ and True True True $A \vee B$ are True False False 4 A is also True here so we can safely add A = True to our KB
Commonly used inference rules



Slide adapted from D. Byron

Work out these on paper for yourself too

Example: Proof in Wumpus KB

Knowledge base





1. Nothing in (1,1)

 $\mathsf{B}_{11} \Leftrightarrow (\mathsf{P}_{12} \lor \mathsf{P}_{21})$

Rule of the game

 $B_{11} \Leftrightarrow (P_{12} \lor P_{21})$ Rule of the game $B_{11} \Rightarrow (P_{12} \lor P_{21}) \land (P_{12} \lor P_{21}) \Rightarrow B_{11}$ Biconditional
elimination

 $(A \Leftrightarrow B) \equiv ((A \Rightarrow B) \land (B \Rightarrow A))$

$$B_{11} \Rightarrow (P_{12} \lor P_{21}) \land (P_{12} \lor P_{21}) \Rightarrow B_{11}$$
$$(P_{12} \lor P_{21}) \Rightarrow B_{11}$$

Biconditional elimination And elimination

 $\frac{A \wedge B}{B}$

 $(P_{12} \lor P_{21}) \Rightarrow B_{11} \qquad An$ $\neg B_{11} \Rightarrow \neg (P_{12} \lor P_{21}) \qquad Co$

And elimination

Contraposition

 $(\mathsf{A} \Rightarrow \mathsf{B}) \equiv (\neg \mathsf{B} \Rightarrow \neg \mathsf{A})$

$$\neg(\mathsf{A}\lor\mathsf{B})\equiv(\neg\mathsf{A}\land\neg\mathsf{B})$$

$$\neg \mathsf{B}_{11} \Rightarrow \neg (\mathsf{P}_{12} \lor \mathsf{P}_{21})$$
$$\neg \mathsf{B}_{11} \Rightarrow \neg \mathsf{P}_{12} \land \neg \mathsf{P}_{21}$$

Contraposition

"De Morgan"

 $\begin{array}{ll} \mathsf{B}_{11} \Leftrightarrow (\mathsf{P}_{12} \lor \mathsf{P}_{21}) & \mathsf{Rule} \\ \mathsf{B}_{11} \Rightarrow (\mathsf{P}_{12} \lor \mathsf{P}_{21}) \land (\mathsf{P}_{12} \lor \mathsf{P}_{21}) \Rightarrow \mathsf{B}_{11} & \begin{array}{l} \underset{\mathsf{elim}}{\mathsf{Bicc}} \\ \mathsf{elim} \\ (\mathsf{P}_{12} \lor \mathsf{P}_{21}) \Rightarrow \mathsf{B}_{11} & \mathsf{Anc} \\ \neg \mathsf{B}_{11} \Rightarrow \neg (\mathsf{P}_{12} \lor \mathsf{P}_{21}) & \begin{array}{l} \mathsf{Corr} \\ \mathsf{Corr} \\ \neg \mathsf{B}_{11} \Rightarrow \neg \mathsf{P}_{12} \land \neg \mathsf{P}_{21} & & \end{array} \end{array}$

Rule of the game Biconditional elimination

And elimination

Contraposition

"De Morgan"

$B_{11} \Leftrightarrow (P_{12} \lor P_{21})$	Rule of the game
$B_{11} \Rightarrow (P_{12} \lor P_{21}) \land (P_{12} \lor P_{21}) \Rightarrow B_{11}$	Biconditional elimination
$(P_{12} \lor P_{21}) \Rightarrow B_{11}$	And elimination
$\neg B_{11} \Rightarrow \neg (P_{12} \lor P_{21})$	Contraposition
$\neg B_{11} \Rightarrow \neg P_{12} \land \neg P_{21}$	"De Morgan"

Thus, we have <u>proven</u>, in four steps, that no breeze in (1,1) means there can be no pit in either (1,2) or (2,1)

This symbolic inference can be a lot more efficient than naive enumeration of models – *if* we can apply rules in the "good" order!

The Resolution rule

An inference algorithm is guaranteed to be complete if it uses the *resolution rule*



Unit resolution

Full resolution

A clause = a disjunction (\lor) of literals A literal = a positive or a negative symbol

The Resolution rule

An inference algorithm is guaranteed to be complete if it uses the *resolution rule*







Note: The resulting clause should only contain one copy of each literal.

Resolution truth table $A \lor B \neg B \lor C A \lor C$ $\neg B$ С В Α \cap 1

 $((A \lor B) \land (\neg B \lor C)) \Rightarrow (A \lor C)$

Resolution truth table $A \lor B \neg B \lor C A \lor C$ $\neg B$ С В Α 1 A 1 \mathbf{O} 1 1 1

 $((A \lor B) \land (\neg B \lor C)) \Rightarrow (A \lor C)$

Proof for the resolution rule

Conjunctive normal form (CNF)

- Every sentence of propositional logic is equivalent to a conjunction of clauses
 - a clause is a finite disjunction of literals
 - a literal is an atomic formula or its negation
- Sentences expressed in this way are in *conjunctive* normal form – CNF
 - there is also DNF (disjunctive normal form), i.e. a disjunction of conjunctive clauses
- A sentence with exactly k literals per clause is said to be in k-CNF

This is good, it means we can get far with the resolution inference rule.

Wumpus CNF example

$B_{11} \Leftrightarrow (P_{12} \lor P_{21})$	Rule of the
	yame
$B_{11} \Rightarrow (P_{12} \lor P_{21}) \land (P_{12} \lor P_{21}) \Rightarrow B_{11}$	Biconditional elimination
$(\neg B_{11} \lor (P_{12} \lor P_{21})) \land (\neg (P_{12} \lor P_{21}) \lor B_{11})$	Implication elimination
$(\neg B_{11} \lor P_{12} \lor P_{21}) \land ((\neg P_{12} \land \neg P_{21}) \lor B_{11})$	"De Morgan"
$(\neg B_{11} \lor P_{12} \lor P_{21}) \land ((\neg P_{12} \lor B_{11}) \land (B_{11} \lor \neg P_{21}))$	Distributivity
$(\neg B_{11} \lor P_{12} \lor P_{21}) \land (\neg P_{12} \lor B_{11}) \land (B_{11} \lor \neg P_{21})$	Voilá – CNF
$(A \Leftrightarrow B) \equiv ((A \Rightarrow B) \land (B \Rightarrow A)) \qquad \neg(A \lor B) \equiv (\neg A \land A)$	∧ ¬B)
$(A\RightarrowB)\equiv(\negA\lorB)\qquad\qquad(A\lor(B\landC))\equiv$	((A ∨ B) ∧ (A ∨ C))

Wumpus CNF example

$\mathsf{B}_{11} \Leftrightarrow (\mathsf{P}_{12} \lor \mathsf{P}_{21})$

 $\mathsf{B}_{11} \Rightarrow (\mathsf{P}_{12} \lor \mathsf{P}_{21}) \land (\mathsf{P}_{12} \lor \mathsf{P}_{21}) \Rightarrow \mathsf{B}_{11}$

 $(\neg B_{11} \lor (P_{12} \lor P_{21})) \land (\neg (P_{12} \lor P_{21}) \lor B_{11})$

Rule of the game

Biconditional elimination Implication elimination

 $(\neg \mathsf{B}_{11} \lor \mathsf{P}_{12} \lor \mathsf{P}_{21}) \land ((\neg \mathsf{P}_{12} \land \neg \mathsf{P}_{21}) \lor \mathsf{B}_{11})$

"De Morgan"

 $(\neg B_{11} \lor P_{12} \lor P_{21}) \land ((\neg P_{12} \lor B_{11}) \land (B_{11} \lor \neg P_{21}))$ Distributivity

 $(\neg \mathsf{B}_{11} \lor \mathsf{P}_{12} \lor \mathsf{P}_{21}) \land (\neg \mathsf{P}_{12} \lor \mathsf{B}_{11}) \land (\mathsf{B}_{11} \lor \neg \mathsf{P}_{21}) \quad \text{Voilá - CNF}$

The **resolution refutation** algorithm

Proves by the principle of contradiction: Shows that $KB \models \alpha$ by proving that $(KB \land \neg \alpha)$ is unsatisfiable.

- Convert (KB $\land \neg \alpha$) to CNF
- Apply the resolution inference rule repeatedly to the resulting clauses
- Continue until: (a) No more clauses can be added, KB $\nvDash \alpha$ (b) The empty clause (\varnothing) is produced, KB $\vDash \alpha$



KB = The set of models that agrees with the knowledge base (the observed facts) [The KB is true in these models]

 α_{l} = The set of models that agrees with conclusion α_{l} [conclusion α_{l} is true in these models]

Wumpus resolution example

 $\mathsf{KB} \land \neg \alpha = (\neg \mathsf{B}_{11} \lor \mathsf{P}_{12} \lor \mathsf{P}_{21}) \land (\neg \mathsf{P}_{12} \lor \mathsf{B}_{11}) \land (\mathsf{B}_{11} \lor \neg \mathsf{P}_{21}) \land \neg \mathsf{B}_{11} \land \mathsf{P}_{21}$

Wumpus resolution example



Not satisfied, we conclude that $KB \vDash \alpha$

Completeness of resolution

S = Set of clauses

RC(S) = Resolution closure of SRC(S) = Set of all clauses that can be derived from S by the resolution inference rule.

RC(S) has finite cardinality (finite number of symbols P₁, P₂, ..., P_k) \Rightarrow Resolution refutation must terminate.

Completeness of resolution

The ground resolution theorem

If a set *S* is unsatisfiable, then RC(S) contains the empty clause \emptyset .

Left without proof.

Your knowledge base (KB) is this:

$B \Rightarrow C$ $B \land C \Rightarrow A$

Prove, using the resolution refutation algorithm, that A is True

Your knowledge base (KB) is this:

KB in CNF



Prove, using the resolution refutation algorithm, that A is True

Hypothesis: A is True $\alpha = A$

 $\mathsf{KB} \land \neg \alpha$

 \boldsymbol{B}



 $-\mathbf{A}$



 $\mathsf{KB} \land \neg \alpha$

 \boldsymbol{B}

 $-B \sqrt{C}$

 $-(B \land C) \lor A = -B \lor -C \lor A$



6



 $\mathsf{KB} \land \neg \alpha$

 \boldsymbol{B}

 $-B \lor C$

 $-(B \land C) \lor A \equiv -B \lor -C \lor A$





-B





(KB $\land \neg \alpha$) is unsatisfiable so α is True.



We could have illustrated the resolution refutation steps with a graph...



Problem with resolution refutation

- It may expand all nodes (all statements)
 exponential in both space and time
- Is there not a more efficient way
 - to only expand those nodes (statements) that affect our query?

Horn clauses and forward- backward chaining

- Restricted set of clauses: Horn clauses
- disjunction of literals where at most one is positive, e.g.,

•
$$(\neg A_1 \lor \neg A_2 \lor \cdots \lor \neg A_k \lor B)$$
 or $(\neg A_1 \lor \neg A_2 \lor \cdots \lor \neg A_k)$

• Why Horn clauses? Every Horn clause can be written as an implication, e.g.,

•
$$(\neg A_1 \lor \neg A_2 \lor \cdots \lor \neg A_k \lor B) \equiv (A_1 \land A_2 \land \cdots \land A_k) \Rightarrow B$$

•
$$(\neg A_1 \lor \neg A_2 \lor \cdots \lor \neg A_k) \equiv (A_1 \land A_2 \land \cdots \land A_k) \Rightarrow False$$

• Inference in Horn clauses can be done using *forward-backward* (F-B) chaining in *linear time*

Slide adapted from V. Pavlovic

Forward or Backward?

Inference can be run forward or backward

Forward-chaining:

 Use the current facts in the KB to trigger all possible inferences

Backward-chaining:

- Work backward from the query proposition Q
- If a rule has Q as a conclusion, see if antecedents can be found to be true
Example

KB

KB in graph form





We are going to check if Q is True

Example



KB in graph form



We are going to check if Q is True

Example of forward chaining



Example of backward chaining





















Wumpus world revisited

Knowledge base (KB) in initial position (ROG = Rule of the Game)



Wumpus world revisited

Knowledge base (KB) in initial position (ROG = Rule of the Game)

$$\begin{array}{lll} 1-16 & B_{i,j} \Leftrightarrow (\mathsf{P}_{i,j+1} \lor \mathsf{P}_{i,j-1} \lor \mathsf{P}_{i-1,j} \lor \ \mathsf{P}_{i+1,j}) & \mathsf{ROG: Pits} \\ 17-32 & S_{i,j} \Leftrightarrow (\mathsf{W}_{i,j+1} \lor \mathsf{W}_{i,j-1} \lor \mathsf{W}_{i-1,j} \lor \mathsf{W}_{i+1,j}) & \mathsf{ROG: Wumpus' odor} \\ 33 & (\mathsf{W}_{1,1} \lor \mathsf{W}_{1,2} \lor \mathsf{W}_{1,3} \lor \cdots \lor \mathsf{W}_{4,3} \lor \mathsf{W}_{4,4}) & \mathsf{ROG: } \#\mathsf{W} \ge 1 \\ 34-153 & \neg(\mathsf{W}_{i,j} \land \mathsf{W}_{k,l}) & \mathsf{ROG: } \#\mathsf{W} \le 1 \\ 154 & (\mathsf{G}_{1,1} \lor \mathsf{G}_{1,2} \lor \mathsf{G}_{1,3} \lor \cdots \lor \mathsf{G}_{4,3} \lor \mathsf{G}_{4,4}) & \mathsf{ROG: } \#\mathsf{G} \ge 1 \\ 155-274 & \neg(\mathsf{G}_{i,j} \land \mathsf{G}_{k,l}) & \mathsf{ROG: } \#\mathsf{G} \le 1 \\ 275 & (\neg\mathsf{B}_{11} \land \neg\mathsf{W}_{11} \land \neg\mathsf{G}_{11}) & \mathsf{ROG: } \mathsf{Start safe} \end{array}$$

There are 5 "on-states" for every square, {W,P,S,B,G}. A 4×4 lattice has $16 \times 5 = 80$ distinct symbols. Enumerating models means going through 2^{80} models!

The physics rules (1-32) are very unsatisfying – no generalization.



Summary

- Knowledge is in the form of sentences in a knowledge representation language
- The representation language has syntax and semantics
- Propositional logic consists of
 - proposition symbols
 - logical connectives
- Inference:
 - Model checking
 - Inference rules (e.g. resolution)
 - Horn clauses