EPISTEMIC LOGICS AND THEIR APPLICATIONS IN ARTIFICIAL INTELLIGENCE

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IJCAI 1993 Chamberry What Are Epistemic Logics?

---Logics of Knowledge

---Logics of Belief

and their extensions

Logics used to reason *about* Knowledge and Belief

What is Reasoning about Knowledge?

- Reasoning about how agents <u>use</u> knowledge
- -- how they reason with their knowledge
- -- how they reason with partial knowledge
- [Pierre doesn't know the directions to Lyon, but does know the number of the automobile club]

Does <u>not</u> include Knowledge-based systems et. al. which use knowledge (facts), but don't reason about it

Aim of Tutorial

- --- motivate the problem
- --- introduce concepts, vocabulary
- --- examine major issues
- --- point to tools, applications

Why do we want to reason about knowledge and belief?

- Inherently important ---central concern of philosophy, psychology
 - --- how do agents acquire knowledge (Theaetetus) ?
 - -- what is the relation between knowledge and belief?
 - -- do agents know all the consequences of their knowledge ?
 - -- how do you explain inconsistent beliefs ?
 - Important for applications
 - --- Al applications :
 - planning, speech act theory, CAI
 - --- Other CS applications:
 - distributed systems, security
 - --- Applications outside AI: economics

AI Applications

- Planning
- Text Understanding
- Active Perception
- Speech Acts
- Intelligent Computer Aided Instruction
- Design of Intelligent Systems
- Nonmonotonic Logic

AI Applications

Planning

- In perfect world (complete knowledge) planning can be done without reasoning about knowledge
- -- Real world ---- incomplete knowledge so planning agent must reason

does agent have enough knowledge to perform action? does other agent know enough to do action?

Knowledge Preconditions Problem for Actions and Plans

From the New York Times, Metropolitan Diary, Nov. 27, 1991

Dear Diary:

This is what happened the other day.

Richard locked himself out of his West Fourth Street apartment. The super wasn't around. Two hours later, Richard was still waiting in his lobby. Then Mary Anne, an upstairs neighbor, came home. She didn't have the keys to Richard's apartment, but she had keys to Carol's apartment next door to her. And Carol, she knew, had keys to Lydia's apartment on the floor below. And Lydia, Richard knew, had keys to his apartment.

So Mary Anne used her keys to get into Carol's apartment where she found a set of keys labeled "Lydia." Then Mary Anne and Richard went to Lydia's apartment where Richard was certain he would find the keys to his apartment. And so he did. A few minutes later he was unlocking his own door.

There's a moral here someplace, maybe about good neighbors, maybe about New York apartment dwellers. On the other hand, it could be a question. Like, didn't Lucy and Ethel have it easy?

The Moral: Intelligent agents reason about knowledge and action

AI Applications

Active Perception

[Having intelligent control for the focus of the sensor]

Using knowledge of sensor characteristics and of external world,

Predict that a given focus for the sensor will gather a desired piece of knowledge

"I can find out what I need to know

- -- for planning
- -- for physical prediction
- -- for disambiguating my perceptual interpretations by focussing my camera 10 degrees to the right"
- -- I can determine whether my wallet is in my pocket by feeling in my pocket
- -- I can determine whether a region is a mark or a shadow by looking for the object casting the shadow

AI Applications

Speech Acts (Grice)

--- modelling communication, use of language

--explains how "la neige est blanche" means snow is white [in contrast to "These clouds mean rain"] based on convention; common knowledge of what a sentence means -- explains why "Dear Sir, Mr. X has an excellent command of English and always comes to class" is a bad letter of recommendation

conversational implicature; our expectations and knowledge

Intelligent CAI (Computer Aided Instruction)

Create Automated Tutor Ideally

Would maintain a model of

- -- what the student knows
- -- how the student reasons
- -- how the student learns

Can initialize the model from

- -- generic model of students
- -- specific student data (e.g. tests)
- Can update the model from knowing
 - -- what the student has been taught
 - -- how the student responds

Can plan an effective teaching strategy

Design of Intelligent Systems

Automate the construction of a specialized AI system

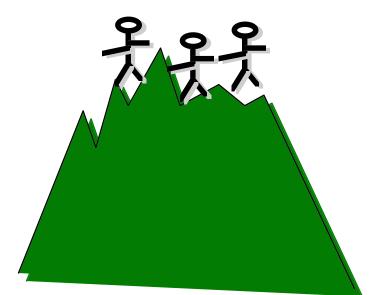
Given a specification of

- -- the kind of knowledge that the system has
- -- the evolution of the system's knowledge
- -- the proper action of the system in a given state of knowledge

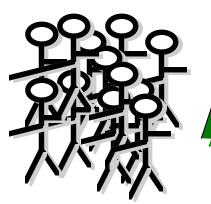
Design a knowledge-base architecture that implements this

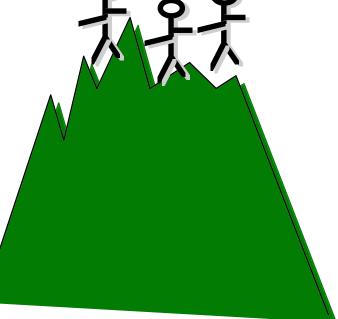
Distributed Systems: Byzantine Generals

> Byzantines must coordinate attack; otherwise, they'll be defeated



Byzantines A





Byzantines B

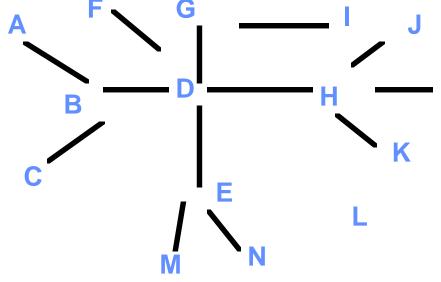
Saracens

Distributed Systems: Byzantine Generals

- 2 Byzantine armies on opposite sides of Saracen army
- If both Byz. armies attack simultaneously, they win
- If only one attacks at a time, they'll be defeated
- **Objective:** To decide on a time of simultaneous attack by sending messages back and forth
- Difficulty: The general sending the messenger can't be sure that he will get through
- **Question:** How long until they can coordinate an attack?
- Theorem: There is no protocol which enables both generals to be sure that the other will attack

Relevance: components in computer system where messages don't always get through

Application: Distributed SystemsGiven: A collection of nodes connected as a free tree.Task: To impose a directed tree structure



Each node knows the general rules:

If u borders v then either u = parent(v) or v = parent(u); The root has no parent; all other nodes have exactly 1 parent.

Protocol for u: if you border v, and you know your relation to v, communicate this to v.

Supports variety of information transmission patterns:

- Assign one node to be root: info spreads from root
- Assign n-1 nodes to root; info spread up from leaves

Applications of Epistemic Logic in General CS

Distributed Systems

Characterize a protocol for a distributed system in terms of

--What each element knows

- --What each element wants to know
- --What each element knows about what other elements know
- --What is known by the union of all elements
- --What is common knowledge throughout the system

Security

Guarantee that someone who does not know P (password) cannot find out Q (information)

Convince another agent that I know a solution to problem P without letting him know the solution

(Public key encryption, zero-knowledge proofs)

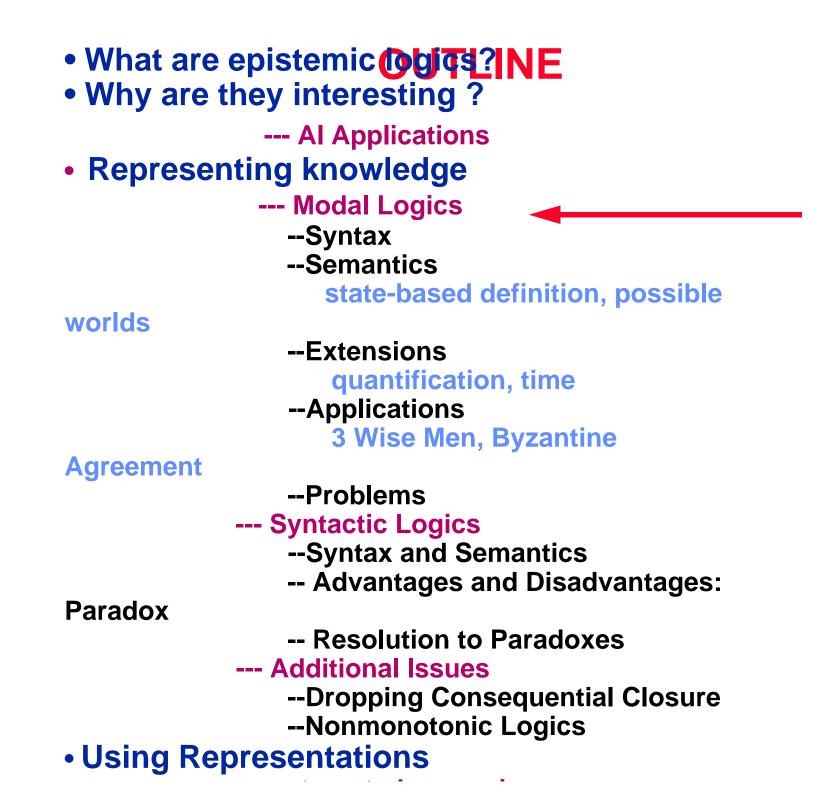
Applications of Epistemic Logic outside CS

GAME THEORY: (Partial knowledge games - Bridge, Poker)

Determine the likely action of the opponent based on his knowledge Use one's knowledge effectively without revealing its source

ECONOMICS:

Determine the expected cost and value of a particular piece of knowledge



Important issues not covered here:

- connection of knowledge to other propositional attitudes: belief, hope, fear, desire
- [nonmonotonic and] probabilistic inference
- knowledge and perception
- natural language use of "know"

Representing Knowledge

- --- Modal Logics
 - --Syntax
 - --Semantics

state-based definition, possible worlds

--Extensions

quantification, time

--Applications

3 Wise Men, Byzantine Agreement

--Problems

--- Syntactic Logics

- --Syntax and Semantics
- -- Advantages and Disadvantages: Paradox
- -- Resolution to Paradoxes

--- Additional Issues

- --Dropping Consequential Closure
- --Nonmonotonic Logics

Representing Knowledge

How do we talk about knowledge? LOGIC ---

Extending standard logic into logic of knowledge Starting point: Propositional Logic Propositions: P Q R

Connectives: ~ v & ==> <==>

Examples:

Temp = 25 Frog-Kermit v Temp = 25 Frog-Kermit ==> Green-Kermit

Note: no way to talk about knowledge

Extending Propositional Logic to Modal Logic of Knowledge

Add modal operator Know (applies to sentences)

Examples:

Know(Frog-Kermit) Know(Frog-Kermit v ~ Frog-Kermit) Temp = 25 & ~ Know(Temp = 25) Know(~ Know(Frog-Kermit)) [nested knowledge] Implicit agent; can make explicit

Know(Beth, Temp = 25) Know(Sally, Frog-Kermit ==> Green-Kermit) Know(Sally, Know(Beth, Temp = 25)) [nested knowledge] Note: difficulty in representing knowledge

Referential Opacity

Most predicates are <u>transparent;</u> you can substitute equals for equals.

John is the father of William Color-of-eyes(John, Brown) is true just in case Color-of-eyes(father(William),Brown) is true

Not true of Know

Scott is the author of Waverly --- but Know(British(Scott)) may be true and Know(British(author(Waverly)) may be false.

Also, the Morning Star is the same as the Evening Star But Know(MorningStar = MorningStar) is true of all; But Know(MorningStar = EveningStar) is not. Know is Opaque Modal Logic of Knowledge: Multiple Agents

Mutual Knowledge

2 or more agents know some fact

A and B mutually know P iff Know(A,P) & Know(B,P)

Common Knowledge

very important for distributed systems!

2 or more agents know some fact and they know that they know, and so on ...

A and B have common knowledge of P iff Know(A,P) and Know(B,P) and Know(A, Know(B, Know(A, P] Know(B, Know(A, Know(B, P]

How can we define the Know operator?

Intuitive definition of Know:

---- whatever is explicitly stated in a knowledge base

--- implicit knowledge in a propositional knowledge base

--- what can be derived

implicit knowledge definition is most accepted

We need to write down axioms to capture this concept of knowledge

[Possible] Axioms on Knowledge

- 1. Veridicality If A knows P, then P is true
- **2. Consequential Closure**
 - If A knows P_1 ... and A knows P_n and P_1 ... $P_n \models Q$ then A knows Q
- 3. Knowledge of Necessary Truths If P is necessarily true, then A knows P
- 4. Positive Introspection
 - If A knows P then A knows that A knows P
- **5. Negative Introspection**
 - If A does not know P, then he knows that he does not know P

Different subsets of these axioms form different systems of modal logic

1. Veridicality If A knows P, then P is true

- **2. Consequential Closure**
 - If A knows P_1 ... and A knows P_n and P_1 ... $P_n \models Q$ then A knows Q
- 3. Knowledge of Necessary Truths If P is necessarily true, then A knows P

== T, a simple modal logic of knowledge

1. Veridicality If A knows P, then P is true

2. Consequential Closure

If A knows P₁ ... and A knows P_n and P₁ ... P_n |- Q then A knows Q

3. Knowledge of Necessary Truths If P is necessarily true, then A knows P

4. Positive Introspection If A knows P then A knows that A knows P

== S4, popular modal logic of knowledge

1. Veridicality

If A knows P, then P is true

2. Consequential Closure

If A knows $P_1 \dots$ and A knows P_n and $P_1 \dots P_n \mid -Q$ then A knows Q

3. Knowledge of Necessary Truths

If P is necessarily true, then A knows P

4. Positive Introspection

If A knows P then A knows that A knows P

5. Negative Introspection

If A does not know P, then he knows that he does not know P

== S5, modal logic for ideal knowledge

2. Consequential Closure

If A believes P₁ ... and A believes P_n and P₁ ... P_n |- Q then A believes Q

3. Knowledge of Necessary Truths

If P is necessarily true, then A believes P

4. Positive Introspection

If A believes P then A believes that A believes P

5. Negative Introspection
 If A does not believe P, then he believes
 that he does not believe P

 == weak S5, better suited for belief

Note: an agent's beliefs are not necessarily true

What does <u>Know</u> mean? When can we say that a sentence like

Know(John, temp = 25)

is true?

Several characterizations:

state-basedpossible-worlds semantics

State-based definition of knowledge

[good for a machine]

Consider a device D that can be in one of a collection of states. We say: D knows P

if the state of D is S and whenever D is in state S, P is true

Examples:

- **D:** a mercury thermometer **S:** the level of mercury
- D knows temperature is 25C because mercury points to 25C and mercury only points to 25C when it is 25C
- D knows temperature is not 0C because because temp. is never 0 when mercury is at 25C

State-based definition of knowledge, cont.

Also

D knows that the temperature > 5C and < 100C
D knows that the temperature in degrees C is a square number
D knows that any planar map can be colored with 4 colors
D kows that either the sun is shining or it is not shining

And ...

D does not know that the sun is shining because sometimes mercury points to 25 when it rains

State-based definition of knowledge

Other examples:

- **D**: an inventory database
- **S** : relation instance in the database

D knows that there are 5 widgets on the shelf

D knows that there are fewer than 8 widgets on the shelf

State-based definition of knowledge Combined knowledge

Devices D1 and D2 together know P if the state of D1 is S1 the state of D2 is S2 whenever the state of D1 is S1 and the state of D2 is S2, P is true

Example:

D1: an inside thermometer D2: an outside thermometer

D1 and D2 together know it's colder outside than inside because D1's mercury points to 22 D2's mercury points to 10 whenever D1's mercury is at 22 and D2's mercury is at 10, it's colder outside than inside

Combined Knowledge: Example

D1: device in store that keeps track of purchases D2: device in store that keeps track of incoming orders

D1 and D2 together knows that there are 10,000 items in the store because D1 indicates 30,000 items have been purchased D2 indicates 40,000 items have come in

State-based definition of knowledge Common Knowledge

Devices D1 and D2 have common knowledge of P iff state of D1 is an element of some set of states SS1 state of D2 is an element of some set of states SS2 whenever state of D1 is in SS1,

P is true and the state of D2 is in SS2 whenever state of D2 is in SS2,

P is true and the state of D1 is in SS1

Example:

D1: a digital clock that displays hour and minute

D2: a digital clock that displays only the hour

D1 and D2 have common knowledge that time is between 12:00 and 1:00 iff

SS1 = the set of all displays 12:xx on D1

SS2 = the set of D2 displaying 12:00

Applications of the State-based Definition: Distributed Systems

Let P = "Printer 0 is free"

Machine M1 knows p iff the internal state of M1 is attained only when P is true.

M1 communicates P to M2 through message C if M1 only sends C to M2 when M1 knows P and Whenever M2 receives C from M1, it enters a state where M2 knows P

M2 knows that M1 knows P if: whenever M2 is in its current internal state, internal state of M1 is one only attained when P is true

Protocol: rules of the form --

if Mx knows Pk, Mx communicates Cxyk to My

Verify concepts of protocols such as: if P true, then in 5 cycles, every machine will know that P is true

State-based definition of knowledge What properties hold?

Important result:

Get: veridicality

consequential closure

necessary truths

positive introspection

negative introspection

That is, we get the modal logic S5 (perfect knowledge)

Problems with State-based definition of knowledge

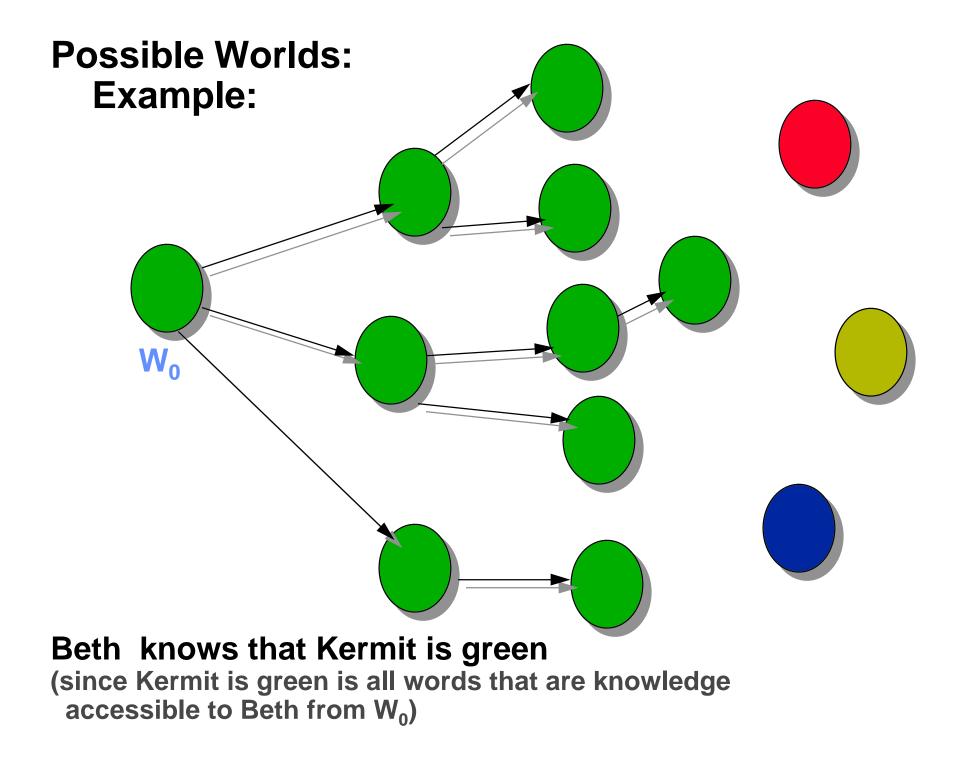
- --- fine for machines; unintuitive definition for intelligent agents
- --- consequential closure: all agents are perfect reasoners
- --- necessary truths: all agents know all axioms (universal facts)
- --- negative introspection: agents never have false beliefs
 - built into the semantics

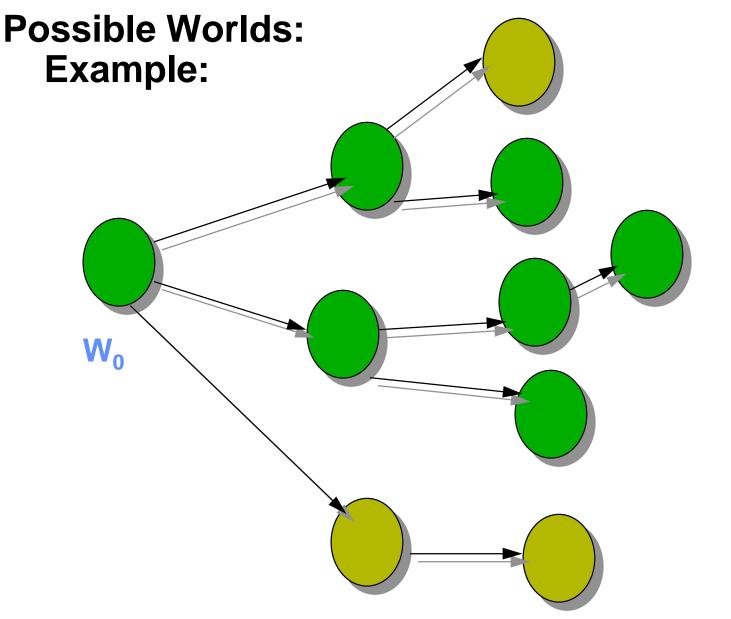
Possible Worlds Definition of Knowledge Kripke-Hintikka

- idea: A knows P iff P is true in all worlds that are knowledge-accessible for A
- W_1 is knowledge-accessible from W_0 for A iff W_1 is consistent with everthing A knows in W_0 ; that is, for all A knows in W_0 , he might as well be in W_1
- Beth knows Kermit is green if he is green in all worlds that are knowledge-accessible to Beth

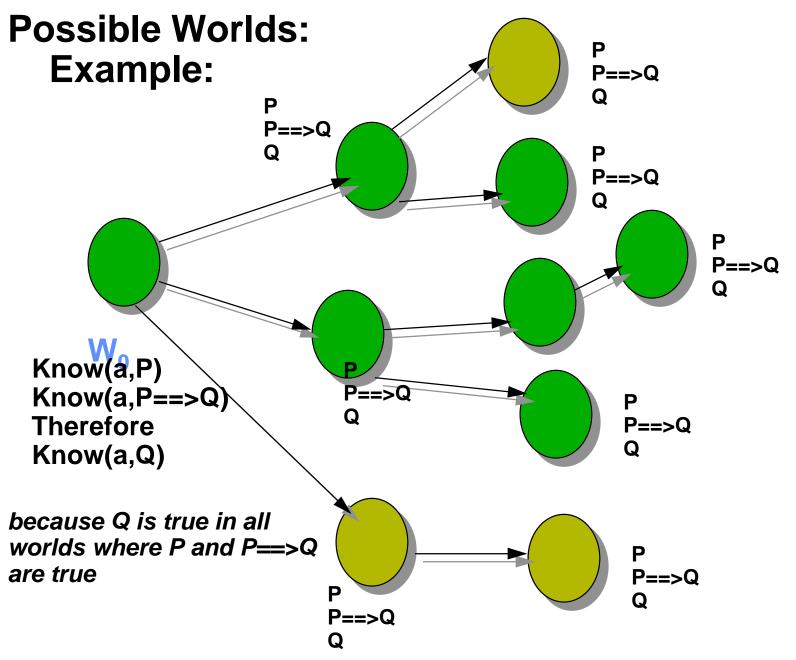
On the other hand, if in some knowledge-accessible world, Kermit is yellow, Beth doesn't know that Kermit is green

True(W_0 , Know(A,P)) iff forall W_1 K(a, W_0 , W_1) ==> True(W_1 ,P)

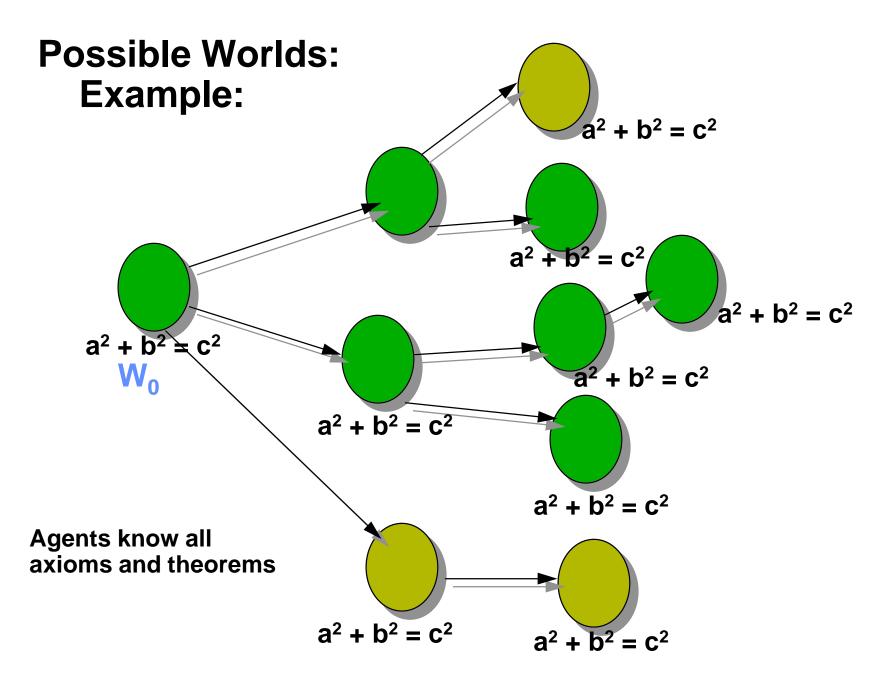




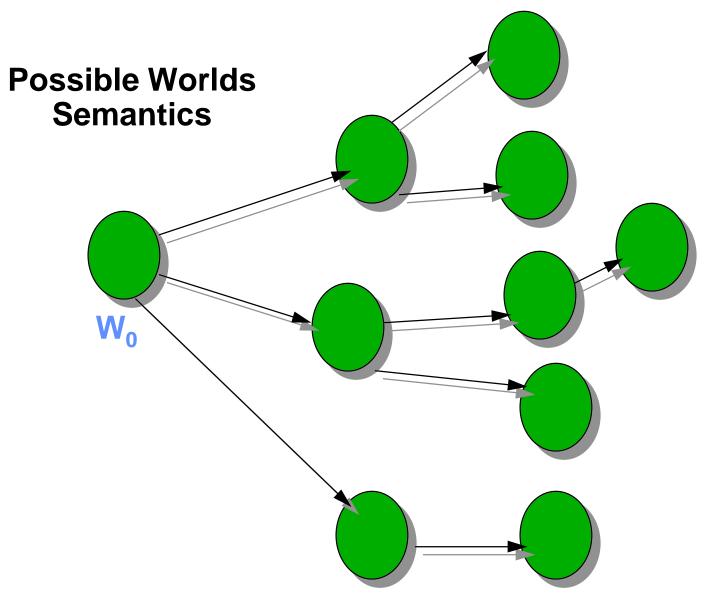
Beth doesn't know whether Kermit is green (since Kermit is green in some worlds; yellow in others)



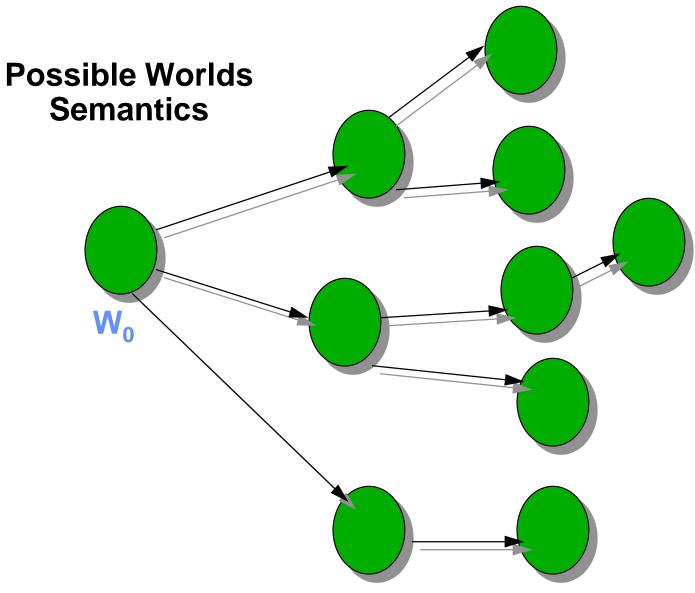
Agents always know the consequences of their knowledge !!



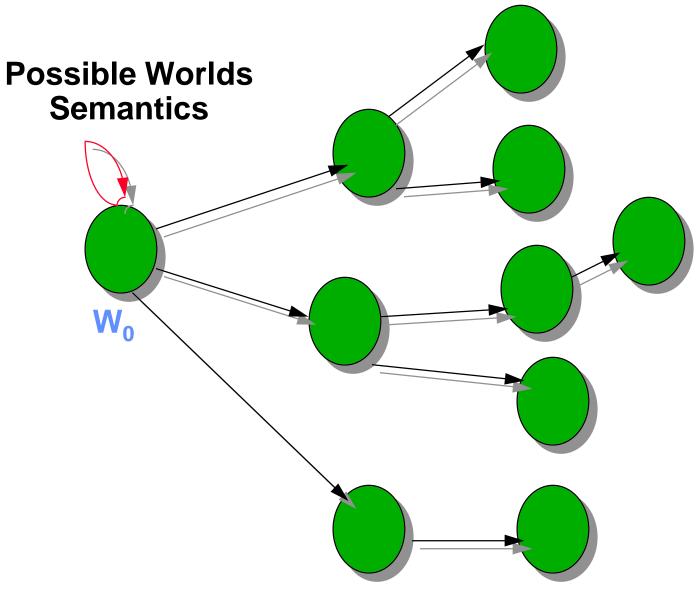
Since Pythagorean theorem is true in all worlds, Beth knows it (even if she's 5 years old)



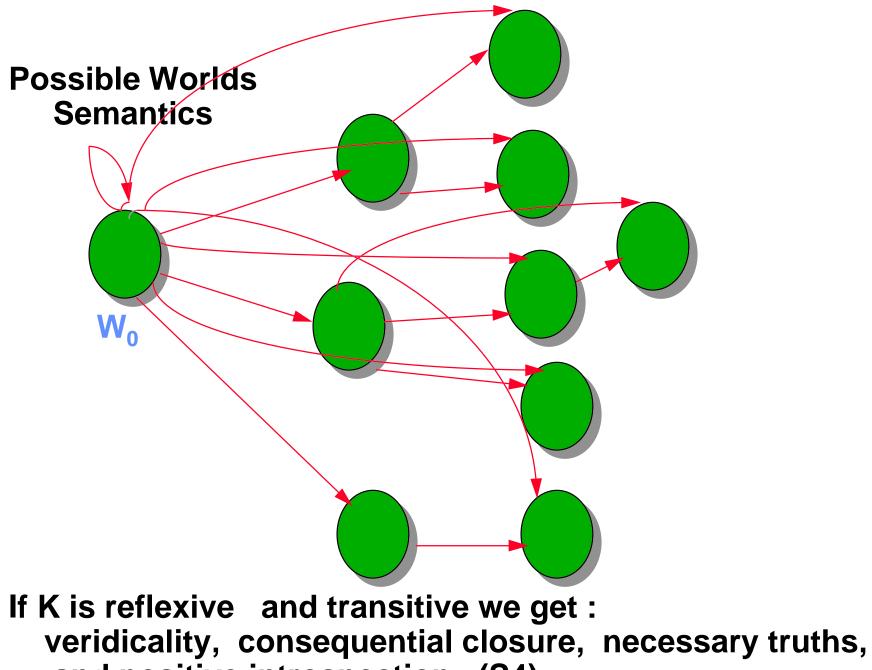
Note: Different modal logics (subsets of axioms) correspond to properties of the knowledge accessibility relation



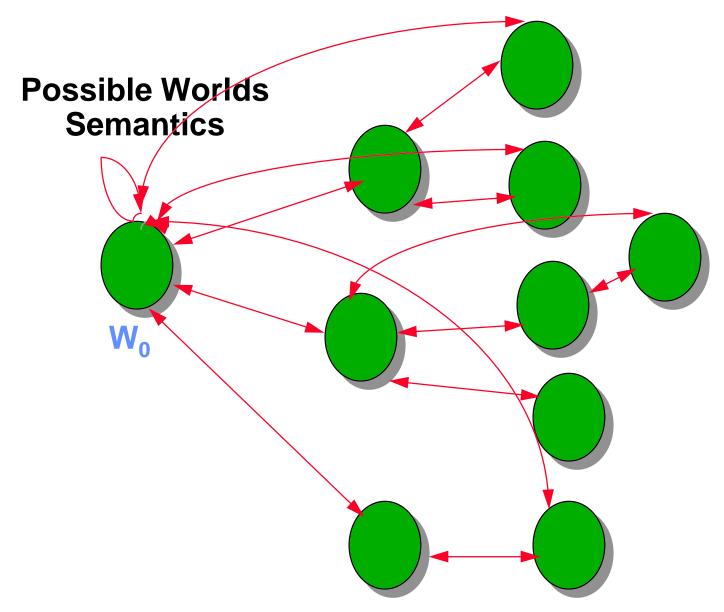
No restrictions: Just get consequential closure and necessary truths (weak T)



If K is reflexive, we get : veridicality, consequential closure, necessary truths (T)



and positive introspection, (S4)



If K is reflexive, symmetric, and transitive we get : veridicality, consequential closure, necessary truths, positive introspection, and negative introspection (S5)

Problems with Possible Worlds definition of Knowledge

- --- Is "knowledge-accessible" any more intuitive than knowledge?
- ---- Consequential closure: all agents are perfect reasoners
- --- Necessitation: all agents know all axioms (universal facts)

Built into the semantics; can't take these out

[restricts us to a very small class of modal logics]

Some Directions for Extensions:

- -- Adding quantification Quantifying into epistemic contexts
- -- Adding the concept of time
- -- Dropping "perfect reasoner" assumption (consequential closure)

Adding Quantification

Before base logic was propositional logic

Now base logic is predicate logic

Quantifying into Epistemic Contexts

Consider the following sentence:

John knows someone is blackmailing him

2 possible readings:

- 1. Know(John, ∃x Blackmailer(x,John)) (de dicto)
- 2. ∃x (Know(John, Blackmailer(x,John))) (de re)

Second reading more fit if John knows who is blackmailing him

Note: 2. implies 1., but 1. does not imply 2

Quantifying into Epistemic Contexts

• When can we say

∃x(Know(John, (x)) [∃x (Know(John, Blackmailer(x)))]

We can deduce it from

Know(John, Blackmailer(Mr.Thorpe, John))

but not necessarily from

Know(John, Blackmailer(Murderer(Sam), John)))

How much knowledge does John have to have ? constant? name? rigid designator?

Adding the concept of time

Till now: no concept of time

Know(Beth, Green(Kermit))

But -- facts of knowledge refer to *time* People know different things at different times

Know(Beth, President(USA, Clinton))

is meaningless When is Clinton President? When does Beth know this?

Need to add concept of time

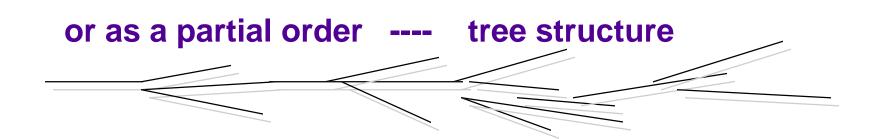
Adding time to epistemic logics

Many methods: Adding extra temporal argument

Know(Beth, President(USA, Clinton, 1993)), 1993)

Know(Beth, President(USA, Bush, 1992)), 1993)

can imagine time as a total order --- time line



Adding Time How is knowledge affected by time? Perfect memory Know(A,P,S1) & S1 < S2 ==> Know(A,P,S2)

Belief is also affected by time:

Changing your mind --

if A believes P, he believes he'll always believe P Bel(A,P,S1) ==> Bel(A, forall S1 S1 < S2 ==> Bel(A,P,S2), S1)

Future beliefs --

if A believes he'll believe P in future, he believes it now Bel(A, ∃ S2 S1 < S2 and Bel(A,P,S2), S1) ==> Bel(A, P, S1) **Adding Time**

Predicting the future: reasoning about the effects of an action

How to say: Susan knows that if she moves block A to block B, block A will be on top of block B

Need to: integrate logic of knowledge with logic of action **Situation Calculus**

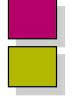
situation = instant of time

situations partially ordered by < : branching model of time

Actions = functions on situations

E.g., Puton(A, B) maps situations in which blocks A and B are clear to situations in which block A is on top of block B





True-in(Result(put-on(A,B),S), on(A,B))

Know(Sam, True-in(Result(put-on(A,B),S), on(A,B)))

Application: Three Wise Men Problem

In other guises: Dirty Children Problem Cheating Husbands Problem

Idea: Three wise men are told that at least one has a black dot on his forehead. Everyone can see if others have black dots, but no-one can see his own forehead.

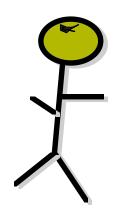
Assume that we start at t = 0.

All are perfect reasoners.

Any round of reasoning takes one unit. If all of the wise men have black dots, how long will

it take them to realize? If 2 have dots? if 1 does?

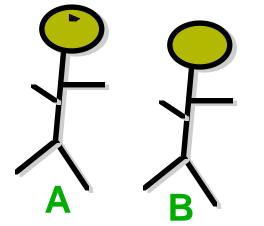
BASE CASE: 1 Wise Man



This is trivial; he knows he has a dot on his forehead so he says it right away, at t = 0.

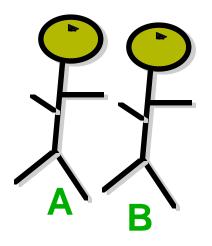
Now suppose there are 2 men

Case I: 1 man has a black dot



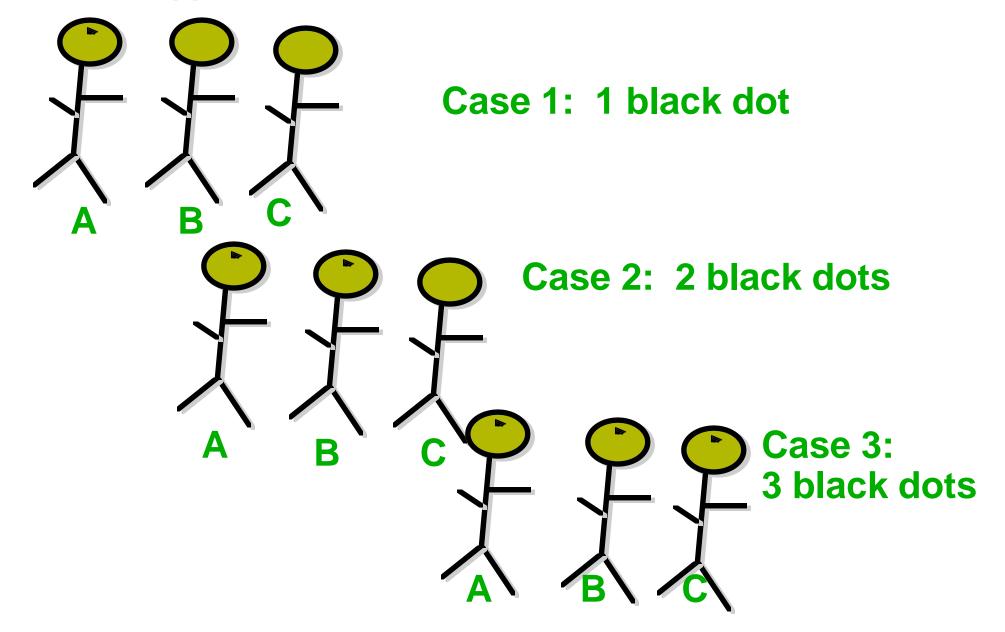
At time t = 0, A sees that B doesn't have a dot. Since he knows that one of them has a dot, he figures that he does. So at t = 1, A says: I have a black dot. (B can't figure anything out.)

Case 2: both men At t have dots Thu

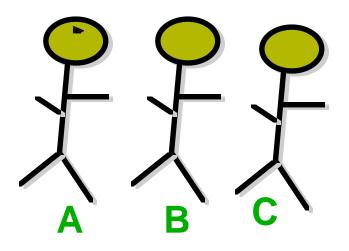


At time t = 0, A sees that B has a dot. Thus, he doesn't know if he does or not. But at time t = 1, B is silent (he doesn't know if it's case 1 or case 2). So A knows that this *can't* be the same as case1; thus he must also have a dot. So he speaks out at time t = 2. B, doing the same reasoning, also speaks at t = 2.

Now suppose there are 3 men



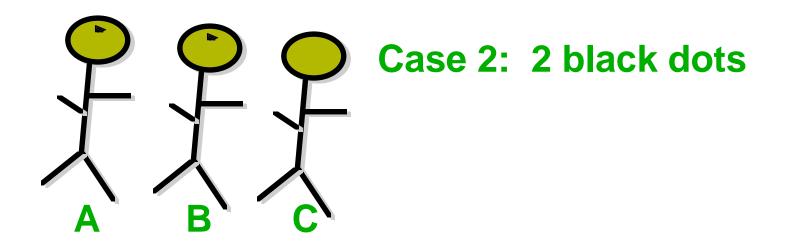
Case of 3 men



Case 1: 1 black dot

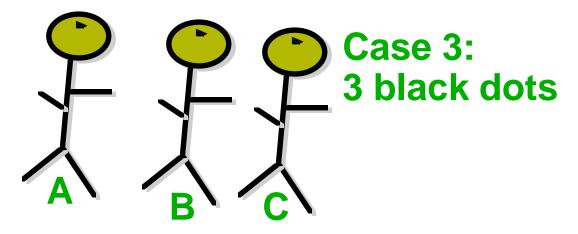
At time t = 0, B and C each see one person with a dot. So they may have dots on their forehead; they don't know. But A doesn't see anyone with a dot on his forehead, so he knows he must have a dot on his forehead. So, at time t = 1, he speaks.

Case of 3 men



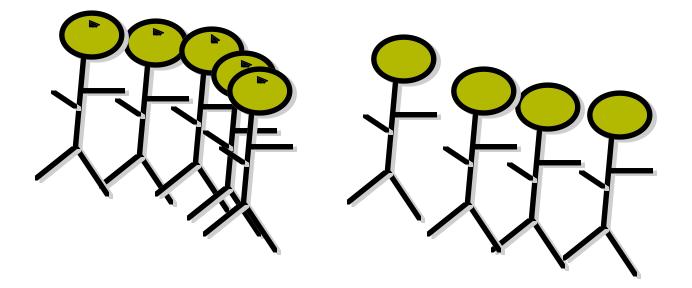
At time t = 0, everybody sees at least one person with a dot, so they don't know if they have dots. A and B each see 1 person with a dot, so they know: either there is 1 person with a dot, or 2 people. At t = 1, no-one speaks. So A [resp. B] knows it can't be that only B [A] has a dot. Because if that were the case, at time 1, B [A] would have spoken. Thus, there must be 2 people with dots -- i.e., A [B] has a dot too. At t = 2, A [and B!] speak.

Case of 3 men



At t = 0, each of A, B, and C see that two other people have dots. So, A [resp. B, C] reasons as follows: Either B and C have dots and I don't, or we all have dots. Now, if it were the case that I did not have a dot, this would reduce to case 2, and at time t = 2, B and C would speak. When t = 2 passes, and B and C do not speak, A realizes that it is not case 2; that all three have dots. B and C, reasoning similarly, come to the same conclusion. Thus at t = 3, all speak.

N - Wise Man Problem



Assume N wise men. K have black dots on forehead. Assuming - common knowledge of at least one black dot all perfect reasoners each round of reasoning takes 1 unit Theorem: K men will speak at t = K The crucial concepts: common knowledge consequential closure

Three Wise Men --- Formulation in Logic

Language:

black(x) - X has a black dot on his forehead
speak(x,t) - X states the color on time T
t + 1 - successor of time T
0 - starting time
know(x,p,t) - X knows P at time T
know-whether(x,p,t) - X knows at T whether P holds

Axioms:

W1. know-whether(x,p,t) <==> [know(x,p,t) v ~know(x,p,t)] (definition of know-whether: X knows whether P if he either knows P or he knows not P)

W2. speak(x,t) <==> know-whether(x,black(x),t)

(a wise man declares the color on his head iff he knows what it is)

Wise Men -- Logical formulation, cont.

W3. x < y == know-whether(x,black(y),t)(The wise men can see the color on everyone else's head) W4. know-color(x,t) ==> speak(x,t) (The wise men speak as soon as they figure it out) W5. know-whether(y,speak(x,t),t+1) (Each wise man knows what has been spoken) W6. know(x,p,t) ==> know(x,p,t+1) (The wise men do not forget what they know) W7. know(x,black(w1) v black(w2) v black(w3), t) (The wise men know that at least one of them has a black dot) W8. if p is an instance of W1. -- W.8, then know(x,p,t)

Inference for 3 Wise Man Problem:

Lemma: If P is a theorem (can be inferred from 1 - 5, W.1 -- W.8, then know(x,p,t) Proof: induction on length of inference (2,3, W.8)

Lemma 1.A ~black(w2) & ~black(w3) ==> speak(w2,0) Proof: From W.7, w2 knows that either w1, w2, or w3 has a black dot. From W.3 and 1, w1 knows that neither w2 nor w3 has a black dot. From 2 and 3, s2 knows that w1 has a black dot. From W.2, w1 will speak.

Analogously Lemma 1.B: ~black(w1) & ~black(w3) ==> speak(w2,0) Lemma 1.C: ~black(w1) & ~black(w2) ==> speak(w3,0)

Inference for 3 wise men, cont.

```
Lemma 2.A:

~black(w3) ==> ∃ x speak(x,0) v speak(s1,1)

Proof:

From Lemma 1.A, if ~ black(w2) as well, then speak(w1,0);

From Lemma 1.B, if ~black(w1) as well, then speak(w2,0);

Suppose, then, that black(w1) & black(w2) & ~speak(w2,0).

From W.3, know(w1,black(w2),1) & know(w1,~black(w3),1)/

From W.5, know(w1, ~speak(w2,0),1).

By the lemma of necessitation, know(w1, Lemma 1.B, 1).

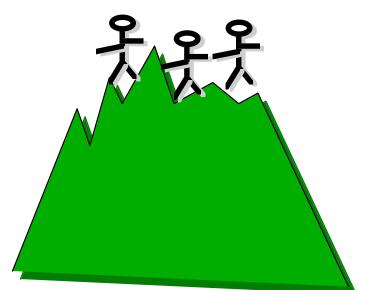
Using the contrapositive of Lemma 1.B and 2,

know(w1,black(w1),1).
```

And so on

Application: Byzantine Problem

Byzantines must coordinate attack; otherwise, they'll be defeated



Byzantines A



Byzantines B

Saracens

Byzantine Agreement:

- t = 0: A sends B message: Attack at 6:00 AM (= M)
- t = 1: B sends A message: received message
- t = 2: A now knows that B received message, but B doesn't know that A knows A sends message to B that A received B's message
- t = 3: B now knows that A knows that B knows M, but A doesn't know that B now knows that A knows that B knows that M
- t = 2n: A sends message to B
 - A knows that B knows that ... that A knows (2n times) (but not 2n+1 times)

Never reach common knowledge.

Thus, can't coordinate attack

Application: Knowledge Preconditions for Actions and Plans

Interrelationship between Knowledge and Action

--- How does knowledge affect action ? --- How do actions affect knowledge?

Focus of research:

- --- agent wants to do an action
- --- he doesn't know all that he needs to know
- --- how can he get the action done anyway?

Knowledge Preconditions for Actions and Plans

Studied by McCarthy and Hayes; Moore presented first concise solution.

Moore's theory based on

--- possible worlds theory of knowledge--- situation calculus

situations = possible worlds

Moore: Knowledge Preconditions Problem (single agent case)

Basic idea:

You know how to do an action Dial(no(Suzanne)) iff you know executable procedure [assumption: all agents know basic procedures] iff you know what the parameters of the actions are

So you know how do perform Dial(no(Suzanne)) if you know what no(Suzanne) is

How do you know what the parameters of an action are? You know what something is iff you know of a rigid designator for that object

rigid designator = something that stays the same in all possible worlds (name, number, constant)

Know how to do Dial(no(Suzanne)) if know some number equal to no(Suzanne)

Moore: Knowledge Preconditions for Plans (single-agent)

Basic idea:

Knowledge Preconditions for Plans reduce to Knowledge Preconditions for Actions

For example:

You know how to do sequence(act1, act2) if you know how to do act1 and as a result of doing act1 you know how to do act2

Consider the plan

sequence(look_up_no(Suzanne), dial(no(Suzanne)))
You can perform the plan if you can do lookupno(Suzanne)
and you can then do dial(no(Suzanne))

Extension to Moore (Morgenstern) : Knowledge Preconditions for Multi-agent Plans

Example:

Pierre wants to drive to Lyon He doesn't know the directions He does know the number of the automobile club How can he plan to drive to Lyon?

We want to show that Pierre can execute the following plan:

What's needed:

ability to reason about one's own ability to do actions, ability to predict other people's actions

Knowledge Preconditions for Multi-agent Plans In general:

- --- need to know that you'll <u>be able</u> to do your part of the plan when it comes up
- --- need to predict that other agents <u>will</u> do their parts of the plan at the proper time

How to predict other agents' actions:

- --- consider interactions between knowledge, goals, and actions (BDI)
- --- agents typically act in their own interests
- --- will often accede to a request if there are no conflicting goals

Note: Importance of communication (establishing goals, relaying information)

Knowledge Preconditions for Multi-agent Plans

Consider Pierre's plan:

Pierre can execute the plan if :

- --- he knows the number of the auto club
- --- he knows how to ask for directions
- --- he can predict that once asked, the officer of the auto club will give him directions
- --- he knows that once the officer gives him directions, he will know how to get to Lyon

works because officer is cooperative and knowledgeable, and knows how to give directions

Application: Speech Acts (Grice)

"Dear Sir, Mr. X has an excellent command of English and always comes to class"

Why will this doom Mr. X?

Cooperative Principle Plus Maxims of Conversation:

Say as much as is needed, no more, no less

Since there is common knowledge of these maxims, and Mr. X's teacher must know more about him, his failure to say more must mean that there's nothing more that is good to say.

Based on common knowledge of convention, of maxims of conversation

Beyond Modal Logic

Disadvantages of Modal Logic:

- Inexpressive Can't quantify over propositions Can't say, e.g. John knows something that Bill doesn't know
- 2. Non-intuitive semantics
 - --- state-based
 - --- possible worlds
- Undesirable consequences of semantics
 --- necessary truths
 --- consequential closure

Representing Knowledge

--- Modal Logics

--Syntax

--Semantics

state-based definition, possible worlds

--Extensions

quantification, time

--Applications

3 Wise Men, Byzantine Agreement

--Problems

--- Syntactic Logics

--Syntax and Semantics

- -- Advantages and Disadvantages: Paradox
- -- Resolution to Paradoxes

--- Additional Issues

--Dropping Consequential Closure

--Nonmonotonic Logics

Alternative to modal logic: Syntactic Logic

standard predicate logic can't say: Know(John, Frog(Kermit)) since Know is a predicate and can't range over the sentence Frog(Kermit)

introduce an invertible map from sentences (wffs) to terms

[Godel mapping maps each wff onto an integer]

denote range of mapping function with quotation marks

```
Know(John, "Frog(Kermit)")
```

Tarskian semantics

Features of Syntactic Logics

Advantages:

- 1. Expressivity ∃x (Know(Bill,x) & ~ Know(John,x)) forall x (Concerns(x,Radiology) ==> Know(Helene,x))
- 2. No need for necessitation, consequential closure

Disadvantages:

- 1. Messiness quasi-quotation
- 2. Paradox

Tarskian Semantics considered advantage by some, disadvantage by others

Messiness of Syntactic Logic

```
Saying simple things gets ugly.
Can't just say [principle of positive introspection]:
```

forall a,x
Know(a,x) ==> Know(a,"Know(a,x)")

This would imply that Know(John,"Frog(Kermit)") ==> Know(John, "Know(a,x)") which is not what we want and meaningless, too!

Need quasi-quotes, which allow us to substitute value
of quoted string:
forall a,x
 Know(a,x) ==> Know(a,"Know(@a,!p!)")

Paradox

akin to Liar Paradox ---Everything I say is a lie P iff ~True("!P!")

Knower Paradox P iff Know(a,"~!P!") P is true iff a knows that it is false

Comes from unrestricted use of quotation Arises in many reasonable languages

Surprise Test Paradox: You'll have a test someday next week, but you won't know which

Pravda: Everything the Times says is a lie New York Times: The Pravda sometimes lies

Whether or not these sentences are paradoxical depends on empirical facts about the world e.g., has NY Times said one true fact? and not only on structure of sentences

Resolutions to Paradox

--- reduce expressivity (Tarski, Konolige) no self-reflexive sentences

--- Three-valued logic: true, false, neither (Kripke, Gupta, Herzberger, Barwise, Morgenstern)

--- Different semantics for Know, True (Perlis)

No free lunch: drawbacks for each

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Dropping Consequential Closure Until now, agents have been assumed to be perfect reasoners: Know(P) & Know(P ==> Q) ==> Know(Q) [consequential closure]

Clearly false:

- --- agents make mistakes
- --- if true, all agents should know whether Fermat's last theorem Riemann's Conjecture is true, but no-one does
- --- agents have inconsistent beliefs but don't believe everything

---doesn't take into account time, resources, focus, etc.

Three types of incompleteness (Konolige)

--- resource incompleteness (running out of time to take a test)

--- fundamental logical incompleteness (not knowing how to do integrals)

---- relevance incompleteness (not knowing which facts to include)

Building a System without Consequential Closure

Issues:

- how can we drop consequential closure
- what can we replace it with

Dropping Consequential Closure

Difficult because it is built into possible worlds semantics, state-based definition of knowledge part of every standard modal logic

Ways to proceed:

- 1. Drop modal logic go to syntactic logic (Konolige, Haas, Elgot-Drapkin)
- 2. Make a distinction between explicit and implicit knowledge Implicit knowledge = standard concept of knowledge Introduce concept of awareness Explicit knowledge = awareness plus implicit knowledge Consequential closure for implicit knowledge only (Levesque, Halpern and Fagin)

Replacing Consequential Closure <u>Problem</u>: If agents don't do perfect reasoning, just what do they do?

Proposal: Limit reasoning rules in some way

--- restricted set of inference rules e.g., math student might not know integrals robot might not know path-finding algorithm --- restricted resources --- specifically time, number of steps --- clear that agents only have limited time to reason (Elgot-Drapkin, Kraus, Nirkhe, and Perlis) ---- "need-to-know" --- idea is that our reasoning is goal-oriented --- plan to reason (Haas)

Problems with Alternatives to Consequential Closure:

--- restricted rules seem arbitrary, counter-intuitive

--- can always find counterexamples

- --- limited resources, e.g., limited number of steps: what makes *n* the cutoff as opposed to *n*+1? If I know p, and q is *n* steps away, I'll know q. But then won't I know r if r is 1 step away from q?
- --- restricted reasoning rules: logicians are thoroughly familiar with rules of logic, and still aren't perfect reasoners.
- --- "need to know" agents seem to chain forward, too.

Nonmonotonic Logic

Commonsense reasoning

often draws conclusions on basis of partial information

• Birds typically fly Tweety is a bird

Counterexamples: penguins, broken wings

Tweety flies

• If I turn the key in the ignition, the car will start

Counterexamples: dead battery, bad starter

Really:

If I turn the key in the ignition, and the starter works, and the battery works, and there's gas in the car and there's no potato in the tailpipe and ... then the car will start

 If I had an older brother, I'd know it I don't know I have an older brother, so I infer that I don't have one Counterexamples: General Hospital, Bill Clinton

Such reasoning (Tweety flying, my car starting, my lack of an older brother) can't be carried out in classical logic

- Classical logic --- Drawing permanent conclusions based on complete information
- What we need --- Drawing conclusions on basis of *incomplete* information --later retract Nonmonotonic Logic

Classical Logic --- monotonic in set of assumptions the more assumptions, the more conclusions

Nonmonotonic Logic --- nonmonotonic in set of assumptions as you add assumptions, you may have to retract conclusions

Bird(Tweety)	but	Bird(Tweety), Penguin(Tweety)
Fly(Tweety)		retracts Fly(Tweety)

How can we capture nonmonotonic reasoning?

1. Default Logic (Reiter) based on <i>default</i> rules: Bird(x) : Fly(x)	new type of inference rule
Fly(x)	
2. NML (McDermott and Doy based on idea of consister Bird(x) & M(Fly(x)) ==> F	ncy rules within the logic
3. Circumscription (McCarthy restricts set of objects; in particular, abnormal of Bird(x) and ~ ab(x) ==>	objects
4. Autoepistemic Logic (Mod if x is true, I'd know x (where x is an "imp Allows inference from x	ortant" statement)

Autoepistemic Logic (Moore)

Commonsense Reasoning: based on one's beliefs --- or lack of them

e.g. how do I know I don't have an older brother? If I had an older brother, I'd know about it

Get from: P is not in my knowledge base to: I don't know P: ~Know(P)

Note: nonmonotonic

If I later find out that a parent previously married and had children, I'd retract this conclusion Nonmonotonic because <u>indexical</u> Autoepistemic Logic --- how it works based on logic of belief (L === belief)

set of formulas T that represent beliefs of reasoning agents should satisfy:

 if P₁ ... P_n in T, and P₁ ... P_n |- Q, then Q in T (consequential closure)
 if P in T, then LP in T (positive introspection)

3. if P not in T, then ~LP in T ("negative introspection")

Theories obeying 1. - 3. are stable. If a stable theory is consistent, you also get:

- 4. if LP in T, then P in T
- 5. if ~LP in T, the P not in T

Def: T is grounded in set of premises A iff every formula of T is included in the tautological consequences of A U {LP | P in T} U {~LP | P not in T}
Theorem: An AE theory T is sound w.r.t. set of premises A iff T is grounded in A

Autoepistemic Logic --- how it works

The older brother example:

```
P = "I have an older brother"
A = {P ==> LP}
```

```
By rule 1., P ==> LP in T.
Also, ~LP ==> ~P in T
```

```
Now P not in T. So by 3., ~LP in T.
So by 1., ~P in T, and by 2., L~P in T.
```

Result: You know that you do not have an older brother. You have reasoned from your own lack of knowledge

Note: stable set semantics gives us weak S5: preferred logic of belief

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Summary

Issue: Construct Theorem Prover for Epistemic Logics

Problem: Complexity

- --- very inefficient
- --- theorem prover "driven" by axioms on knowledge

Idea: Circumvent

Approaches:

- --- multiple contexts
- --- direct representation, procedural attachment
- --- inference with possible worlds
- --- vivid reasoning

Multiple Contexts

Basic Idea:

For each state of knowledge, or state of imbedded knowledge, create a separate context of "object-level" facts. Inference within a context uses "ordinary" automated reasoning. Inference from one context to the next uses special-purpose inference.

Example of Multiple Contexts

Given:

In S1, A knows p. In S1, B knows that p ==> q. In S1, A knows that B knows that p ==> q. A tells p to B during [S1,S2]

Initialize:

Context S1A (what A knows in S1): {p} Context S1B (what B knows in S1): {p==>q} Context S1AB (what A knows that B knows in S1): {p==>q}

Create corresponding contexts for time S2:

Context S2A = { ... } Context S2B = { ... } Context S2AB = { ... }

Example of multiple contexts, continued

Frame Inferences:

Context S2A : { p ... } (A still knows p) Context S2B: { p ==> q ... } (B still knows p ==> q} Context S2AB: { p ==> q ... } (A knows that B still knows that p ==> q)

Inferences associated with "tell" :

If X tells P to Y during [S1,S2] then in S2 Y knows P and X knows that Y knows P

Context S2B: { p ==> q, p} (B now knows P) Context S2AB: { p ==> q, p) (A now knows that B knows p)

Modus Ponens within context:

Context S2B: { p ==> q, p, q} (B infers Q) Context S2AB: { p ==> q, p, q} (A infers that B infers Q} **Implementation Remark:**

Since different contexts are likely to share a lot of knowledge, inference will be more efficient if facts are labelled by context, as in CONNIVER and ATMS, rather than copying the whole knowledge base. Limitations, Issues:

--- Limited expressivity: Difficult to express

A knows p or A knows q A knows who the president of the Congo is The man with the white hat knows p A will know p when the bell rings (partial spec. of time) A knows that B does not know p

--- When do you generate new contexts?
--- What are the cross-context inference rules?
--- How is the closed-world assumption to be applied?

If S1A does not contain q, should we conclude
A knows in S1 that q is false? or
A does not know in S1 whether q is true or false?

If S1AB does not contain q, should we conclude

In S1, A knows that B knows that q is false? or
In S1, A knows that B does not know whether q? or
In S1, A does not know whether B knows q?

Explicit Syntactic Representation

Express arbitrary sentences about knowledge in syntactic representation Use first-order theorem prover incorporating theory of strings

String operations implemented partly or wholly by procedural attachment Axioms of knowledge implemented largely by special-purpose inference rules

Example:

Axiom 1: Joe knows that a person always knows whether he's hungry
Know(Joe,"forall x know-whether(x,"hungry(@x)")")
Axiom 2: Joe knows that Fred is hungry
Know(Joe,"hungry(Fred)")
To Prove: Joe knows that Fred knows that he is hungry
Know(Joe,"Know(Fred,"hungry(Fred)")")

Proof:

Applying the inference rule R1, consequential closure, and the definition know-whether(A,q) <==> Know(A,q) v Know(A, ~q) to Axiom 1 gives: 3. Know(Joe,"forall x Know(x, "hungry(@x)" v

Know(x,"~hungry(@x)")")

Applying R1 plus the axiom of veridicality plus the propositional axiom (P ==> Q) ==> (P ==> (P & Q)) to 3. gives 4. Know(Joe,

"forall x (hungry(x) & Know(x,"hungry(@x)")) v

~ hungry(x) & Know(x,~hungry(@x)"))")

Applying R1 to 2. and 4. gives

5. Know(Joe, "hungry(Fred) & Know(Fred, "hungry(Fred)")") Applying R1 to 5 gives

6. Know(Joe, "Know(Fred, "hungry(Fred)")")

Problem: Immense search space. How to control search?

Inference with Possible Worlds

Technique: Translate all statements into first-order language of possible worlds.

Apply first-order theorem proving techniques.

Example:

Axiom 1: Joe knows that a person always knows if he's hungry forall W1

K(Joe,w0,W1) ==>

forall x ((forall W2 K(X,W1,W2) ==> hungry(X,W2)) or forall W3 K(X,W1,W3) ==> ~ hungry(X,W3))) Axiom 2: Joe knows that Fred is hungry forall W4 K(Joe,w0,W4) ==> hungry(Fred,W4) To prove: Joe knows that Fred knows that he is hungry forall W5 K(Joe,w),W5) ==>

forall W6 K(Fred, W5, W6) ==> hungry(Fred, W6)

Skolemizing:

- 1. ~K(Joe,w0,W1) v ~K(X,W1,W2) v hungry(X,W2) v ~K(X,W1,W3)) v ~hungry(X,W3)
- 2. ~K(Joe,s0,S4) v hungry(fred,W4)

Negation of 3:

3A. K(Joe,w0,w5) {w5 and w6 are Skolem constants}3B. K(fred,w5,w6)3C. ~hungry(Fred,w6)

Skolemization of reflexivity:

4. K(X,W,W)

The resolution proof is then immediate

Comparison to syntactic representation

--- Much more controlled inference path

--- Somewhat less expressive language

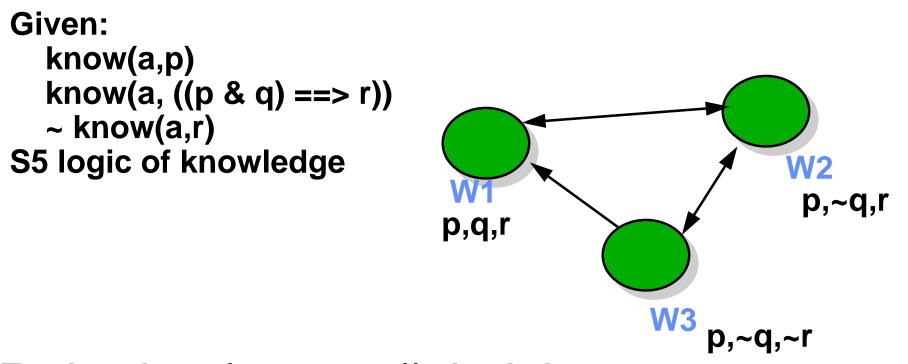
--- Substantially less intuitive representation and proof structure

Vivid Representation (Grove and Halpern, '92)

Construct an actual model of the theory as a set of possible worlds (or a collection of models).

What is true in the model(s) may be a consequence of the theory.

Example:



To show know(a, $\sim r ==> \sim q$)) check that $\sim r ==> \sim q$ holds in every accessible world.

To show ~know(a,q) show that q is false in some accessible world.

Problem: Distinguish between the consequences of the theory and random features of the model (e.g., ~know(a, q <==> r) holds because q<==> r false in W2. But it's not a consequence of the theory.

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Implementations

--- Restricted to toy programs --- Planning PAWTUCKET UWL

UWL (Etzioni et. al.) Modified TWEAK planner, find out variable bindings

Goal: (satisfy (color chair ?c) (satisfy (color table ?c) (handsoff (color table ?tc))

Make the chair the same color as the table, but not by changing the color of the table

Actions: (SENSE-COLOR ?object ! color) Effects: ((observe ?object !color))

Action: (GET-PAINT ? color) Effects: (have-color ?color)

Name: (PAINT ?obj ?color) Preconds: (satisfy ((have-color ?color))) Effects(cause ((color ?obj ?color)))



(sense-color table !color) (get-paint !color) (paint chair ! color)

Summary

--- Logics of knowledge and belief are needed for many Al applications --- planning, speech acts, distributed systems

---- Modal logics, Syntactic logics can be used to represent knowledge

--- Many extensions needed for commonsense reasoning:

--- time, default reasoning --- Much future work ahead

--- concrete applications, multiple agents, consequential closure

Pawtucket (Davis, unpublished)

Situation:

John knows that Bill knows Mary's phone no John knows that phone1 is a telephone

Wanted:

A plan for John to call Mary's no.

Causal rules:

A way to call x is to dial x's no. on the phonme The preconditions of B telling P to A in S are that A and B are at the same place and that B knows P is true

Plan:

do(john,request(bill,do(bill, tell(john,a_q(n)))),
do(bill,tell(john,a_q(n))),
do(john, dial(a_q(n), phone1))