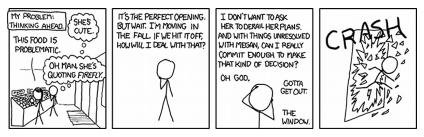
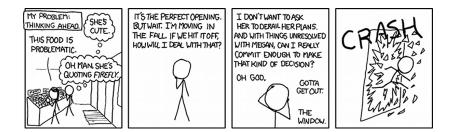
CS325 Artificial Intelligence Chs. 10, 11 – Planning

Cengiz Günay, Emory Univ.



Spring 2013

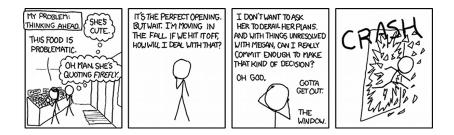
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Planning is coming up with a solution for an agent:

• it's at the heart of AI

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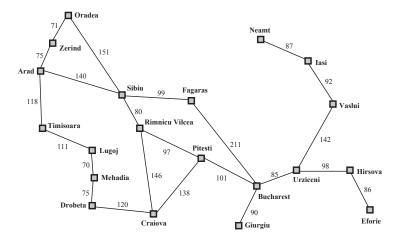
Exit survey: Knowledge Representation and Inference

- What knowledge representation do you think our brains have?
- What knowledge base topic/domain would you wish you had?

Entry survey: Planning (0.25 points of final grade)

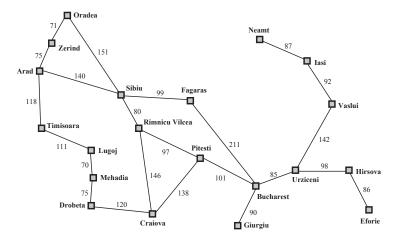
- What previous class topic would count as a planning algorithm?
- Where do you think you used an automated planning system?

Graph Search Is a Form of Planning



• Search algorithm like A^* is a planning method

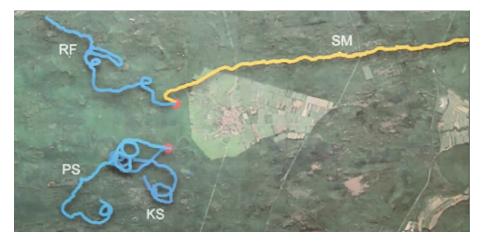
Graph Search Is a Form of Planning



- Search algorithm like A* is a planning method
- Only works for observable and deterministic environment
- What if you cannot plan all the way?

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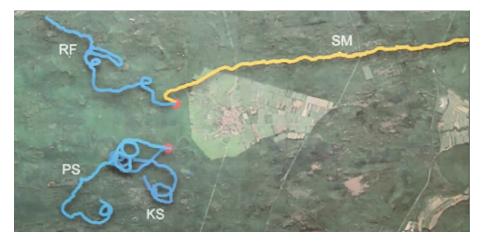
Blind Walking





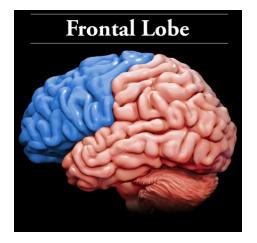
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Blind Walking



• Need to alternate plan & execution

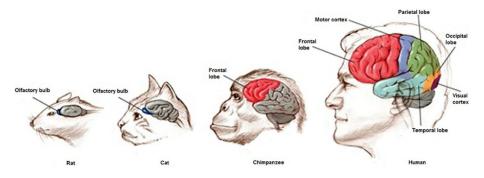
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Chs. 10, 11 – Planning

How Does the Brain Do It?

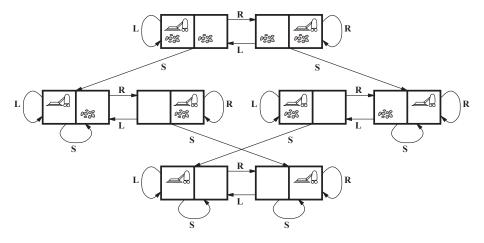


• Frontal lobes are biggest in humans, dedicated to **planning**, **reasoning** and other high-order skills

- Stochastic environment
- Multi-agent situation
- Partial observable
- Unknown
- Hierarchical

- Stochastic environment
- Multi-agent situation
- Partial observable
- Unknown
- Hierarchical

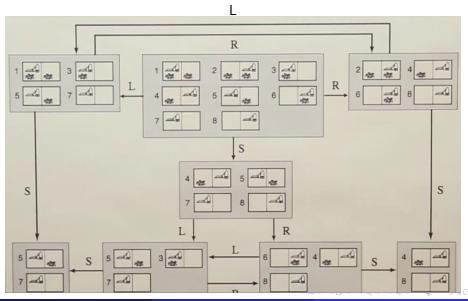
In these cases, may need to plan in belief states



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Sensorless World

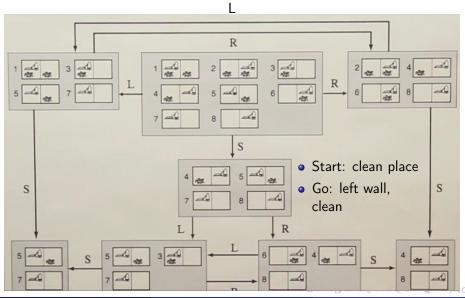


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Sensorless World

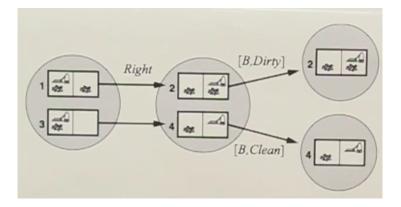


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Chs. 10, 11 - Planning

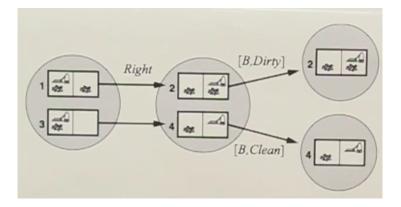
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Partially Observable Environment

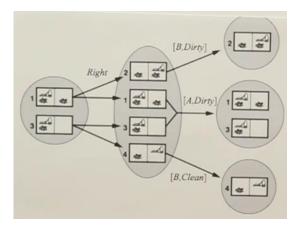


• Deterministic, but sense only local dirt & location

Partially Observable Environment

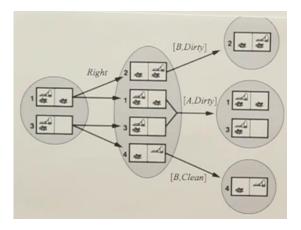


- Deterministic, but sense only local dirt & location
- Belief states get smaller with actions



• Still local sensing

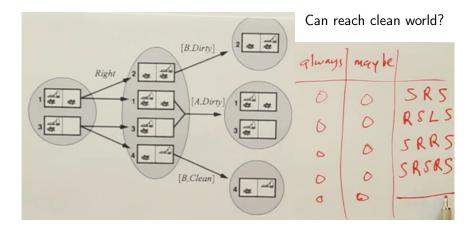
• Robot has slippery wheels, may not move



- Still local sensing
- Robot has slippery wheels, may not move
- Belief states get larger with actions

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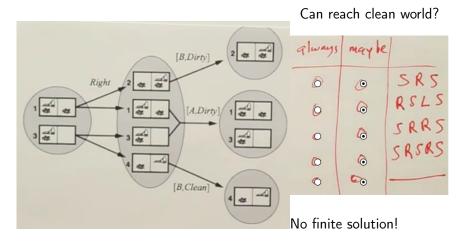
Chs. 10, 11 – Planning



- Still local sensing
- Robot has slippery wheels, may not move
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Chs. 10, 11 – Planning

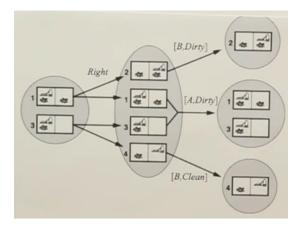


- Still local sensing
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Chs. 10, 11 – Planning

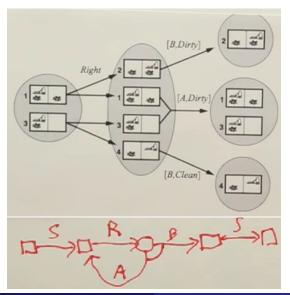
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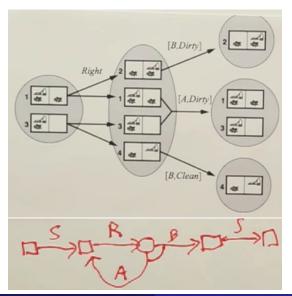
- Finite sequences don't work:
 - [R, S, R, S]

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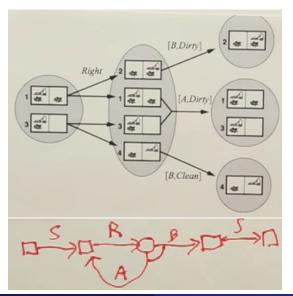
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- Finite sequences don't work:
 - [*R*, *S*, *R*, *S*]
- Make a finite plan with branches



- Finite sequences don't work:
 - [*R*, *S*, *R*, *S*]
- Make a finite plan with branches
- Gives an infinite sequence!

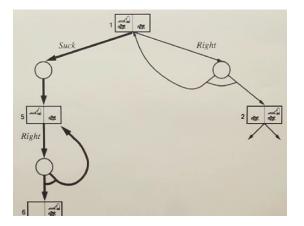


- Finite sequences don't work:
 - [R, S, R, S]
- Make a finite plan with branches
- Gives an infinite sequence!

Can write it like:

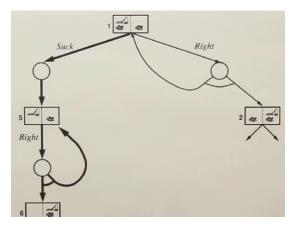
 $[S, \mathrm{While}\, A:\, R, S]$

Searching For a Plan



Like searching for paths, find a plan that reaches the goal.

Searching For a Plan



Unbounded solution:

- Some leaf goal
- Every leaf goal
- No

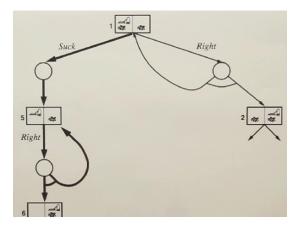
Bounded solution:

- No branch
- No loops

• No

Like searching for paths, find a plan that reaches the goal.

Searching For a Plan



Unbounded solution:

- Some leaf goal
- Every leaf goal
- No

Bounded solution:

- No branch
- No loops

• No

Like searching for paths, find a plan that reaches the goal.

[A, S, F] Result(Result($A, A \rightarrow S$), $S \rightarrow F$) \in Goals Deterministic:

$$s' = \operatorname{Result}(a, s)$$

where s is state and a is action.

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 $[A,S,F] \quad {\rm Result}({\rm Result}(A,A\to S),S\to F)\in {\rm Goals}$ Deterministic:

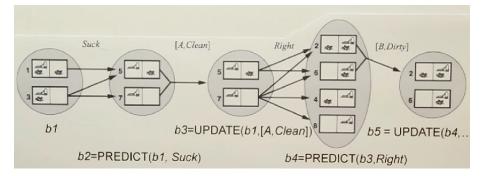
$$s' = \operatorname{Result}(a, s)$$

where *s* is state and *a* is action. Non-deterministic: Predict-update beliefs (*b*) cycle:

b' = Update(Predict(b, a), o)

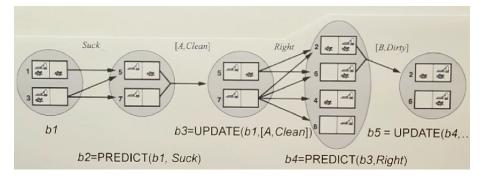
where *o* is observation.

Predict-Update Cycle



• Babies randomly dirtying floors

Predict-Update Cycle



• Babies randomly dirtying floors

Problems:

• Large solution, simpler to list world state as variables

State space: *k* boolean variables; size:



State space: k boolean variables; size: 2^k



State space: k boolean variables; size: 2^k World state: Complete assignment Belief state: Complete/partial/arbitrary formula

• Called an action schema

- Action
- Precondition
- In Effect



- Action
- Precondition
- I Effect

```
Example:
Action(Fly(p, from, to),
Pre: At(p, from) \land Plane(p) \land Airport(from) \land Airport(to)
Eff: \neg At(p, from) \land At(p, to))
```

- Action
- Precondition
- I Effect

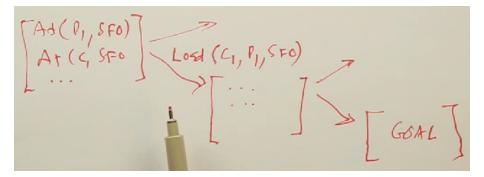
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Example:
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```

Is it FOL?

 $Init(At(C_1, SFO) \land At(C_2, JFK) \land At(P_1, SFO) \land At(P_2, JFK))$ $\wedge Cargo(C_1) \wedge Cargo(C_2) \wedge Plane(P_1) \wedge Plane(P_2)$ $\land Airport(JFK) \land Airport(SFO))$ $Goal(At(C_1, JFK) \land At(C_2, SFO))$ Action(Load(c, p, a))**PRECOND:** $At(c, a) \land At(p, a) \land Cargo(c) \land Plane(p) \land Airport(a)$ EFFECT: $\neg At(c, a) \land In(c, p)$ Action(Unload(c, p, a)),**PRECOND:** $In(c, p) \land At(p, a) \land Cargo(c) \land Plane(p) \land Airport(a)$ EFFECT: $At(c, a) \land \neg In(c, p)$) Action(Fly(p, from, to), **PRECOND:** $At(p, from) \land Plane(p) \land Airport(from) \land Airport(to)$ EFFECT: $\neg At(p, from) \land At(p, to))$

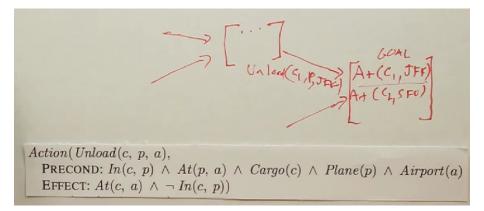
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• Like regular search.

Regression/Backward Search



Start with goal, have many unknowns.

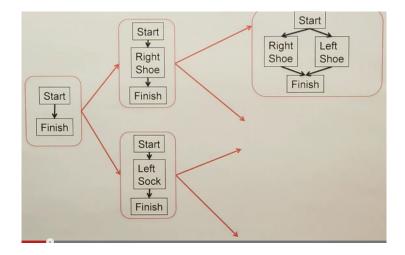
When to Choose Forward vs. Backward?

Actin (Buy (b) PRE: ISBN (b) EFF: OWN (b) EFF: OWN (b) Goal (Oum (0136042597))

- Remember when to choose breadth-first vs. depth first.
- Fewer options when starting from the goal.
- Can think of other examples?

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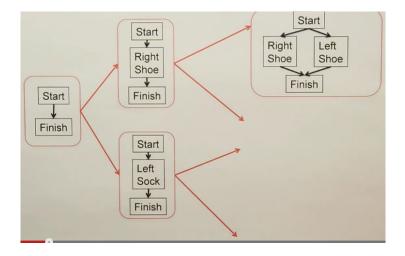
Searching the "Plan Space"



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Searching the "Plan Space"



• How would one use genetic algorithms?

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Action(Slide(t, a, b), Pre: On(t, a) \land Tile(t) \land Blank(b) \land Adj(a, b) Eff: On(t, b) \land Blank(a) $\land \neg$ On(t, a) $\land \neg$ Blank(b))



Action(Slide(t, a, b), Pre: On(t, a) \land Tile(t) \land Blank(b) \land Adj(a, b) Eff: On(t, b) \land Blank(a) $\land \neg$ On(t, a) $\land \neg$ Blank(b))

How to use heuristics? Remember Romania routes.



Action(Slide(t, a, b), Pre: On(t, a) \land Tile(t) \land Blank(b) \land Adj(a, b) Eff: On(t, b) \land Blank(a) $\land \neg$ On(t, a) $\land \neg$ Blank(b))

How to use heuristics? Remember Romania routes.

- Choose approximate solutions and use during search
- Delete terms from schema

Action(Slide(t, a, b), Pre: On(t, a) \land Tile(t) \land Blank(b) \land Adj(a, b) Eff: On(t, b) \land Blank(a) $\land \neg$ On(t, a) $\land \neg$ Blank(b))

How to use heuristics? Remember Romania routes.

- Choose approximate solutions and use during search
- Delete terms from schema
- Can also do this programmatically

Uses First Order Logic (FOL) for planning: Actions: are objects; e.g., Fly(p, x, y)Situations: objects; e.g., s' = Result(s, a)Fluents: change at each situation: e.g., At(p, x, s)Possible actions: objects; Poss(a, s)Written as in: $Pre(s) \Rightarrow Poss(a, s)$ Uses First Order Logic (FOL) for planning: Actions: are objects; e.g., Fly(p, x, y)Situations: objects; e.g., s' = Result(s, a)Fluents: change at each situation: e.g., At(p, x, s)Possible actions: objects; Poss(a, s)Written as in: $Pre(s) \Rightarrow Poss(a, s)$

Plane example:

$$\begin{split} \text{Plane}(p,s) \land \text{Airport}(x,s) \land \text{Airport}(y,s) \land \text{At}(p,x,s) & \Rightarrow \\ & \text{Poss}(\text{Fly}(p,x,y),s) \end{split}$$

Effects: Successor state axioms:

 $\forall a, s \operatorname{Poss}(a, s) \Rightarrow (\operatorname{fluent} \Leftrightarrow \operatorname{caused} \operatorname{it} \land \operatorname{didn't} \operatorname{undo} \operatorname{it})$



Effects: Successor state axioms:

 $\forall a, s \operatorname{Poss}(a, s) \Rightarrow (\operatorname{fluent} \Leftrightarrow \operatorname{caused} \operatorname{it} \land \operatorname{didn't} \operatorname{undo} \operatorname{it})$

Cargo example:

$$\operatorname{Poss}(a, s) \Rightarrow \operatorname{In}(c, p, \operatorname{Result}(s, a)) \Leftrightarrow (a = \operatorname{Load}(c, p, x) \lor (\operatorname{In}(c, p, s) \land a \neq \operatorname{Unload}(c, p, x)))$$

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Chs. 10, 11 – Planning

≣ ▶ ৰ ≣ ▶ ≣ ৩৫৫ Spring 2013 24 / 24 Effects: Successor state axioms:

 $\forall a, s \operatorname{Poss}(a, s) \Rightarrow (\operatorname{fluent} \Leftrightarrow \operatorname{causedit} \land \operatorname{didn't} \operatorname{undoit})$

Cargo example:

$$\operatorname{Poss}(a, s) \Rightarrow \operatorname{In}(c, p, \operatorname{Result}(s, a)) \Leftrightarrow$$

 $(a = \text{Load}(c, p, x) \lor (\text{In}(c, p, s) \land a \neq \text{Unload}(c, p, x)))$

- Can use power of FOL
- Can use existing theorem provers
- Problem: slow