

# Principles of AI Planning

February 9, 2007 — Complexity of nondeterministic planning with partial observability

## Introduction

### Complexity results

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- Membership in 2-EXP
- Idea of the 2-EXP hardness reduction
- 2-EXP hardness reduction
- 2-EXP-completeness proof

## Summary

# Principles of AI Planning

Complexity of nondeterministic planning with partial observability

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## Introduction

## Introduction

- ▶ Earlier, we showed how **deterministic Turing Machines with polynomial space** can be translated to **deterministic planning tasks**.
- ▶ Later, we saw how **alternation** in Turing Machines can be translated into **nondeterminism** in the planning task.
- ▶ We also saw how **exponential space** in Turing Machines can be modeled by using **unobservable** planning tasks.

Now, we will combine the latter two proof techniques to show that nondeterministic planning with partial observability is 2-EXP-complete.

## Complexity Problem

## The strong planning problem for partial observability

**PARTIALPLANEX** (plan existence for partial observability)

GIVEN: nondeterministic planning task  $\langle A, I, O, G, V \rangle$

QUESTION: Is there a strong plan for the task?

- ▶ We do **not** consider the analog of the bounded plan existence problem (PLANLEN).

## Membership in 2-EXP

### PARTIALPLANEX $\in$ 2-EXP

For input  $\mathcal{T}$ :

- ▶ Use the reduction algorithm presented in the previous lecture to generate an equivalent nondeterministic planning task with full observability  $\mathcal{T}'$  in exponential time.
  - ▶ This requires exponential time and creates a task of exponential size in  $\|\mathcal{T}\|$ .
- ▶ Solve the resulting task using an EXP algorithm.
  - ▶ This requires exponential time in  $\|\mathcal{T}'\|$ , which is doubly exponential in  $\|\mathcal{T}\|$ .

Thus, the problem can be solved within 2-EXP.

## Reduction idea

- ▶ We want to prove that PARTIALPLANEX is 2-EXP-hard.
- ▶ To do this, we need to reduce **all** problems in 2-EXP to PARTIALPLANEX.
- ▶ A problem is in 2-EXP iff there exists a DTM that accepts instances of the problem in doubly exponential time.
- ▶ Equivalently, by Chandra et al.'s theorem, a problem is in 2-EXP iff there exists an **ATM** that accepts instances of the problem in exponential space (since AEXPSPACE = 2-EXP).
- ▶ We exploit the latter relationship by providing a **generic reduction** from word acceptance for ATMs with exponential space to PARTIALPLANEX.

## Proving hardness for 2-EXP

### Overview

- ▶ For a fixed polynomial  $p$ , given ATM  $M$  and input  $w$ , generate planning task which is solvable by a strong plan iff  $M$  accepts  $w$  in space  $2^{p(|w|)}$ .

For simplicity, we only consider ATMs with two restrictions (no loss of generality):

- ▶ ATM never moves to the left of the initial head position.
- ▶ If several ATM transitions are possible in universal state  $q$  reading the symbol  $a$ , then the resulting state  $q'$  is different for all these transitions.

(The second restriction is so that the planning agent can know which transition was taken by looking at the current state.)

## Idea of the reduction

### Dealing with alternation

- ▶ **Existential states** of the ATM are modeled by states of the planning task where there are **several applicable operators** to choose from.
- ▶ **Universal states** of the ATM are modeled by states of the planning task where there is **a single applicable operator with a nondeterministic effect**.

## Idea of the reduction

Dealing with exponential space

- ▶ Only keep track of the contents of **one** tape cell  
 $\rightsquigarrow$  **watched tape cell**.
- ▶ **Which** tape cell is watched is unobservable.
- ▶  $\rightsquigarrow$  Plan must work correctly for **all possible choices**.
- ▶  $\rightsquigarrow$  Plan must remain faithful to the TM computation.

## Reduction: state variables

Let  $p$  be a polynomial such that  $2^{p(n)} - 2$  is a space bound for inputs of size  $n$ .

Given: ATM  $\langle \Sigma, \square, Q, q_0, l, \delta \rangle$  and input  $w_1 \dots w_n$ .

### State variables

Convention:

Use  $\overline{\phantom{x}}$  to denote **vectors** of  $p(n)$  state variables encoding a number in the range  $0 \dots 2^{p(n)} - 1$ .

- ▶  $\text{state}_q$  for all  $q \in Q$  – current TM state
- ▶  $\overline{\text{head}}$  – head position
- ▶  $\overline{\text{watched}}$  – position of the watched tape cell
- ▶  $\text{content}_a$  for all  $a \in \Sigma \setminus \square$  – contents of the watched tape cell

The  $\overline{\text{watched}}$  variables are unobservable.

All other variables are observable.

## Spelling it out

- ▶  $(\overline{\text{head}} = 1) \equiv \neg \text{head}_1 \wedge \dots \wedge \neg \text{head}_{p(n)-1} \wedge \text{head}_{p(n)}$
- ▶  $(\overline{\text{head}} = 5) \equiv \neg \text{head}_1 \wedge \dots \wedge \neg \text{head}_{p(n)-3} \wedge \text{head}_{p(n)-2} \wedge \neg \text{head}_{p(n)-1} \wedge \text{head}_{p(n)}$
- ▶  $(\overline{\text{head}} = \overline{\text{watched}}) \equiv$   
 $(\neg \text{head}_1 \vee \text{watched}_1) \wedge (\text{head}_1 \vee \neg \text{watched}_1)$   
 $\wedge (\neg \text{head}_2 \vee \text{watched}_2) \wedge (\text{head}_2 \vee \neg \text{watched}_2)$   
 $\wedge \dots$
- ▶  $\overline{\text{head}} := \overline{\text{head}} + 1 \equiv$   
 $(\neg \text{head}_{p(n)} \supset \text{head}_{p(n)})$   
 $\wedge ((\neg \text{head}_{p(n)-1} \wedge \text{head}_{p(n)}) \supset (\text{head}_{p(n)-1} \wedge \neg \text{head}_{p(n)}))$   
 $\wedge \dots$
- ▶  $\overline{\text{head}} := \overline{\text{head}} - 1 \equiv \dots$

## Reduction: initial state formula

### Initial state formula

$$\begin{aligned}
 I = & \text{state}_{q_0} \wedge \bigwedge_{q \in Q \setminus \{q_0\}} \neg \text{state}_q \\
 & \wedge \overline{\text{head}} = 1 \\
 & \wedge \left( \bigwedge_{i=1}^n ((\overline{\text{watched}} = i) \rightarrow \text{content}_{w_i}) \right) \\
 & \wedge (\overline{\text{watched}} = 0 \vee \overline{\text{watched}} > n) \rightarrow \text{content}_{\square} \\
 & \wedge \bigwedge_{a \in \Sigma \setminus \square} \bigwedge_{a' \in \Sigma \setminus \{a\}} \neg (\text{content}_a \wedge \text{content}_{a'})
 \end{aligned}$$

Note: watched tape cell **unspecified**

## Reduction: operators

### Operators

For each transition rule  $((q, a), (q', a', \Delta)) \in \delta$ , define:

► **precondition:**

$$\begin{aligned} \text{pre}_{q,a} := & \text{state}_q \\ & \wedge ((\overline{\text{head}} = \overline{\text{watched}}) \rightarrow \text{content}_a) \\ & \wedge \overline{\text{head}} > 0 \\ & \wedge \overline{\text{head}} < 2^{p(n)} - 1 \end{aligned}$$

► **effect:**

$$\begin{aligned} \text{eff}_{q,a,q',a',\Delta} := & \neg \text{state}_q \wedge \text{state}_{q'} \\ & \wedge ((\overline{\text{head}} = \overline{\text{watched}}) \triangleright \neg \text{content}_a) \\ & \wedge ((\overline{\text{head}} = \overline{\text{watched}}) \triangleright \text{content}_{a'}) \\ & \wedge (\overline{\text{head}} := \overline{\text{head}} + \Delta) \end{aligned}$$

If  $q = q'$ , omit the effects in the first line.

If  $a = a'$ , omit the effects in the second and third line.

## Reduction: operators (continued)

### Operators (ctd.)

For **existential** states  $q \in Q_{\exists}$ ,  $a \in \Sigma_{\square}$ :

Let  $(q'_j, a'_j, \Delta_j)_{j \in \{1, \dots, k\}}$  be those triples with  $((q, a), (q'_j, a'_j, \Delta_j)) \in \delta$ .

For each  $j \in \{1, \dots, k\}$ , introduce one operator:

► precondition:  $\text{pre}_{q,a}$

► effect:  $\text{eff}_{q,a,q'_j,a'_j,\Delta_j}$

## Reduction: operators (continued)

### Operators (ctd.)

For **universal** states  $q \in Q_{\forall}$ ,  $a \in \Sigma_{\square}$ :

Let  $(q'_j, a'_j, \Delta_j)_{j \in \{1, \dots, k\}}$  be those triples with  $((q, a), (q'_j, a'_j, \Delta_j)) \in \delta$ .

Introduce only one operator:

► precondition:  $\text{pre}_{q,a}$

► effect:  $\text{eff}_{q,a,q'_1,a'_1,\Delta_1} \mid \dots \mid \text{eff}_{q,a,q'_k,a'_k,\Delta_k}$

## Reduction: goal

### Goal

$$G = \bigvee_{q \in Q_{\forall}} \text{state}_q$$

## 2-EXP-completeness of strong planning with partial observability

Theorem (Rintanen, 2002)

$\text{PARTIALPLANEX}$  is 2-EXP-complete.

Proof.

Membership in 2-EXP has been shown by providing doubly exponential-time algorithms that generate strong plans (and decide if one exists as a side effect).

Hardness follows from the previous generic reduction for ATMs with exponential space bound and Chandra et al.'s theorem (showing  $\text{AEXPSPACE} = 2\text{-EXP}$ ).  $\square$

## Summary

- Nondeterministic planning with partial observability is very hard.