

Length of plans

Let $\langle A, I, O, G \rangle$ be a deterministic succinct transition system.

- 1 There is a plan of length 0 iff $I \models G$.
- 2 Shortest plans may not be longer than $2^n - 1$: If a plan is longer, then it visits some state s more than once and has the form $\sigma_1^s \sigma_2^s \sigma_3$: the plan $\sigma_1 \sigma_3$ is shorter.
- 3 Shortest plan may have length $2^n - 1$: Reach the goal state $111 \dots 1$ from the initial state $000 \dots 0$ by an operator that increments the corresponding binary number $2^n - 1$ times.

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Deterministic planning: expressivity

Definition

The decision problem SAT: test whether a given propositional formula ϕ is satisfiable.

Reduction from SAT to deterministic planning

A = the set of propositional variables occurring in ϕ

I = any state, e.g. all state variables have value 0

$O = (\{T\} \times A) \cup (\{\{T, \neg a\} \mid a \in A\})$

There is a plan for $\langle A, I, O, \phi \rangle$ if and only if ϕ is satisfiable.

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Deterministic planning: expressivity

- Because there is a polynomial-time translation from SAT into deterministic planning, and SAT is an NP-complete problem, there is a polynomial time translation from **every decision problem in NP** into deterministic planning. Hence the problem is NP-hard.
- Does deterministic planning have the power of NP, or is it still more powerful?
- We show that it is more powerful: The decision problem of testing whether a plan exists is PSPACE-complete.

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Definition

A **Turing machine** $\langle \Sigma, Q, \delta, q_0, g \rangle$ consists of

- 1 an alphabet Σ (a set of symbols),
- 2 a set Q of internal states,
- 3 a transition function δ that maps $\langle q, s \rangle$ to a tuple $\langle s', q', m \rangle$ where $q, q' \in Q$, $s \in \Sigma \cup \{ |, \square \}$, $s' \in \Sigma$ and $m \in \{ L, N, R \}$.
- 4 an initial state $q_0 \in Q$, and
- 5 a labeling $g : Q \rightarrow \{ \text{accept}, \text{reject}, \exists \}$ of states.

Turing machines

Example

TM accepting strings $\epsilon, a, ab, aba, abab, \dots$ is $\langle \Sigma, Q, \delta, q_1, g \rangle$
where

$$\Sigma = \{a, b\}$$

$$Q = \{q_1, q_2, q_3, q_4\}$$

$$g(q_1) = \exists$$

$$g(q_3) = \text{accept}$$

$$\delta(q_1, a) = \langle a, q_2, R \rangle$$

$$\delta(q_2, b) = \langle b, q_1, R \rangle$$

$$\delta(q_1, \square) = \langle a, q_3, R \rangle$$

$$\delta(q, s) = \langle a, q_4, R \rangle \text{ for all other } q, s$$

$$g(q_2) = \exists$$

$$g(q_4) = \text{reject}$$

$$\delta(q_1, b) = \langle b, q_4, R \rangle$$

$$\delta(q_2, a) = \langle a, q_4, R \rangle$$

$$\delta(q_2, \square) = \langle b, q_3, R \rangle$$

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What does the TM do with the string $ababb$?

$$q_1 \mid \widehat{a}babb \square$$
$$q_2 \mid a\widehat{b}abb \square$$
$$q_1 \mid ab\widehat{a}bb \square$$
$$q_2 \mid ab\widehat{a}bb \square$$
$$q_1 \mid ababb\widehat{b} \square$$
$$q_4 \mid ababb\widehat{\square}$$

The label $g(q_4) = \text{reject}$. The TM does not accept the string.

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Some complexity classes

Definition

$DTIME(f)$ is the class of decision problems solved by a **deterministic** Turing machine in $\mathcal{O}(f(n))$ time when n is the input string length. $NTIME(f)$ is defined similarly for **nondeterministic** Turing machines.

$DSPACE(f)$ is the class of decision problems solved by a **deterministic** Turing machine in $\mathcal{O}(f(n))$ space when n is the input string length.

$$EXSPACE = \bigcup_{k \geq 0} DSPACE(2^{n^k})$$

$$NEXPTIME = \bigcup_{k \geq 0} NTIME(2^{n^k})$$

$$EXPTIME = \bigcup_{k \geq 0} DTIME(2^{n^k})$$

$$PSPACE = \bigcup_{k \geq 0} DSPACE(n^k)$$

$$NP = \bigcup_{k \geq 0} NTIME(n^k)$$

$$P = \bigcup_{k \geq 0} DTIME(n^k)$$

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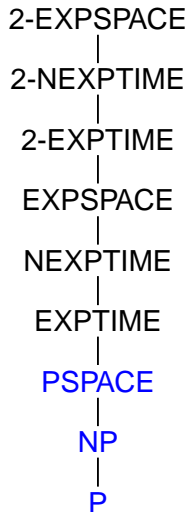
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Some complexity classes



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Simulation of PSPACE Turing machines

Close match between space-bounded Turing machines and planning problems.

- 1 Turing machine configurations \sim states
- 2 Turing machine transitions \sim operators
- 3 initial configuration \sim initial state
- 4 accepting configurations \sim goal states

For simulation of PSPACE TMs a number of state variables and operators that is **polynomial** in input string length suffices.

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Simulation of PSPACE Turing machines

We show how polynomial-space Turing machines can be simulated by planning.

- Tape cell contents are represented in state variables.
- R/W head location is represented in state variables.
- Internal TM state is represented in state variables.
- Transitions are represented as operators.

Theorem

A Turing machine M accepts an input string σ if and only if $T(M, \sigma) = \langle A, I, O, G \rangle$ has a plan.

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Simulation of PSPACE Turing machines

Turing machine with $\Sigma = \{a, b, c\}$, input string of length $n = 4$, space bound $p(n) = n^2 = 16$, internal states $Q = \{q_1, q_2, q_3\}$.

State variables in the corresponding planning problem:

tape cell:	state variables for tape cells						
	0	1	2	3	...	15	16
R/W head:	h_0	h_1	h_2	h_3	...	h_{15}	h_{16}
symbol a :	a_0	a_1	a_2	a_3	...	a_{15}	a_{16}
symbol b :	b_0	b_1	b_2	b_3	...	b_{15}	b_{16}
symbol c :	c_0	c_1	c_2	c_3	...	c_{15}	c_{16}
symbol \square :	\square_0	\square_1	\square_2	\square_3	...	\square_{15}	\square_{16}
symbol $ $:	$ _0$	$ _1$	$ _2$	$ _3$...	$ _{15}$	$ _{16}$
state q_1 :	q_1						
state q_2 :	q_2						
state q_3 :	q_3						

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TM config.	state variable values					plan	
	1	2	3	4	5		
$q_1 \widehat{a}babbb \square$	$\widehat{1}00$	$0\widehat{1}0$	100	010	010	1000	$o_{a,q_1,1}$
$q_2 \widehat{b}babbb \square$	010	$\widehat{0}10$	100	010	010	0100	$o_{b,q_2,2}$
$q_4 bc\widehat{a}bb \square$	010	001	$\widehat{1}00$	010	010	0001	$o_{a,q_4,3}$
$q_3 bcb\widehat{b}b \square$	010	001	010	$\widehat{0}10$	010	0010	$o_{b,q_3,4}$
$q_1 bcb\widehat{c}cb \square$	010	001	$\widehat{0}10$	001	010	1000	$o_{b,q_1,3}$
$q_4 bc\widehat{a}acb \square$	010	$\widehat{0}01$	100	001	010	0001	$o_{c,q_4,2}$
$q_4 \widehat{b}aacb \square$	$\widehat{0}10$	100	100	001	010	0001	$o_{b,q_4,1}$

Operator $o_{s,q,h}$ is applicable when current symbol is s , current TM state is q , and current tape cell is h .

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PSPACE simulation

Simulate a TM $\langle \Sigma, Q, \delta, q_0, g \rangle$ that needs at most $p(n)$ tape cells on an input string of length n .

State variables in the succinct transition system are

- 1 $\{q_1, \dots, q_{|Q|}\} = Q$ for the current state of the TM,
- 2 s_i for every symbol $s \in \Sigma \cup \{|\, \square\}$ and tape cell $i \in \{0, \dots, p(n)\}$,
- 3 h_i for every $i \in \{0, \dots, p(n)\}$ (position of the R/W head).

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Initial state

- 1 $I(q_0) = 1$ and $I(q) = 0$ for all $q \in Q \setminus \{q_0\}$.
- 2 $I(s_i) = 1$ if $i \leq n$ and input symbol i is s .
- 3 $I(s_i) = 0$ if $i \leq n$ and $s \in S$ and input symbol i is not s .
- 4 $I(\square_i) = 1$ iff $i \in \{n + 1, \dots, p(n) - 1\}$
- 5 $I(|_i) = 1$ iff $i = 0$
- 6 $I(h_i) = 1$ iff $i = 1$

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Goal formula

Goal formula requires that the Turing machine is in an accepting state.

$$G = \bigvee \{q \in Q \mid g(q) = \text{accept}\}.$$

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Operators

For all $s \in \Sigma \cup \{ \sqcup, \square \}$ and $q \in Q$ and $i \in \{0, \dots, p(n)\}$ with $\delta(q, s) = \langle s', q', m \rangle$ such that $m \neq R$ or $i < p(n)$ define

$$o_{s,q,i} = \langle h_i \wedge s_i \wedge q, \nu \wedge \chi \wedge \mu \rangle$$

where

ν describes the writing operation,

χ describes the change in the internal state of the TM,

μ describes the movement of the R/W head.

The requirement $m \neq R$ or $i < p(n)$ means that no transition violating the space bound is possible.

Operator $o_{s,q,i}$ corresponds to the **unique transition** from a configuration where current symbol is s , internal state is q , and R/W head location is i .

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Operators' effects

symbol written onto the tape

$$\nu = \begin{cases} \top & \text{if } s \in \{|\, s'\} \\ \neg s_i \wedge s'_i & \text{otherwise} \end{cases}$$

change in the internal state

$$\chi = \begin{cases} \neg q \wedge q' & \text{if } q \neq q' \\ \top & \text{otherwise} \end{cases}$$

movement of the R/W head

$$\mu = \begin{cases} \neg h_i \wedge h_{i-1} & \text{if } i > 0 \text{ and } m = L \\ \neg h_i \wedge h_{i+1} & \text{if } i < p(n) \text{ and } m = R \\ \top & \text{otherwise} \end{cases}$$

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- 1 Turing machine $\langle \{a, b\}, \{q_1, q_2, q_{acc}\}, \delta, q_1, g \rangle$ where δ is

	a	b	$ $	\square
q_1	$\langle a, q_1, R \rangle$	$\langle b, q_2, N \rangle$	$\langle , q_2, R \rangle$	$\langle b, q_1, N \rangle$
q_2	$\langle a, q_1, L \rangle$	$\langle a, q_{acc}, N \rangle$	$\langle , q_1, R \rangle$	$\langle a, q_2, L \rangle$
q_{acc}	—	—	—	—

and $g(q_{acc}) = \text{accept}$, $g(q_1) = \exists$ and $g(q_2) = \exists$.

(This Turing machine does not do anything interesting!)

- 2 Input string is abaab.
- 3 Let the space bound be $p(5) = 7$ for some polynomial p .

PSPACE simulation

Example

The succinct transition system corresponding to the Turing machine is $\langle A, I, O, G \rangle$ where

- 1 $A = \{q_1, q_2, q_{acc}, h_0, \dots, h_7, a_0, \dots, a_7, b_0, \dots, b_7, \dots\}$,
- 2 $I \models h_1 \wedge |_0 \wedge a_1 \wedge b_2 \wedge a_3 \wedge a_4 \wedge b_5 \wedge \square_6 \wedge \square_7 \wedge \neg h_0 \wedge \neg a_0 \wedge \neg b_0 \wedge \neg \square_0 \wedge \dots$,
- 3 operators in O are on the next slide, and
- 4 $G = q_{acc}$.

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Only part of the about

$|\{0, 1, \dots, 7\}| \times |\{q_1, q_2\}| \times |\{a, b, \sqcup, \square\}|$ operators are given below, for R/W head position 1 and input symbols a and b:

$$O = \{ \langle h_1 \wedge a_1 \wedge q_1, \neg h_1 \wedge h_2 \rangle, \dots, \\ \langle h_1 \wedge b_1 \wedge q_1, \neg q_1 \wedge q_2 \rangle, \dots, \\ \langle h_1 \wedge a_1 \wedge q_2, \neg q_2 \wedge q_1 \wedge \neg h_1 \wedge h_0 \rangle, \dots, \\ \langle h_1 \wedge b_1 \wedge q_2, \neg b_1 \wedge a_1 \wedge \neg q_2 \wedge q_{acc} \rangle, \dots \}$$

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Deterministic planning is solvable in PSPACE

- The PSPACE-hardness result provides a **lower bound** on the complexity of deterministic planning. Is the problem hard for a complexity class more difficult than PSPACE?
- We next give an **upper bound** on the complexity by showing that the problem belongs to PSPACE.
- Hence the problem is **PSPACE-complete**, locating the problem exactly in one complexity class.
- It is not known whether $NP \neq PSPACE$ or even $P \neq PSPACE$, but the result is still useful because for all practical purposes we can assume that $NP \neq PSPACE$.
- For example, we may conclude that **there is no polynomial-time translation from planning to the satisfiability problem** (the translation we gave earlier is **linear in the plan length**, which may be **exponential in the problem instance size**).

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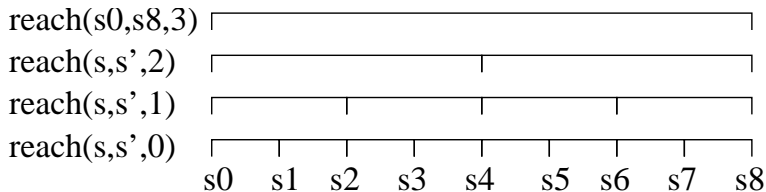
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Proof idea

Recursive algorithm for testing m -step reachability between two states with $\log m$ memory consumption.



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Algorithm

Testing whether a plan of length $\leq 2^n$ exists:

```
PROCEDURE reach( $s, s', n$ )  
IF  $n = 0$  THEN  
    IF  $s = s'$  OR  $s' = \text{app}_o(s)$  for some  $o \in O$   
    THEN RETURN true  
    ELSE RETURN false;  
ELSE  
    FOR all states  $s''$  DO  
        IF reach( $s, s'', n - 1$ ) AND reach( $s'', s', n - 1$ )  
        THEN RETURN true  
    END  
    RETURN false;
```

This algorithm does not store the plan anywhere (it could not without violating the space bound!) but could be modified to output it.

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Correctness of the algorithm

Correctness

For a succinct transition system N with n state variables, N has a plan if and only if $\text{reach}(I, s', n)$ returns true for some s' such that $s' \models G$.

Memory consumption

If number of states is 2^n , then recursion depth is n . At each recursive call only one state s'' is represented, taking space $\mathcal{O}(n)$, which means that total memory consumption at any time point is $\mathcal{O}(n^2)$, which is polynomial in the size of the succinct transition system.

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- For n Boolean state variables shortest plans have length $\leq 2^n - 1$.
- Testing for the existence of a plan is PSPACE-hard: The halting problem of every deterministic polynomial-space Turing machine can be translated into a deterministic planning problem.
- Testing for the existence of a plan can be done in PSPACE.

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