

# Analyzing decision algorithms and problems

# Outline

1. The Mathematical Method
2. Asymptotic notation and time complexity
3. Problem Complexity Classes: P, NP, PSPACE, EXPTIME

# Methods of Learning

(This is not a well-established taxonomy, just Prof. Sunberg's thoughts)

## Testimonial (?) Method

1. Read or hear a piece of knowledge
2. Decide whether to believe it
3. Learn by believing it

## Scientific Method

1. Formulate a question
2. Formulate a hypothesis
3. Test the hypothesis with an experiment
4. Learn by analyzing the results of the experiment

## Mathematical Method

1. Formulate a question
2. Create a conjecture that answers the question
  1. *Define* all concepts
  2. State as a theorem
3. Prove or disprove the conjecture
4. Learn by considering the theorem/false conjecture and the proof

**Most RL research fits into the scientific method.**

# Mathematical Proofs

A mathematical proof *must* (according to Sunberg):

1. Be logically correct
2. Allow each step to be easily verified by a member of the intended audience

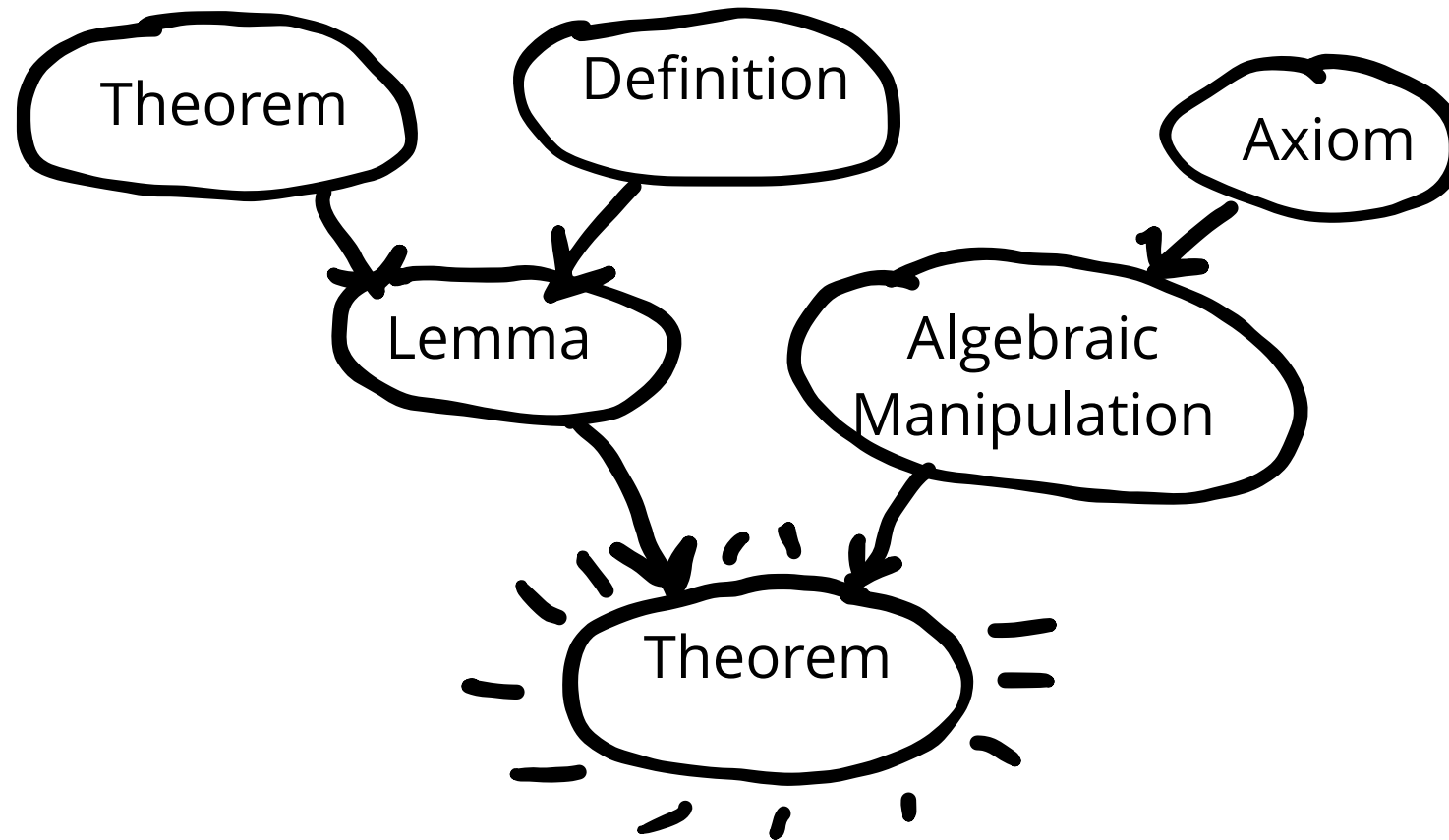
"Nice-to-have"s for theorems, definitions, proofs:

1. General (Definitions and Theorems)
2. Easy for audience to understand
3. Follows patterns the audience is familiar with
4. Composable: allows for construction of other definitions, lemmas, theorems, proofs
5. Answers the question "why" (Proofs)
6. Interesting, engaging, fun for humans

## Proof Strategies

1. **Direct proof**
2. **Proof by contradiction**
3. **Proof by induction**
4. Proof by contrapositive
5. Proof by exhaustion

# Direct Proof



# Direct Proof Example

Definition: An integer  $x$  is *even* if there exists an integer  $a$  such that  $x = 2a$

Theorem: The sum of two even integers is even.

Proof: Let  $x$  and  $y$  be even integers and  $z$  be their sum. By definition, there exist integers  $a$  and  $b$  such that

$$x = 2a \text{ and}$$

$$y = 2b.$$

Then  $z = x + y = 2a + 2b = 2(a + b)$ . Since  $a + b$  is an integer,  $z$  is even.

# Proof by Contradiction

To prove proposition  $P$ :

1. Assume  $P$  is false, i.e. assume  $\neg P$
2. Show that  $\neg P$  implies two mutually contradictory assertions,  $Q$  and  $\neg Q$
3.  $Q$  and  $\neg Q$  cannot both be true (a "contradiction"), so  $P$  must be true

# Proof by Contradiction

## Example

Theorem:  $\sqrt{2}$  is irrational.

Proof: Suppose  $\sqrt{2}$  were rational. Then there exist integers  $a$  and  $b$  such that  $\sqrt{2} = a/b$ , **at least one of which is odd** (If  $a$  and  $b$  were both even, the fraction could be reduced).

Without loss of generality, suppose  $a$  is even. Then  $a^2$  is a multiple of 4, and thus  $2b^2$  is a multiple of 4. Therefore  $b^2$  is even, and  $b$  is also even. **Thus, both  $a$  and  $b$  are even.** This is a contradiction.



# Proof by Induction

To prove  $P(n)$  holds for every natural number  $n$ :

1. Prove  $P(1)$  (the base case)
2. Prove that  $P(k) \implies P(k + 1)$  (the induction step)

# Proof by Induction Example

Theorem: For any natural number  $n$ , the sum of  $n$  and all natural numbers less than  $n$  is  $\frac{n^2+n}{2}$

Proof: By induction:

- Base case: The sum of 1 and all natural numbers less than 1 is  $1 = \frac{1^2+1}{2}$
- Induction step: If the sum of  $k$  and all natural numbers less than  $k$  is  $\frac{k^2+k}{2}$ , then the sum of  $k+1$  and all natural numbers less than  $k+1$  is  $\frac{k^2+k}{2} + k + 1 = \frac{k^2+3k+2}{2} = \frac{(k+1)^2+k+1}{2}$

# Algorithm Analysis

# Runtime Analysis

$s \leftarrow 0$

for  $i \in \{1 \dots n\}$

$s \leftarrow s + i$

end

$C_1$  1

$C_2$   $n$

$C_3$   $n$

$$C_1 + n C_2 + n C_3$$

$s \leftarrow 0$

for  $i \in \{1 \dots n\}$

for  $j \in \{1 \dots n\}$

$s \leftarrow s + i$

end

end

$C_1$  1

$C_2$   $n$

$C_3$   $n^2$

$C_4$   $n^2$

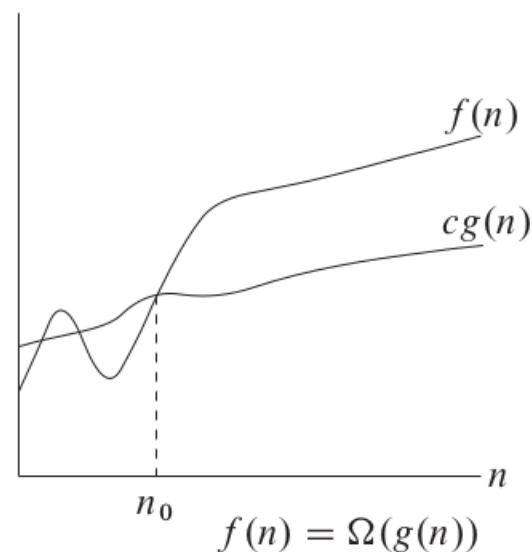
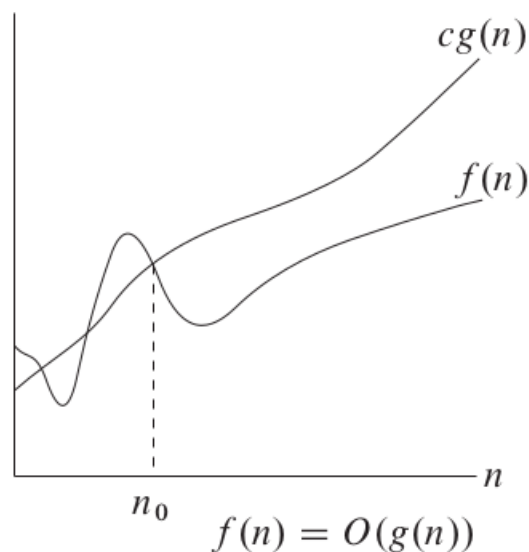
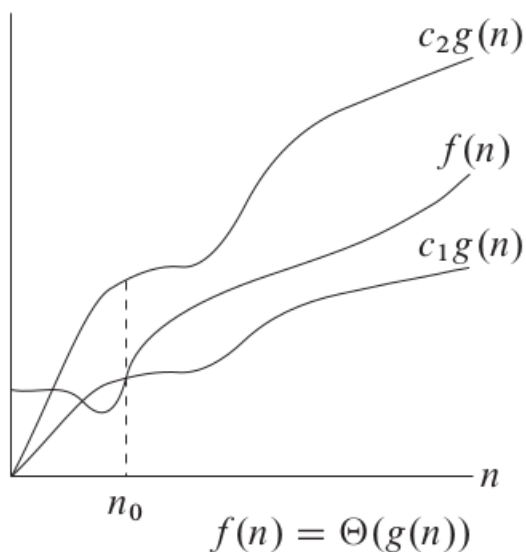
$$C_1 + C_2 n + C_3 n^2 + C_4 n^2$$

# Asymptotic Notation

$O(g(n)) = \{f(n) : \text{there exist positive constants } c \text{ and } n_0 \text{ such that}$   
 $0 \leq f(n) \leq cg(n) \text{ for all } n \geq n_0\}.$

$\Omega(g(n)) = \{f(n) : \text{there exist positive constants } c \text{ and } n_0 \text{ such that}$   
 $0 \leq cg(n) \leq f(n) \text{ for all } n \geq n_0\}.$

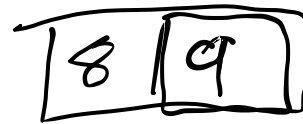
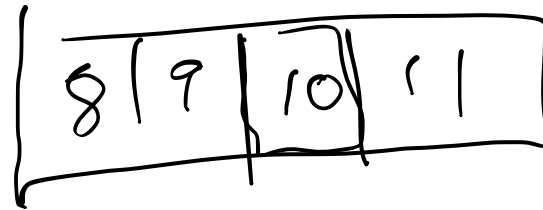
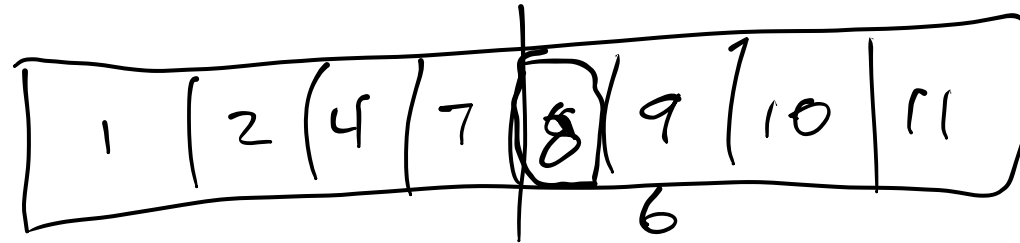
$\Theta(g(n)) = \{f(n) : \text{there exist positive constants } c_1, c_2, \text{ and } n_0 \text{ such that}$   
 $0 \leq c_1g(n) \leq f(n) \leq c_2g(n) \text{ for all } n \geq n_0\}.$ <sup>1</sup>



# Runtime Analysis

function search(A, x)	$T(n)$
mid $\leftarrow \lceil  A /2 \rceil$	$c_1$
if x = A[mid]	
return mid	$c_2$
elseif x < A[mid]	
return search(A[1:mid-1], x)	$T(\lfloor n/2 \rfloor)$
else	
return search(A[mid+1: A ], x)	$T(\lfloor n/2 \rfloor)$
end	
end	
$T(n) = \begin{cases} c_1 + c_2 & \text{if } x = A[\text{mid}] \\ c_1 + T(\lfloor n/2 \rfloor) & \text{otherwise} \end{cases}$	

Assume  $T(n) = O(\log(n))$  and prove via induction.



larger



$$x = 9$$

$$O(1)$$

$$O(\log n)$$

$$O(n)$$

$$O(n^2)$$

$$O(n^3)$$

$$O(n^k)$$

---


$$O(k^n)$$

# Tractability

What definition should we use to separate tractable from intractable problems?

$$O(1)$$

$$O(\log(n))$$

$$O(n)$$

$$O(n^2)$$

$$O(n^k)$$

$$O(k^n)$$

**Table 2.1** The running times (rounded up) of different algorithms on inputs of increasing size, for a processor performing a million high-level instructions per second. In cases where the running time exceeds  $10^{25}$  years, we simply record the algorithm as taking a very long time.

	$n$	$n \log_2 n$	$n^2$	$n^3$	$1.5^n$	$2^n$	$n!$
$n = 10$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	4 sec
$n = 30$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	18 min	$10^{25}$ years
$n = 50$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	11 min	36 years	very long
$n = 100$	< 1 sec	< 1 sec	< 1 sec	1 sec	12,892 years	$10^{17}$ years	very long
$n = 1,000$	< 1 sec	< 1 sec	1 sec	18 min	very long	very long	very long
$n = 10,000$	< 1 sec	< 1 sec	2 min	12 days	very long	very long	very long
$n = 100,000$	< 1 sec	2 sec	3 hours	32 years	very long	very long	very long
$n = 1,000,000$	1 sec	20 sec	12 days	31,710 years	very long	very long	very long

We say that a problem is *tractable* if it can be solved in  $O(n^k)$  for some  $k$ .



# Exercise: Complexity of Value Iteration for fixed horizon

**Fixed Horizon Case:**

for  $t \in \{T, T - 1, \dots, 0\}$

for  $s \in \mathcal{S}$

$$V_t(s) = \max_{a \in \mathcal{A}} (R(s, a) + \gamma \sum_{s' \in \mathcal{S}} T(s' | s, a) V_{t+1}(s'))$$

end

end

$$O(\tau |\mathcal{S}|^2 |\mathcal{A}|)$$

$|\tau|$   
 $|\mathcal{S}|$

$|\mathcal{A}| |\mathcal{S}|$

# Complexity Theory

## Which problems are hard?

# Complexity Classes

Optimization Problem

Example: Find an optimal policy for the Tiger POMDP

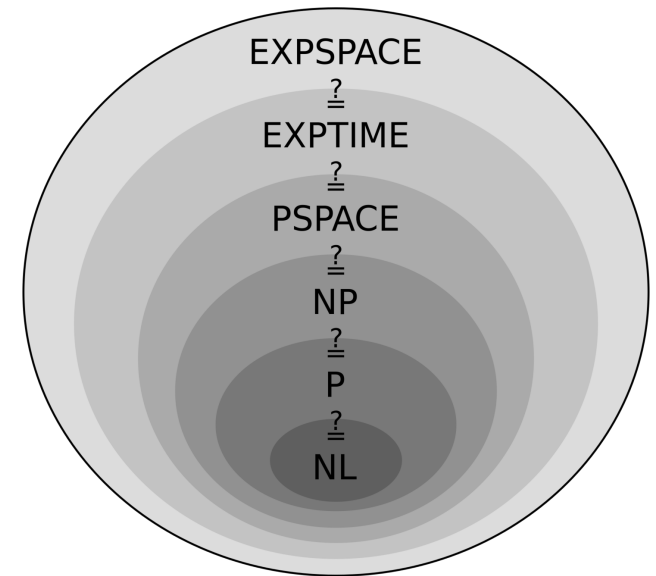
Decision Problem: answer in  $\{0, 1\}$

Example: Is "listen" an optimal first action for the Tiger POMDP

- Abstract problem: A mapping from every *instance* in a set to a solution
  - Example abstract problem: Is  $a$  an optimal first action in POMDP  $(S, A, O, R, T, Z, \gamma, b_0)$ ?
  - Example instance: Is "listen" an optimal first action for the Tiger POMDP?

# Complexity Classes

- P: Any instance can be solved in polynomial time in the size of the input
  - Example: Finite Horizon MDPs
- EXPTIME: Any instance can be solved in exponential time in the size of the input



# The NP Complexity Class

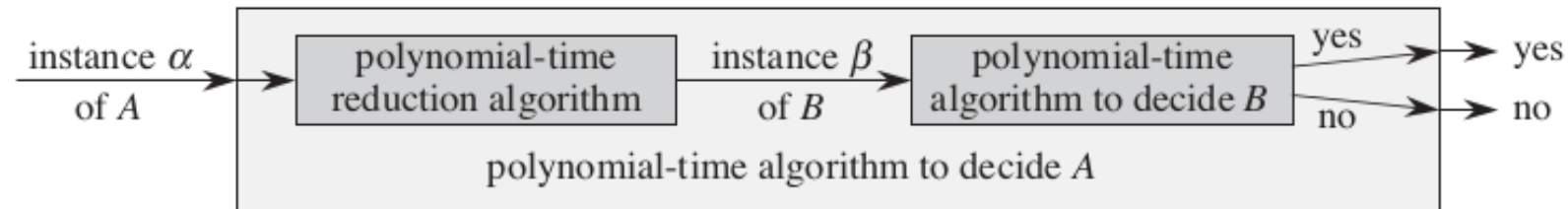
- "Nondeterministic Polynomial"
- Solution can be *verified* in polynomial time
- Is  $P \subseteq NP$ ?
- Example of a problem that is in NP but not P?

$$P \subset NP$$

(We think - this is one of the biggest  
unsolved problems in CS)

# Polynomial Reduction

How can we prove that problem B is at least as hard as problem A?



$$A \leq_P B$$

A is no harder than B

# NP Completeness

Problem  $A$  is said to be *NP-complete* if  $A \in \text{NP}$  and

$$B \leq_P A \quad \forall B \in \text{NP}.$$

Example: 3-SAT:

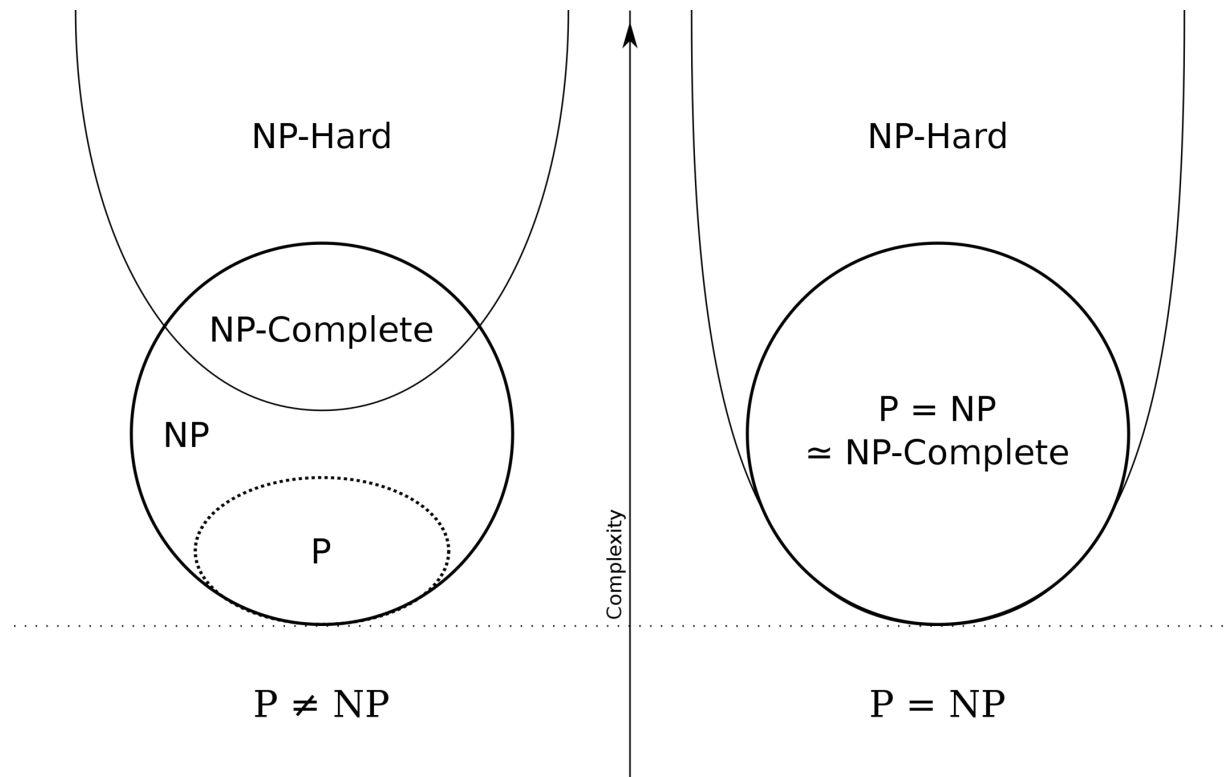
Given a set of clauses,  $C_1, \dots, C_k$ , each of length 3, over a set of binary variables  $X = \{x_1, \dots, x_n\}$ , does there exist a satisfying truth assignment?

$$(x_1 \vee x_3 \vee x_2) \wedge (\neg x_1 \vee \neg x_2 \vee \neg x_3) \wedge (\neg x_1 \vee x_2 \vee x_3)$$

- 3-SAT is NP-complete (Cook, 1971)
- If 3-SAT can be reduced to  $A \in \text{NP}$ ,  $A$  is also NP-complete!

# NP-Hardness

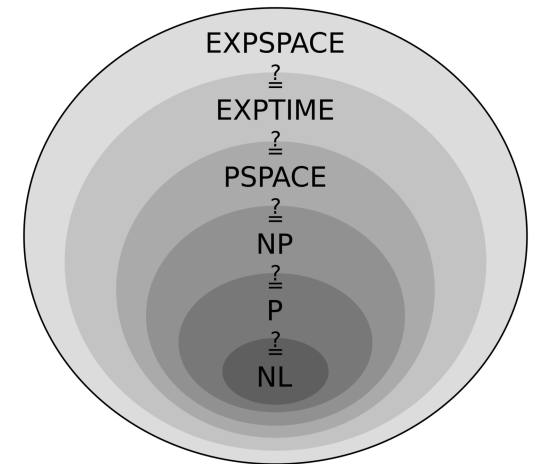
- NP-hard: at least as hard as all of the problems in NP
- NP-complete: NP-hard and in NP

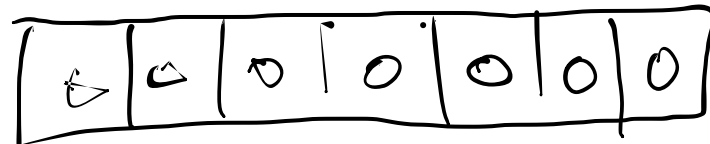




# PSPACE Complexity Class

- PSPACE: Set of all problems that can be solved using polynomial *space*
- Observation 1:  $P \subseteq PSPACE$
- Observation 2: There are algorithms that use exponential time, but only polynomial space
- Observation 3: There is an algorithm that solves 3-SAT using polynomial space, therefore  $NP \subseteq PSPACE$
- $NP \subset PSPACE$  ???





0 0 0 0 0 0 1

0 0 0 0 1 0

# PSPACE-Complete Problems

QSAT: Let  $\Phi(x_1, \dots, x_n)$  be a 3-SAT expression with odd  $n$ , then

$$\exists x_1 \forall x_2 \cdots \exists x_{n-2} \forall x_n \Phi(x_1, \dots, x_n)?$$

(for comparison 3-SAT asked  $\exists x_1 \exists x_2 \cdots \exists x_{n-2} \exists x_n \Phi(x_1, \dots, x_n)?$ )

- QSAT is PSPACE-complete (Stockmeyer and Meyer, 1973)
- If QSAT can be reduced to A, A is PSPACE-hard