- 1. What does  $M_x$  do on input 01011?
- 2. What does  $M_x$  do on input 111?
- 3. What does  $M_x$  do on an arbitrary input w?
- 4. When will  $M_x$  halt on a given input?
- 5. What language does  $M_x$  accept?



On any input string w,  $M_{\rm x}$  erases the tape and then runs  $M_{\rm b}$  on a blank tape.

If  $M_b$  halts when started on a blank tape,  $M_x$  halts on every input.

If  $M_b$  does not halt when started on a blank tape,  $M_x$  does not halt on any inputs.

So the language  $M_x$  accepts is either  $\Sigma^*$  or  $\Phi$ .



Suppose we have an algorithm for this question:

(c) Given  $M_c$ : Is there any string on which  $M_c$  halts?

Use  $M_x$  as input.

If the answer is "yes" then  $M_b$  halts when started on a blank tape.

If the answer is "no" then  $M_b$  does not

halt when started on a blank tape.



Suppose we have an algorithm for this question:

**M**~:

(d) Given  $M_d$ : Does  $M_d$  halt on every string?

Use  $M_x$  as input.

If the answer is "yes" then  $M_b$  halts when started on a blank tape.

If the answer is "no" then  $M_b$  does not

halt when started on a blank tape.



But last class we showed problem (b) was not decidable. Therefore (c) and (d) are also not decidable. Try to prove (e) is not decidable.

Theorem 5.4.2 (p. 255): The following problems are not Turing-decidable:

- (a) Given  $M_a$ , w: Does  $M_a$  halt on input w?
- (b) Given  $M_b$ : Does  $M_b$  halt on input  $\epsilon$ ?
- (c) Given  $M_c$ : Is there any string on which  $M_c$  halts?
- (d) Given  $M_d$ : Does  $M_d$  halt on every string?

(e) Given two TM's  $M_1$  and  $M_2$ : Do they halt on the same input strings?

1. Which of these sequences correspond to Hamilton cycles in the graph?

(a) 0 1 3 5 4 2

- (b) 0 1 2 4 3 5
- (c) 0 1 4 3 1 2
- (d) 0 2 3 5 4 1



(e) 0 1 3 5 6 2

2. Give pseudocode for an algorithm to check if a seqence S[0..(n-1)] for a graph G stored in an adjacency matrix A[0..(n-1)][0..(n-1)] gives a Hamilton cycle.

#### Announcements:

There is a tutorial today. Assignment #5 is due on Friday at the beginning of class.

Please do course evaluations.

Office hours for final exam:

Tuesday August 1: 12:30pm Thursday August 3: 12:30pm Friday August 4: 12:30pm

Sunday August 13: 3pm-6pm, Room TBA

Introduction to NP-completeness.

The class of NP-complete problems is defined.

Satisfiability, the most famous NP-complete problem is introduced.

To prove new problems are undecidable, we use halting problem reductions.

To prove new problems are NP-complete, reductions are also used but the transformation must preserve polynomial running time complexity.

# Table 1: Comparing polynomial and exponential time complexity.Assume a problem of size one takes 0.000001 seconds (1 microsecond).

	Size <i>n</i>							
	10	20	30	40	50	60		
n	0.00001 second	0.00002 second	0.00003 second	0.00004 second	0.00005 second	0.00006 second		
n²	0.0001 second	0.0004 second	0.0009 second	0.0016 second	0.0025 second	0.0036 second		
n <sup>3</sup>	0.001 second	0.008 second	0.027 second	0.064 second	0.125 second	0.216 second		
n <sup>5</sup>	0.1 second	3.2 second	24.3 second	1.7 minutes	5.2 minutes	13.2 minutes		
2 <sup>n</sup>	0.001 second	1.0 second	17.9 minutes	12.7 days	35.7 years	366 centuries		
3 <sup>n</sup>	0.059 second	58 minutes	6.5 years	3855 centuries	2*10 <sup>8</sup> centuries	1.3*10 <sup>13</sup> centuries		

(from M. R. Garey and D. S. Johnson, *Computers and Intractability: A Guide to the Theory of NP-completeness*, W. H. Freeman, New York, 1979.)

Table 2: Effect of improved technology on several polynomial andexponential time algorithms. The following table represents the size of thelargest problem instance solvable in 1 hour.

Time Complexity function	With present computer	With computer 100 times faster	With computer 1000 times faster
n	N1	100 N1	1000 N1
n²	N2	10 N2	31.6 N2
n <sup>3</sup>	N3	4.46 N3	10 N3
n <sup>5</sup>	N4	2.5 N4	3.98 N4
2 <sup>n</sup>	N5	N5+6.64	N5+9.97
3 <sup>n</sup>	N6	N6+4.19	N6+6.29

(from M. R. Garey and D. S. Johnson, *Computers and Intractability: A Guide to the Theory of NP-completeness*, W. H. Freeman, New York, 1979.)

#### Class P

A decision problem (yes/no question) is in the class P if there is a polynomial time algorithm for solving it.

Polynomial time:  $O(n^c)$  for some constant c.

- If a problem is solvable in polynomial time for
- some sensible encoding of the input
- •some reasonable machine (TM/RAM/PC)

it can be solved in polynomial time for all other sensible encodings/reasonable machines.

A program that runs on a Random Access Machine (such as our modern day computers) in time T(n) can be simulated on a single tape Turing machine in time  $O(T^{3}(n))$ .

This distinction is made between polynomial time and anything else (sometimes referred to as exponential time) because of the blowup in computation times which appear for exponential algorithms.

### Class NP

A decision problem (yes/no question) is in the class NP if it has a nondeterministic polynomial time algorithm. Informally, such an algorithm:

1. Guesses a solution (nondeterministically).

2. Checks deterministically in polynomial time that the answer is correct.

Or equivalently, when the answer is "yes", there is a certificate (a solution meeting the criteria) that can be verified in polynomial time (deterministically). Example problem which is in P and NP:

Minimum Weight Spanning Tree (CSC 225).

Input: Graph G, integer k.

Question: Does G have a spanning tree of weight at most k?

If you are provided with a tree with weight at most k as part of the solution, the answer can be verified in O(n) time.



Matching is in NP: Given a graph G and integer k, does G have a k-edge matching?

Matching: disjoint edges.

Certificate, k=5: (a,f) (b,g) (c,h) (d,i)(e,j)



Hamilton Cycle is in NP: Input graph G.



Certificate:

- 1, 2, 3, 5,
- 6, 7, 11, 12,
- 10, 8, 9, 4



Picture from: http://mathoverflow.net/faq

Does P= NP? the Clay Mathematics Institute has offered a \$1 million US prize for the first correct proof.

Some problems in NP not known to be in P:

- Hamilton Path/Cycle
- Independent Set
- Satisfiability

Note: Matching is in P. Learn more in a graph algorithms class.

# NP-completeness



I guess I'm just too dumb.





I can't find an efficient algorithm, but neither can all these famous people.

## NP-complete Problems

The class of problems in NP which are the "hardest" are called the NP-complete problems.

A problem Q in NP is NP-complete if the existence of a polynomial time algorithm for Q implies the existence of a polynomial time algorithm for all problems in NP.

Steve Cook in 1971 proved that SAT is NPcomplete. Proof: will be given in our last class. Other problems: use reductions. Bible for NPcompleteness:

M. R. Garey and D. S. Johnson, Computers and Intractability: A Guide to the Theory of NP-Completness, W. H. Freeman, 1st ed. (1979).

COMPUTERS AND INTRACTABILITY A Guide to the Theory of NP-Completeness Michael R. Garey / David S. Johnson

# SAT (Satisfiability)

**Variables:**  $u_1, u_2, u_3, ..., u_k$ .

A literal is a variable  $u_i$  or the negation of a variable  $\neg u_i$ .

If u is set to true then  $\neg$  u is false and if u is set to false then  $\neg$  u is true.

A clause is a set of literals. A clause is *true* if at least one of the literals in the clause is *true*.

The input to SAT is a collection of clauses.

This SAT problem has solution  $u_1=T$ ,  $u_2=F$ ,  $u_3=T$ ,  $u_4=F$ 

 $(u_1 \text{ OR } u_2 \text{ OR } u_4) \text{ AND } (\neg u_2 \text{ OR } u_4) \text{ AND}$  $(\neg u_1 \text{ OR } u_3) \text{ AND } (\neg u_4 \text{ OR } \neg u_1)$ 

Does this SAT problem have a solution?

 $(u_1 OR u_2) AND (\neg u_2 OR u_3) AND$ 

 $(\neg u_3 OR \neg u_1)$  AND  $(\neg u_2 OR \neg u_3)$  AND

$$(u_3 OR - u_1)$$

## SAT (Satisfiability)

The output is the answer to: Is there an assignment of *true/false* to the variables so that every clause is satisfied (satisfied means the clause is *true*)?

If the answer is yes, such an assignment of the variables is called a **truth** assignment.

SAT is in NP: Certificate is true/false value for each variable in satisfying assignment.



History of NP-completeness Reductions

3-SAT- each clause must contain exactly 3 variables (assignment- at most 3). Given: SAT is NP-complete (proof later) Theorem: 3-SAT is NP-Complete. The first step in any NP-completeness proof is to argue that the problem is in NP. The problem 3-SAT is a yes/no question. Certificate: truth assignment, can be checked in polynomial time. Next, we show that a polynomial time algorithm for 3-SAT implies the existence of one for SAT.

To convert a SAT problem to 3-SAT:

1. Clauses of size 1.

SAT:  $\{z\}$ 3-SAT:  $\{ z, y_1, y_2 \}, \\ \{ z, \neg y_1, y_2 \}, \\ \{ z, \gamma_1, \neg \gamma_2 \}, \\ \{ z, \gamma_1, \neg \gamma_2 \},$  $\{z, \neg y_1, \neg y_2\}$ 

 $y_1$  and  $y_2$  are new variables.

2. Clauses of size 2.

SAT: 
$$\{z_1, z_2\}$$

### 3. Clauses of size 3.

Leave these as they are since they are already acceptable for 3-SAT.

### 4. Clauses of size 4 or more.

3-SAT: { $z_1$ ,  $z_2$ ,  $y_1$ }, { $\neg y_1$ ,  $z_3$ ,  $y_2$ }, { $\neg y_2$ ,  $z_4$ ,  $y_3$ },

. . .

$$\{\neg y_{k-4}, z_{k-2}, y_{k-3}\}, \\ \{\neg y_{k-3}, z_{k-1}, z_{k}\}$$

 $y_1, y_2, \dots y_{k-3}$ , are new variables.

This does not constitute a proof of NPcompleteness unless we can argue that the size of the new 3-SAT problem problem is polynomially bounded by the size of the old SAT problem. Consider each case:

Size of clause	# new literals	size before	size after
1	2	1	12
2	1	2	6
3	0	3	3
k ≥ 4	k-3	k	k + 2(k-3)

In all cases, the size after is at most 12 times the original problem size.

2-SAT: All clauses have at most 2 literals.

There is a linear time algorithm for 2-SAT so 2-SAT is in P.

The 3-SAT problem is as hard as SAT but unless P=NP, 2-SAT is easier than 3-SAT or SAT.