# Algorithms and Data Structures

Week 13 - Lecture A

## Computational Complexity

- All problems can be classified into one of a number of classes, depending on how difficult they are to solve.
- These classes include:
  - P the set of all problems which can be solved in *Polynomial* time;
  - EXP the set of all problems which can be solved in *Exponential* time;
  - R the set of all problems which can be solved in *Finite* time.
- Beyond R there are still problems, these are *unsolvable*.
  - There are more of these than in any other class. ☺
- Almost all of the problems we have looked at in this subject are members of P.

# Some Sample Problems

- Negative-Weight Cycle Detection:
  - ∈ P.
- *n* × *n* Chess:
  - ∈ EXP;
  - ∉ P.
- Tetris:
  - ∈ EXP;
  - We don't know if it is in P.
- Halting Problem:
  - ∉ R.

### Most Decision Problems ∉ R

- Decision problems are ones with a Yes/No answer.
- Program
  - ⇒ Binary string
  - ⇒ Natural number.
- So there are no more programs than there are integers.
- Decision Problem
  - $\Rightarrow$  function: inputs  $\rightarrow$  {Yes, No}
  - $\Rightarrow$  function: binary string  $\rightarrow$  {0,1}
  - $\Rightarrow$  function: natural number  $\rightarrow$  {0,1}.

## Most Decision Problems ∉ R

• We can tabulate any decision problem for each of its possible inputs...

	Input	1	2	3	4	5	6	7	8	
• We (	Yes/No	0	1	0	0	1	1	0	1	

- We described any desired problem in terms of the number sequence of bits in the Yes/No row.
- If we express it as a binary fraction e.g. .01001101... we can equate each decision problem to a real number between 0 and 1.

## Most Decision Problems ∉ R

- So:
  - Every program  $\in \mathcal{Z}$ ;
    - $\mathcal{Z}$  is *countably* infinite.
  - Every decision problem  $\in \Re$ .
    - $\Re$  is *uncountably* infinite.
- $|\Re| >> |\mathcal{Z}|$ .
  - There are far more real numbers than integers;
  - There are far more decision problems than programs;
  - Almost every problem is non-computable;
  - Almost every problem is unsolvable by any program.
- Depressed?

#### One More Class?

- NP the set of all problems solvable in polynomial time using a "lucky" algorithm.
  - Lucky = always makes the right guess.
- Nondeterministic model:
  - Algorithm makes guesses;
  - Guess leads to Yes or No;
  - Guesses will always lead to Yes if there is at least one solution in the decision tree;
  - This is a sort of "magic greedy algorithm".

#### **Tetris**

- If we play a finite game of Tetris we can make guesses for each piece:
  - Where to drop it;
  - How to rotate it;
  - Whether to make last-minute adjustments.
- If we have a "lucky" algorithm for Tetris we can answer the question "Can I survive?" in a number of guesses which is polynomial in the number of pieces.
- This means that Tetris  $\in$  NP.

#### NP – an Alternative View

- NP the set of all problems whose "solutions" can be "checked" in polynomial time.
  - Whenever the answer is Yes there exists a proof which can be verified in polynomial time.
- It is usually easier to check a solution than to produce it.
  - E.g. Is 54727067 a composite number? hard to test.
  - The factors of 54727067 are 6701 and 8167. easy to verify.
- It should be obvious that every problem in P is also in NP.
  - $P \subseteq NP$
  - P = NP ? (Probably not.)

## Tetris Again

- If P≠NP we can prove that some problems, including Tetris, are in NP-P.
- We do this by demonstrating that such problems are as hard as possible while still being in NP.
- We can demonstrate that Tetris is NP-hard;
  - At least as hard as every problem in NP.
- In fact, Tetris is NP-complete.
  - NP-hard and in NP.

# Reduction: Defining "As Hard As"

- Given a problem, A, that you want to solve.
- If we can convert it into some other problem, B, then we can solve B and use this solution to solve A.
- This process is called *reduction*.
- We say the problem A is reducible to problem B.
- Note: this may not be the best way to solve problem A.
- If A is reducible to B we can say that B is at least as hard as A.

## An Example of Reduction

- A:
  - Given a weighted graph, G, find the path between S and D which minimizes the *product* of the path weights.
- B:
  - Given a weighted graph, G', find the path between S and D which minimizes the *sum* of the path weights.
- Reduction A→B:
  - Replace the weights in G with their logarithms.

## Proving a Problem is NP-complete

- To prove a problem is NP-complete we need only to prove:
  - The problem is in NP;
  - Some other problem that is already known to be NP-complete is reducible to the new problem.
- This avoids the issue of finding the first NP-complete problem.
- Stephen Cook did this in 1971:
  - Formula Satisfiability (Sat).
- Richard Karp extended Cook's work:
  - In fact, he showed that 21 different problems were all NP-complete.

## Some other NP-complete Problems

- The knapsack problem:
  - Given n elements,  $\{w_1, ..., w_n\}$  and a target W, is there a subset of elements which adds up exactly to W?
- The Hamiltonian cycle problem:
  - Given a graph, G=(V, E), is there a sequence of distinct edges e ∈ E forming a closed path such that each vertex v ∈ V is visited exactly once?
- The three colour problem.
  - Given a graph, G=(V, E), can colours be assigned to its vertices such that for any two vertices  $v_1$ ,  $v_2$  such that:
    - if  $(v_1, v_2) \in E$  then  $colour(v_1) \neq colour(v_2)$  using no more than 3 colours.

## Some other NP-complete Problems

- The travelling salesman problem:
  - Given a weighted graph, G=(V, E, W), is there a sequence of distinct edges  $e \in E$  forming a closed path such that each vertex  $v \in V$  is visited exactly once and the sum of the weights of the traversed edges is minimized?
- Satisfiability:
  - Given a set of Boolean variables  $b_1$ ,  $b_2$ , ...,  $b_n$  and the operators AND, OR and NOT can values be assigned to each variable such that a well formed expression involving only these components such that the expression evaluates to true?
  - A version of this was the first problem shown to be NP-complete.

#### Who Wants to be a Millionaire?

- You can win \$1,000,000.
- All you have to do is either:
  - Find a polynomial time algorithm to solve any NP-Complete problem, (your choice of problem), or;
  - Prove that at least one NP-complete problem has no polynomial time solution algorithm.
- Note: "I can't find one" ≠ "There isn't one"