

# Lecture 19: Introduction to NP-Completeness

**Steven Skiena**

Department of Computer Science  
State University of New York  
Stony Brook, NY 11794-4400

<http://www.cs.stonybrook.edu/~skiena>

# Topic: Introduction to NP-Completeness

---

# Reporting to the Boss

---

Suppose you fail to find a fast algorithm. What can you tell your boss?

- “I guess I’m too dumb...” (dangerous confession)
- “There is no fast algorithm!” (lower bound proof)
- “I can’t solve it, but no one else in the world can, either...” (NP-completeness reduction)

# The Theory of NP-Completeness

---

Several times this semester we have encountered problems for which we couldn't find efficient algorithms, such as the traveling salesman problem.

We also couldn't prove exponential-time lower bounds for these problems.

The theory of NP-completeness, developed by Stephen Cook and Richard Karp, provides the tools to show that all of these problems were really the same problem.

## The Main Idea

---

Suppose I gave you the following algorithm to solve the *bandersnatch* problem:

Bandersnatch( $G$ )

    Convert  $G$  to an instance of the Bo-billy problem  $Y$ .

    Call the subroutine Bo-billy on  $Y$  to solve this instance.

    Return the answer of Bo-billy( $Y$ ) as the answer to  $G$ .

Such a translation from instances of one type of problem to instances of another type such that answers are preserved is called a *reduction*.

## What Does this Imply?

---

Now suppose my reduction translates  $G$  to  $Y$  in  $O(P(n))$ :

1. If my Bo-billy subroutine ran in  $O(P'(n))$  I can solve the Bandersnatch problem in  $O(P(n) + P'(n'))$
2. If I know that  $\Omega(P'(n))$  is a lower-bound to compute Bandersnatch, then  $\Omega(P'(n) - P(n'))$  must be a lower-bound to compute Bo-billy.

The second argument is the idea we use to prove problems hard!

# My Most Profound Tweet

---

An NP-completeness proof ensures that a dumb algorithm that is slow isn't a slow algorithm that is dumb.

**Questions?**

# Topic: Problems and Reductions

---

# What is a Problem?

---

A *problem* is a general question, with parameters for the input and conditions on what is a satisfactory answer or solution.

**Example:** The Traveling Salesman

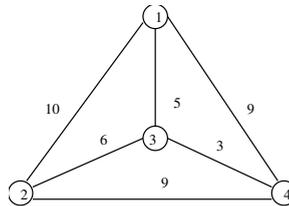
**Problem:** Given a weighted graph  $G$ , what tour  $\{v_1, v_2, \dots, v_n\}$  minimizes  $\sum_{i=1}^{n-1} d[v_i, v_{i+1}] + d[v_n, v_1]$ .

## What is an Instance?

---

An instance is a problem with the input parameters specified.

TSP instance:  $d[v_1, v_2] = 10$ ,  $d[v_1, v_3] = 5$ ,  $d[v_1, v_4] = 9$ ,  
 $d[v_2, v_3] = 6$ ,  $d[v_2, v_4] = 9$ ,  $d[v_3, v_4] = 3$



Solution:  $\{v_1, v_2, v_3, v_4\}$  cost= 27

# Decision Problems

---

A problem with answers restricted to *yes* and *no* is called a *decision problem*.

Most interesting optimization problems can be phrased as decision problems which capture the essence of the computation.

For convenience, from now on we will talk *only* about decision problems.

# The Traveling Salesman Decision Problem

---

Given a weighted graph  $G$  and integer  $k$ , does there exist a traveling salesman tour with cost  $\leq k$ ?

Using binary search and the decision version of the problem we can find the optimal TSP solution.

# Reductions

---

Reducing (transforming) one algorithm problem  $A$  to another problem  $B$  is an argument that if you can figure out how to solve  $B$  then you can solve  $A$ .

We showed that many algorithm problems are reducible to sorting (e.g. element uniqueness, mode, etc.).

A computer scientist and an engineer wanted some tea. . .

**Questions?**