# Introduction to Algorithms

6.046J/18.401J



Lecture 25

Prof. Piotr Indyk



### 🤝 Final Exam

- May 19, 2008, 9:00am 12 pm
- · Johnson Ice Rink
- Closed book:
  - two handwritten crib sheets
- · Coverage: everything except
  - L17 Hidden Markov Models II
  - L18 Computational Biology
  - L26 Parallel Algorithms
- Quiz review: Fri, 3-5pm, 32-155

© Piotr Indyk

Introduction to Algorithms

May 13, 2008 L25.2



# 🔙 Dealing with Hard Problems

- · What to do if:
  - Divide and conquer
  - Dynamic programming
  - Greedy
  - Linear Programming/Network Flows

does not give a polynomial time algorithm?

© Piotr Indvk

Introduction to Algorithms

May 13, 2008 L25,3



## Dealing with Hard Problems

- Idea I: Ignore the problem
  - Can't do it! There are thousands of problems for which we do not know polynomial time algorithms
  - For example:
    - Traveling Salesman Problem (TSP)
    - Set Cover

© Piotr Indvk

Introduction to Algorithms

May 13, 2008 L25.4



### **Traveling Salesman Problem**

- · Traveling Salesman Problem (TSP)
  - Input: undirected graph with lengths on edges
  - Output: shortest cycle that visits each vertex exactly
- Best known algorithm: Ω(2<sup>n</sup>) time



May 13, 2008 L25.5

# **Set Covering**

- Set Cover:
  - Input: subsets  $S_1 ... S_n$  of X,  $\bigcup_i S_i = X, |X| = m$
  - Output:  $C\subseteq \{1\dots n\}$  , such that  $\bigcup_{i\in C} S_i = X,$  and |C| minimal
- · Vertex cover: special case
  - -X = edges
  - $-S_v =$ edges incident to vertex v
- Best known algorithm:  $\Omega(2^n)$  time

App: bank robbery

- X={plan, shoot, safe, drive, scary}
- - S<sub>Steve</sub> = {plan, safe}
  - S<sub>Stevie</sub>={shoot, scary, drive}
  - S<sub>Stevo</sub> = {plan, drive}

© Piotr Indvk

Introduction to Algorithms

May 13, 2008 L25.6



### Dealing with Hard Problems, ctd.

- Exponential time algorithms for small inputs
  - E.g., 1.274<sup>n</sup> time is not bad for n < 50 (such algorithm exists for Vertex Cover)
- Polynomial time algorithms for some inputs (e.g., average-case)
- Polynomial time algorithms for all inputs, but which return approximate solutions

© Piotr Indyk

Introduction to Algorithms

May 13, 2008 L25.7



# 🥽 Approximation Algorithms

- An algorithm A is  $\rho$ -approximate, if, on any input of size n:
  - The cost  $C_{\Lambda}$  of the solution produced by the algorithm, and
  - The cost  $C_{OPT}$  of the optimal solution are such that  $C_A \leq \rho \ C_{OPT}$
- · We have seen:
  - 2-approximation algorithm for finding a median string (PS7)
- We will see:
  - 2-approximation algorithm for TSP in the plane
- ln(m)-approximation algorithm for Set Cover

Introduction to Algorithms

May 13, 2008 L2



# 🥽 Comments on Approximation

- "  $C_{A} \leq \rho$   $C_{OPT}$  " makes sense only for minimization problems
- For maximization problems, replace by " $C_A \ge 1/\rho \ C_{OPT}$ "
- Additive approximation "C<sub>A</sub>≤ ρ + C<sub>OPT</sub>" also makes sense, although difficult to achieve

© Piotr Indyk

Introduction to Algorithms

May 13, 2008 L25.9



### **TSP**

© Piotr Indyk

Introduction to Algorithms

May 13, 2008 L25.10



### 2-approximation for TSP

- Compute MST T
  - An edge between any pair of points
  - Weight = distance between endpoints
- Compute a tree-walk W of T
  - Each edge visited twice
- Convert W into a cycle C using shortcuts



© Piotr Indy

Introduction to Algorithm:

May 13, 2008 L25.11



### 2-approximation: Proof

- Let C<sub>OPT</sub> be the optimal cycle
- $Cost(T) \le Cost(C_{OPT})$ 
  - Removing an edge from C gives a spanning tree, T is a spanning tree of minimum cost
- Cost(W) = 2 Cost(T)
  - Each edge visited twice
- $Cost(C) \le Cost(W)$ 
  - Triangle inequality
- $\Rightarrow$  Cost(C)  $\leq$  2 Cost(C<sub>OPT</sub>)

© Piotr Indyk

Introduction to Algorithms

May 13, 2008 L25.12



### **Set Cover**

© Piotr Indyk

Introduction to Algorithms

May 13, 2008 L25.13



### 🔼 Approximation for Set Cover

• Input: subsets  $S_1...S_n$  of X,  $\bigcup_i S_i = X$ , |X| = m• Output:  $C \subseteq \{1...n\}$ , such that  $\bigcup_{i \in C} S_i = X$ , and |C| minimal

### Greedy algorithm:

- Initialize C=∅
- · Repeat until all elements are covered:
  - Choose S<sub>i</sub> which contains largest number of yet-not-covered elements
  - Add i to C
  - Mark all elements in S; as covered

© Piotr Indyk

Introduction to Algorithms

May 13, 2008 L25.14



## Greedy Algorithm: Example

- $X=\{1,2,3,4,5,6\}$
- · Sets:
  - $-S_1=\{1,2\}$
  - $-S_2=\{3,4\}$
  - $-S_3=\{5,6\}$
  - $-S_4=\{1,3,5\}$
- Algorithm picks C=all sets
- · Not optimal!

© Piotr Indyk

Introduction to Algorithms

May 13, 2008 L25.15



### ln(m)-approximation

- · Notation:
  - C<sub>OPT</sub> = optimal cover
  - $k = |C_{OPT}|$  (we do not know k)
- Fact: At any iteration of the algorithm, there exists S, which contains at least 1/k fraction of yet-not-covered elements
- Proof:
  - $C_{OPT}$  covers the (uncovered) elements using k sets
  - One of those sets must cover ≥1/k fraction of yet-not-covered elements
- Conclusion: greedy algorithm covers ≥1/k fraction of yetnot-covered elements in each step

© Piotr Indyk

Introduction to Algorithms

May 13, 2008 L25.16



### In(m)-approximation

- Let  $u_i$  be the number of yet-not-covered elements at the end of step  $i=0,1,2,\ldots$
- · We have

$$u_{i+1} \leq u_i (1-1/k) u_0 = m$$

- Therefore, after t=k ln m steps, we have  $u_t \le u_0 \, (1 1/k)^t \le m \, (1 1/k)^{k \, \ln m} < m \, 1/e^{\ln m} = 1$
- I.e., all elements are covered by the k ln m sets chosen by greedy algorithm
- Opt size is  $k \Rightarrow$  greedy is ln(m)-approximate

Piotr Indyl

Introduction to Algorithm.

May 13, 2008 L25.17



### Approximation Algorithms

- · Very rich area
  - Algorithms use greedy, linear programming, dynamic programming
    - E.g., 1.01-approximate TSP in the plane
  - Sometimes can show that approximating a problem is as hard as finding exact solution!
    - E.g., 0.99 ln(m)-approximate Set Cover

© Piotr Indyk

ntroduction to Algorithms

May 13, 2008 L25.18