

CSE 421

Introduction to Algorithms

Lecture 25

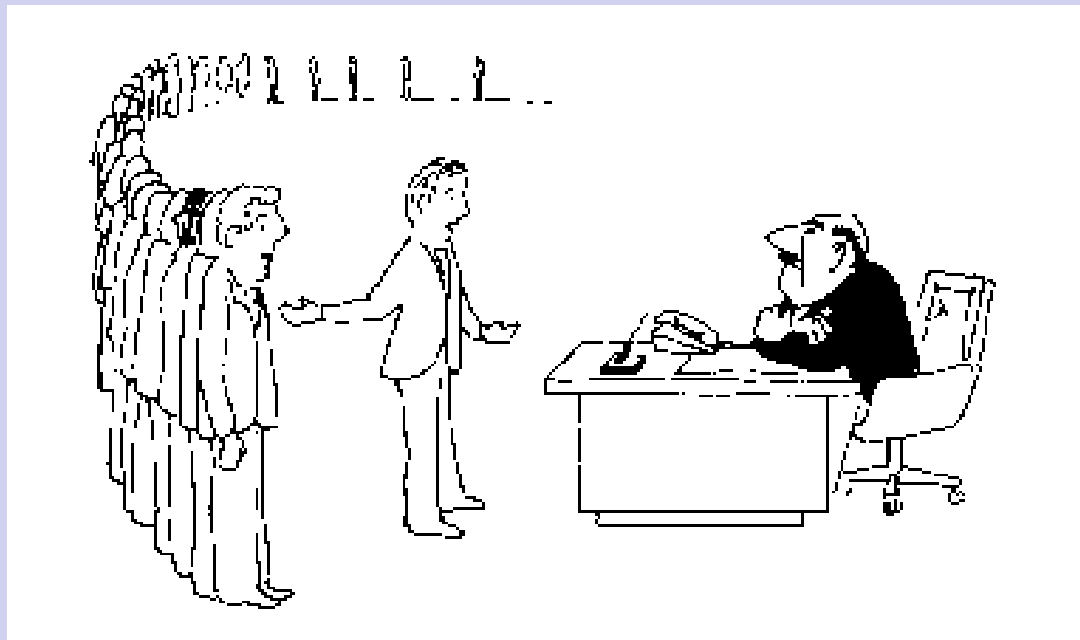
Coping with NP-Completeness

Announcements

- Today, Coping with NP-Completeness
 - Chapters 11 and 12
- Friday, Beyond NP-Completeness
 - Section 8.9, Chapter 9
- Homework 9, Due Friday, March 8
- Final exam,
 - Monday, March 11, 2:30-4:20 pm PDT
 - Comprehensive (~60% post midterm, ~40% pre midterm)
 - Old finals / answers on home page

Coping with NP-Completeness

- Approximation Algorithms
- Exact solution via Branch and Bound
- Local Search



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

Approximation Algorithms

- K-Approximation Algorithm
- Worst case ratio of solution and optimal as input size goes to infinity
- Minimization problems
 - Find a solution at most K times the optimum
- Maximization problems
 - Find a solution at most $1/K$ times the optimum

Vertex Cover

- A vertex cover is a subset of the vertices that is adjacent to every edge
- VC is NP-Complete

```
W = {};
```

```
E' = E
```

```
while E' is not empty
```

```
    Select e = (u,v) from E'
```

```
    Add u and v to W
```

```
    Remove all edges adjacent to u or v from E'
```

$W = \{\};$

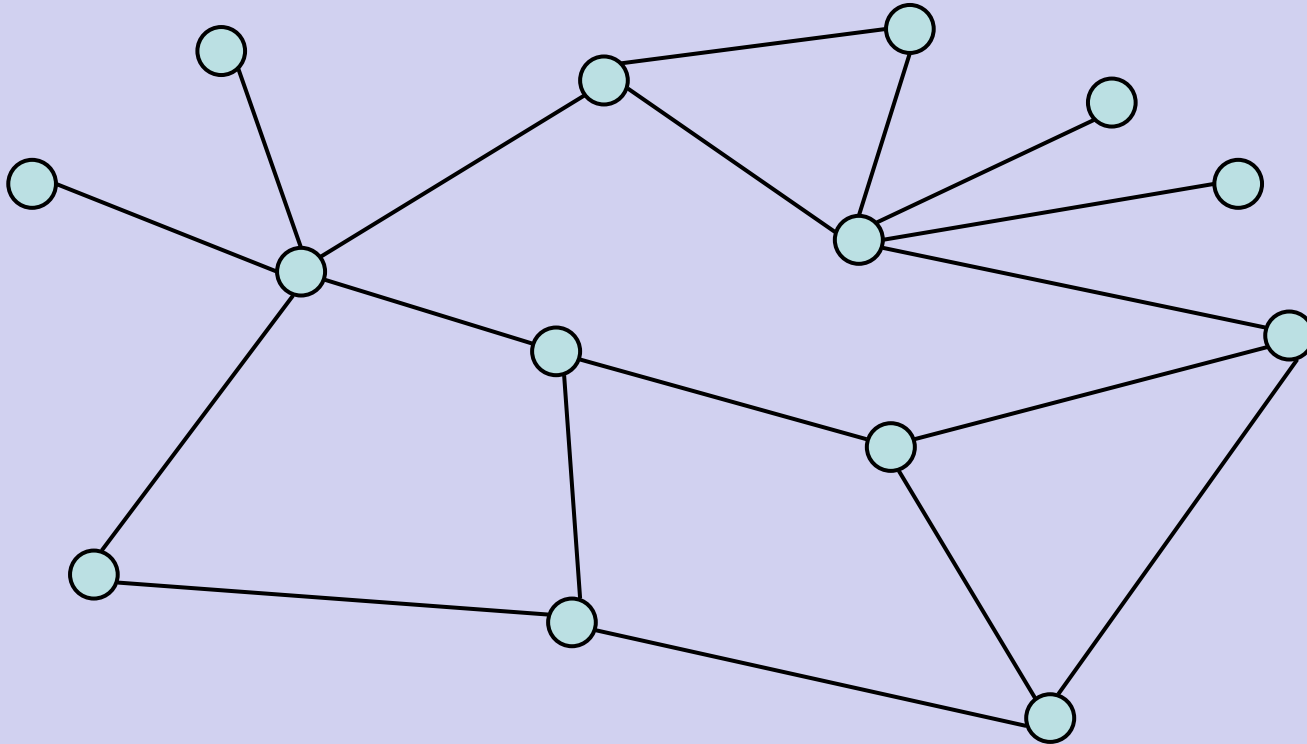
$E' = E$

while E' is not empty

 Select $e = (u, v)$ from E'

 Add u and v to W

 Remove all edges adjacent to u or v from E'



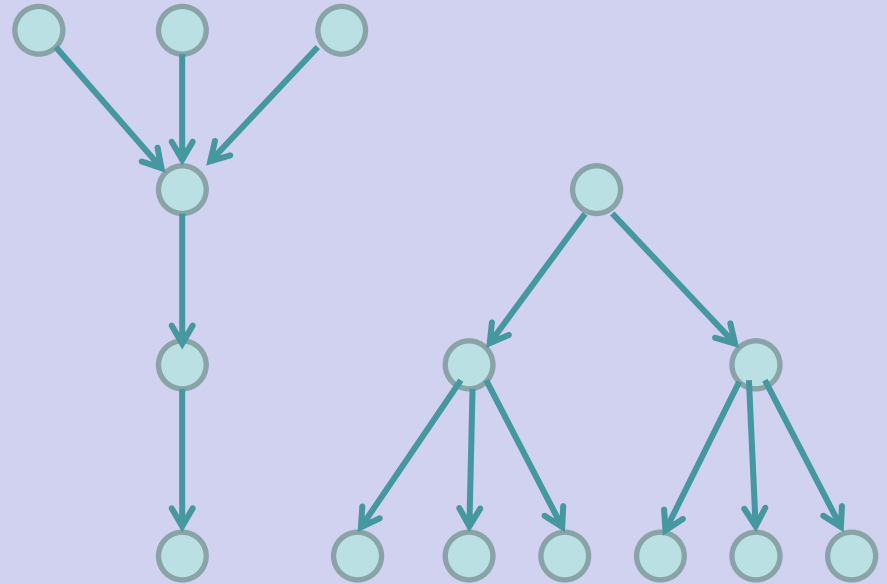
VC 2-Opt Bound

- When edge $e = (u,v)$ is selected, neither u nor v is in W
- At least one of u or v must be in the VC to cover e
- Thus, at least $\frac{1}{2}$ the vertices placed in W are necessary

Multiprocessor Scheduling

- Unit execution tasks
- Precedence graph
- K-Processors

- Polynomial time for $k=2$
- Open for $k = \text{constant}$
- NP-complete if k is part of the problem



Highest level first is 2-Optimal

Choose k items on the highest level

Claim: number of rounds is at least twice the optimal.

Suppose the maximum height of a task is H

A **partial round** removes $< k$ elements

A **full round** removes k elements

2-Opt Proof for HLF

The number of partial rounds is at most H

$$\text{Opt} \geq H$$

The number of full rounds is at most N / k

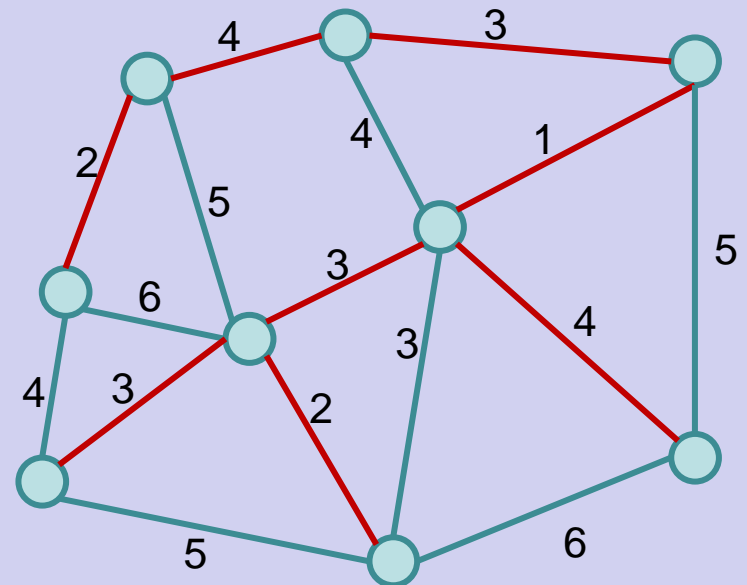
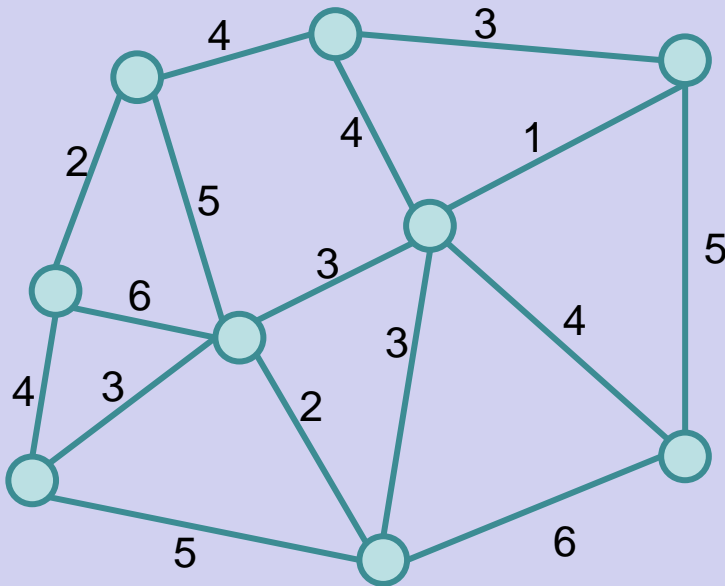
$$\text{Opt} \geq N / k$$

$$\text{Partial} + \text{Full} \leq H + N / K \leq 2 \text{Opt}$$

MST Bound for TSP

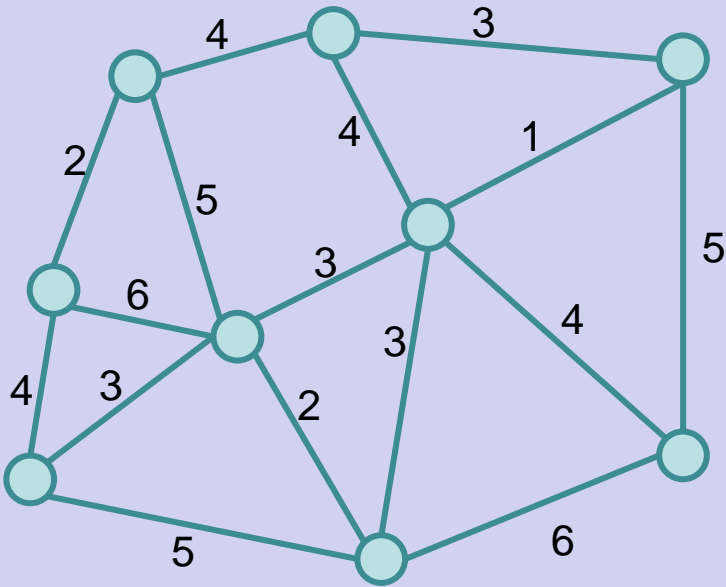
Undirected graph satisfying triangle inequality

$\text{MST Cost} \leq \text{TSP Cost} \leq 2 \text{ MST Cost}$



Christofides TSP Algorithm

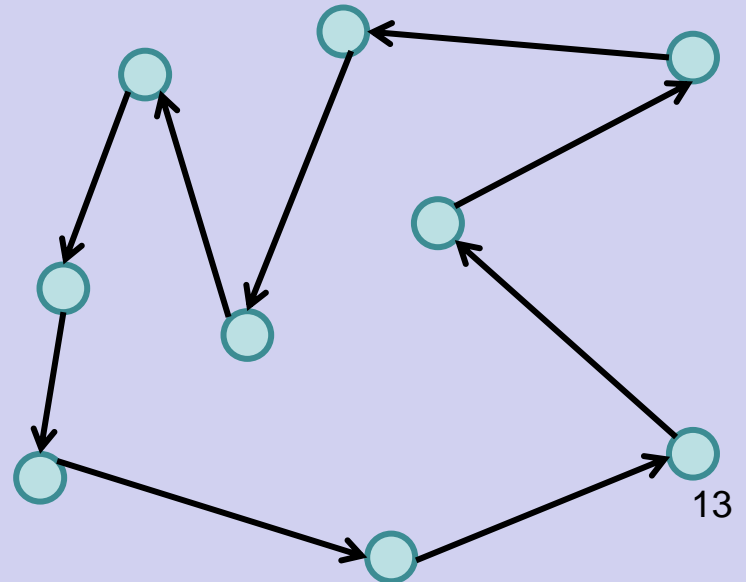
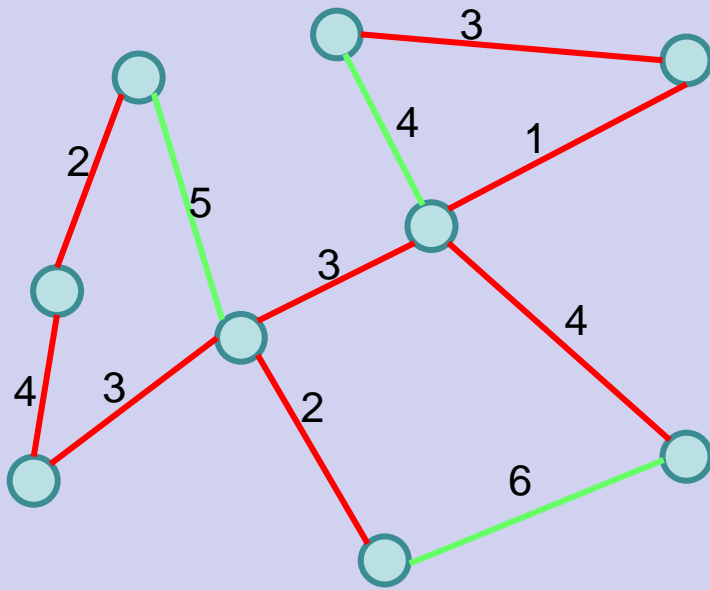
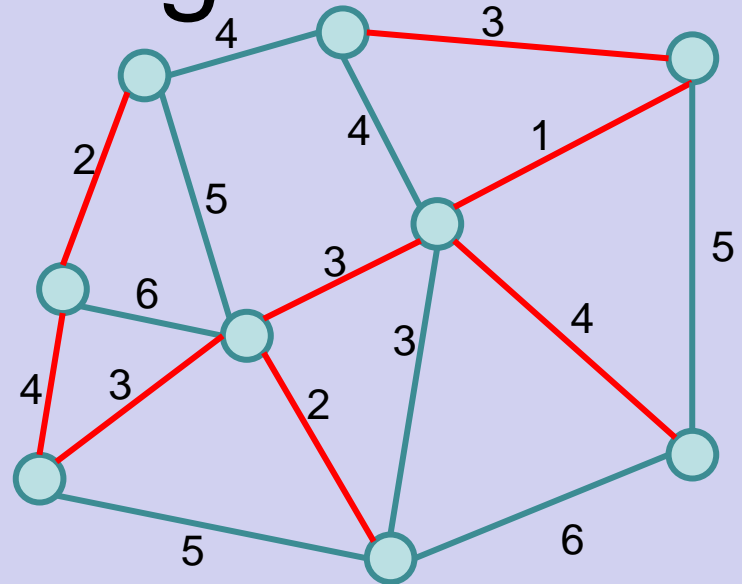
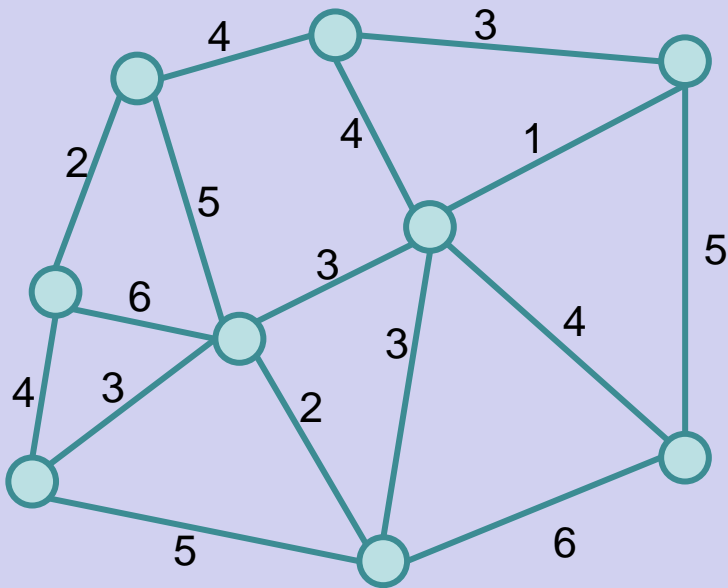
- Undirected graph satisfying triangle inequality



1. Find MST
2. Add additional edges so that all vertices have even degree
3. Build Eulerian Tour

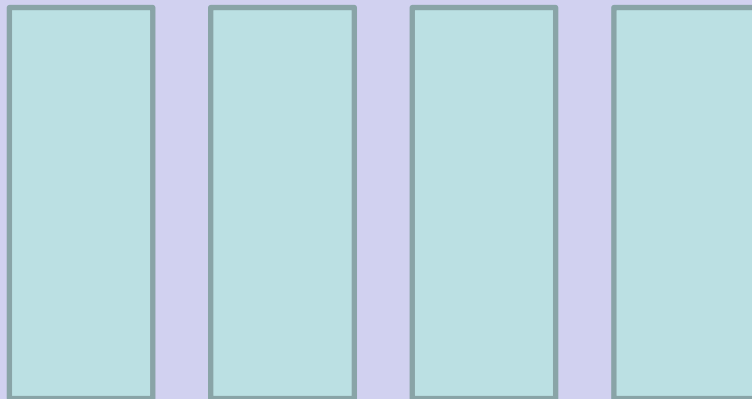
3/2 Approximation

Christofides Algorithm



Bin Packing

- Given N items with weight w_i , pack the items into as few unit capacity bins as possible
- Example: .3, .3, .3, .3, .4, .4



First Fit Packing

- First Fit
 - Theorem: $FF(I)$ is at most $17/10 \text{ Opt}(I) + 2$
- First Fit Decreasing
 - Theorem: $FFD(I)$ is at most $11/9 \text{ Opt}(I) + 4$

Knapsack

- Items $\{I_1, I_2, \dots, I_n\}$
 - Weights $\{w_1, w_2, \dots, w_n\}$, Values $\{v_1, v_2, \dots, v_n\}$
- Find set S of indices to maximize:
 - $\sum_{i \in S} v_i$ such that $\sum_{i \in S} w_i \leq K$
- Dynamic Programming solution:
 - Find the smallest set of a given value
 - Runtime $O(nV)$ where V is the sum of the values
- Goal – for any $\varepsilon > 0$, we want a polynomial time algorithm that finds a solution of at least $(1-\varepsilon) \text{Opt}$

PTAS (Polynomial time approximation scheme)

- Idea for approximation algorithm*
- Scale values so that $\frac{1}{2} \leq \text{Opt} \leq 1$
- Let $\varepsilon = 2^{-k}$
- Round the values down to multiples of ε^2
- Solve the DP using ε^2 values
- Runtime $O(n\varepsilon^2)$, Approximation $(1-\varepsilon)$

*Some details omitted in dealing with very small items.

Branch and Bound

- Brute force search – tree of all possible solutions
- Branch and bound – compute a lower bound on all possible extensions
 - Prune sub-trees that cannot be better than optimal

Branch and Bound for SAT

- Solving SAT by setting one variable at a time
- Setting a literal to 1 removes the clause
- Setting a literal to 0 removes the literal
 - Removing the last literal kills the subtree
- Heuristics for variable ordering
- Very important algorithms in practice, especially for software verification

Local Optimization

- Improve an optimization problem by local improvement
 - Neighborhood structure on solutions
 - Travelling Salesman 2-Opt (or K-Opt)
 - Independent Set Local Replacement

Enhancements to Local Search

- Randomized Local Search
 - Start from lots of places
- Metropolis Algorithm
 - Choose random neighbor
 - Move if cheaper
 - If worse, move with some probability
- Simulated Annealing
 - Like Metropolis, but adjust probabilities to simulate cooling