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## Phase Transitions

- P or NP classification based entirely on worst-case analysis
- Problem banished from P if one instance requires exponential solution time
- Overconstrained and underconstrained are easy
- Really hard problems occur on boundary between these regions
- Probability of a Hamiltonian Circuit as average connectivity varies
- Almost fully connected almost always has HC
- With connectivity of just over 2, almost never has HC

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## Graph Coloring

4 colors, connectivity increases

3 colors, connectivity increases

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## Traveling Salesman

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## Approximation Algorithms

- NP-Complete problems require exponential running times to find optimal solutions
  - If the problem instance is small, then you can wait
  - If not, a near-optimal solution in polynomial-time may be acceptable
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### Ratio Bound

How does the approximate solution compare to the optimal solution?

An approximation algorithm has a \_\_\_\_\_  $p(n)$  if for any input of size  $n$ , the cost  $C$  of the approximate solution is within a factor  $p(n)$  of the cost  $C^*$  of the optimal solution:

$$1 \leq p(n) \leq \max\left(\frac{C}{C^*}, \frac{C^*}{C}\right)$$

where  $C/C^*$  is used for minimization problems, and  $C^*/C$  for maximization problems.

Alternatively, an approximation algorithm has a \_\_\_\_\_

$$\frac{|C - C^*|}{C^*} \leq \epsilon(n)$$

where  $\epsilon(n) \leq p(n) - 1$ .

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### Approximation Scheme

An approximation \_\_\_\_\_ is an approximation algorithm that takes as input an instance of the problem and a value  $\epsilon > 0$ , such that the algorithm has a relative error bound  $\epsilon$ .

The approximation scheme is \_\_\_\_\_ if it runs in time polynomial in the size  $n$  of the input.

The approximation scheme is \_\_\_\_\_ if it runs in time polynomial in  $1/\epsilon$  and  $n$ .

$$T(n) = f(1/\epsilon, n)$$

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## Vertex-Cover Problem

Approximation algorithm:

Keep grabbing edges whose vertices are not already in the cover

Approx-Vertex-Cover( $G$ ) ;  $G = (V, E)$

$C = \{\}$

$E' = E$

while  $E' \neq \{\}$

$(u,v)$  = arbitrary edge from  $E'$

$C = C \cup \{u,v\}$

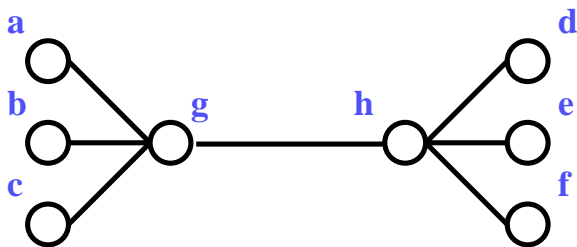
    remove from  $E'$  every edge incident on  $u$  or  $v$

return  $C$

This algorithm takes \_\_\_\_\_ time.

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## Example



$$\begin{aligned}\text{Optimal} &= \{g, h\} \\ \text{Approximate} &= \{a, g, d, h\}\end{aligned}$$

The size of the approximate vertex cover is never more than twice the size of the optimal vertex cover.

### **Theorem 37.1**

Approx-Vertex-Cover has a ratio bound of 2.

**Proof:**

The approximate solution  $C$  is a vertex cover.

Let  $A$  be the set of edges chosen by the algorithm. Since each such edge's endpoints were not in  $C$  at the time,  $|C| = 2|A|$ . An optimal cover must have at least  $|A|$  vertices,  $|C^*| \geq |A|$ . Thus  $|C^*| \geq 1/2|C|$  and  $\frac{|C|}{|C^*|} \leq 2 = p$ .

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## **Traveling Salesman Problem**

### **Triangle Inequality**

$$c(u,w) \leq c(u,v) + c(v,w)$$

(generally satisfied)

An approximation algorithm with ratio bound  $p = 2$  exists for TSPs exhibiting triangle inequality.

Approx-TSP-Tour( $G, c$ ) ;  $G = (V, E)$ ,  $c =$  edge costs

select root vertex  $r$  in  $V$

$T = \text{MST-Prim}(G, c, r)$

$L =$  list of vertices in preorder traversal of  $T$

return cycle with vertices ordered as in  $L$

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## Theorem 37.2

For TSP with triangle inequality, Approx-TSP-Tour is an approximation algorithm with a ratio bound of 2.

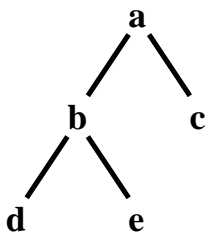
$$\text{If } H \text{ is approximate tour, } c(H) \leq 2c(H^*)$$

### Proof:

If  $H^*$  is the optimal tour and  $T$  is a  $\text{MST}(G)$ , then  $c(T) \leq c(H^*)$ .

Consider a **full walk**  $W$  of a MST with cost  $c(W)$ .

Example



$$W = a \underline{b} d \underline{b} e b a c a$$

$$c(W) = 2c(T) \longrightarrow c(W) \leq 2c(H^*)$$

$W$  is not a tour, but by triangle inequality, we can change  $w \rightarrow x \rightarrow w \rightarrow y$  to  $w \rightarrow x \rightarrow y$ , without increasing cost to yield approximate tour  $H$ .

$$c(H) \leq c(W) \leq 2c(H^*)$$

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## Theorem 37.3

If  $P \neq NP$ , there is no poly-time approximation algorithm with ratio bound  $p \geq 1$  for the general TSP (i.e., no triangle inequality).

### Proof:

If such an algorithm A exists, then we can use A to solve the Hamiltonian Cycle problem, which is NP-Complete, in polynomial time.

From graph G for Hamiltonian Cycle problem construct complete graph  $G' = (V, E')$ , where edges appearing in G have cost 1, and remaining edges have cost  $p|V| + 1$ .

If HC in G, then there is a tour of cost  $|V|$  in  $G'$ , and A must return it to satisfy its ratio bound of p. If no HC in G, the TSP tour costs at least

$$(p|V| + 1) + (|V| - 1) > p|V|.$$

Thus, can determine if HC in G based on whether TSP tour cost is  $|V|$ . But, unless  $P = NP$ , such an algorithm cannot exist, because it solves an NP-complete problem in polynomial time.

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## Set-Covering Problem

Algorithm:

- Greedy approach
- Grab the set covering the largest number of uncovered members

Greedy-Set-Cover( $X, F$ )

$U = X$  ; uncovered

$C = \{\}$

while  $U \neq \{\}$

    select  $S \in F$  maximizing  $|S \cap U|$

$U = U - S$

$C = C \cup \{S\}$

return C

## Corollary 37.5

Greedy-Set-Cover has a ratio bound of  $(\ln|x| + 1)$ .

Proof in book.

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## Subset-Sum Problem

Algorithm:

- Variation
- Return largest sum  $\leq t$  of elements in S

Exact-Subset-Sum(S, t) ; S =  $\{x_1, \dots, x_n\}$

n = |S|

$L_0 = \langle 0 \rangle$

for i = 1 to n

$L_i = \text{Merge-Lists}(L_{i-1}, L_{i-1} + x_i)$

    remove from  $L_i$  elements  $> t$

return largest element in  $L_n$

### Example

S = {1, 2, 3}

$L_0 = \langle 0 \rangle$

$L_1 = \langle 0, 1 \rangle$

$L_2 = \langle 0, 1, 2, 3 \rangle$

$L_3 = \langle 0, 1, 2, 3, 4, 5, 6 \rangle$

## Analysis

- L could double in size after each iteration

- Final merge could take  $2^n$  steps
  - Exponential running time
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## Fully Poly-Time Approximation Scheme

- Returned value is largest  $< t$  to within some percentage error
- Let  $\epsilon =$  error bound,  $0 < \epsilon < 1$
- Trim  $L$  of all values whose relative error is no more than  $\delta$  away from a value in  $L$

Trim( $L, \delta$ ) ;  $L = \langle y_1, \dots, y_m \rangle$  sorted in non-decreasing order

```

1  m = | L |
2  L' = ⟨y1⟩
3  last = y1
4  for i = 2 to m
5      if last < (1 - δ)yi
6          then append yi on end of L'
7          last = yi
8  return L'
```

Approx-Subset-Sum( $S, t, \epsilon$ )

```

1  n = |S|
2  L0 = ⟨0⟩
3  for i = 1 to n
4      Li = Merge-Lists(Li-1, Li-1 + xi)
5      Li = Trim(Li,  $\frac{\epsilon}{n}$ )
```



- 6        remove from  $L_i$  elements  $> t$
- 7        return largest value in  $L_n$

Error passed to Trim is  $\frac{\epsilon}{n}$  to prevent too much inaccuracy after repeated trimmings.

### **Theorem 37.6**

Approx-Subset-Sum is a fully poly-time approximation scheme for the subset-sum problem.

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## **Approximation Algorithms**