Today's Outline \_\_\_\_\_ CS 362, Lecture 24 • Reduction Wrapup • Approximation algorithms for NP-Hard Problems Jared Saia University of New Mexico 1 Hamiltonian Cycle \_\_\_\_\_ The Reduction \_\_\_\_\_ • To do the reduction, we need to show that we can solve • A Hamiltonian Cycle in a graph is a cycle that visits every Vertex Cover in polynomial time if we have a polynomial vertex exactly once (note that this is very different from an time solution to Hamiltonian Cycle. *Eulerian cycle* which visits every *edge* exactly once) • Given a graph G and an integer k, we will create another • The Hamiltonian Cycle problem is to determine if a given graph G' such that G' has a Hamiltonian cycle iff G has a graph G has a Hamiltonian Cycle vertex cover of size k• We will show that this problem is NP-Hard by a reduction • As for the last reduction, our transformation will consist of from the vertex cover problem. putting together several "gadgets"

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# Edge Gadget and Cover Vertices \_\_\_\_\_ Edge Gadget \_\_\_\_\_ • For each edge (u, v) in G, we have an edge gadget in G' consisting of twelve vertices and fourteen edges, as shown below • The four corner vertices (u, v, 1), (u, v, 6), (v, u, 1), and (v, u, 6)(u,v,1) (u,v,2) (u,v,3) (u,v,4) (u,v,5) (u,v,6)each have an edge leaving the gadget • A Hamiltonian cycle can only pass through an edge gadget in one of the three ways shown in the figure • These paths through the edge gadget will correspond to one or both of the vertices u and v being in the vertex cover. (v,u,1) (v,u,2) (v,u,3) (v,u,4) (v,u,5) (v,u,6)An edge gadget for (u, v) and the only possible Hamiltonian paths through it. 5 4 Cover Vertices Vertex Chains \_\_\_\_\_ • For each vertex u in G, we string together all the edge gadgets for edges (u, v) into a single vertex chain and then connect the ends of the chain to all the cover vertices • Specifically, suppose u has d neighbors $v_1, v_2, \ldots, v_d$ . Then G' • G' also contains k cover vertices, simply numbered 1 through has the following edges: k-d-1 edges between $(u, v_i, 6)$ and $(u, v_{i+1}, 1)$ (for all i between 1 and d-1) -k edges between the cover vertices and $(u, v_1, 1)$ -k edges between the cover vertices and $(u, v_d, 6)$

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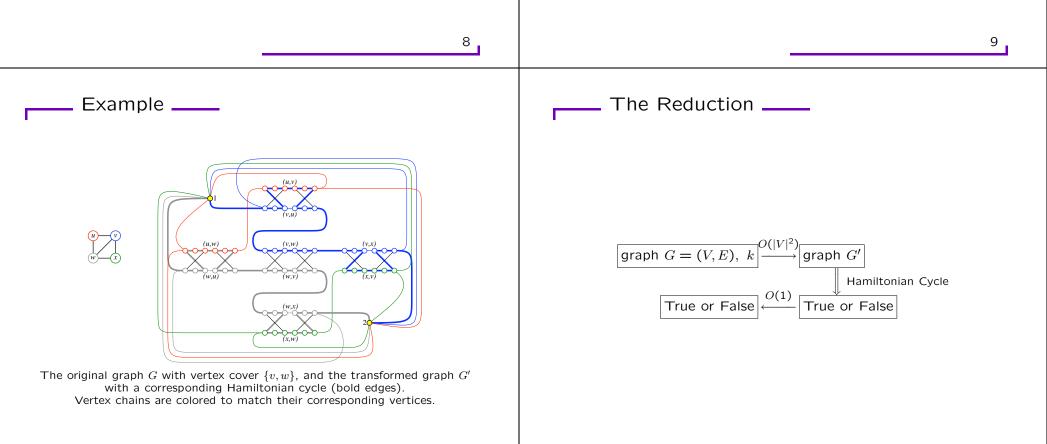
#### The Reduction \_\_\_\_\_

. The Reduction \_\_\_\_\_

- It's not hard to prove that if  $\{v_1, v_2, \ldots, v_k\}$  is a vertex cover of G, then G' has a Hamiltonian cycle
- To get this Hamiltonian cycle, we start at cover vertex 1, traverse through the vertex chain for  $v_1$ , then visit cover vertex 2, then traverse the vertex chain for  $v_2$  and so forth, until we eventually return to cover vertex 1
- Conversely, one can prove that any Hamiltonian cycle in G' alternates between cover vertices and vertex chains, and that the vertex chains correspond to the k vertices in a vertex cover of G

Thus, G has a vertex cover of size k iff G' has a Hamiltonian cycle

- The transformation from G to G' takes at most  $O(|V|^2)$  time, so the Hamiltonian cycle problem is NP-Hard
- Moreover we can easily verify a Hamiltonian cycle in linear time, thus Hamiltonian cycle is also in NP
- Thus Hamiltonian Cycle is NP-Complete



### Traveling Sales Person \_\_\_\_\_

### \_ NP-Hard Games \_\_\_\_\_

- A problem closely related to Hamiltonian cycle is the famous *Traveling Salesperson Problem(TSP)*
- The TSP problem is: "Given a weighted graph *G*, find the shortest cycle that visits every vertex.
- Finding the shortest cycle is obviously harder than determining if a cycle exists at all, so since Hamiltonian Cycle is NP-hard, TSP is also NP-hard!

- In 1999, Richard Kaye proved that the solitaire game Minesweeper is NP-Hard, using a reduction from Circuit Satifiability.
- Also in the last few years, Eric Demaine, et. al., proved that the game Tetris is NP-Hard

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Challenge Problem	Challenge Problem

- Consider the *optimization* version of, say, the graph coloring problem: "Given a graph *G*, what is the smallest number of colors needed to color the graph?" (Note that unlike the *decision* version of this problem, this is not a yes/no question)
- Show that the optimization version of graph coloring is also NP-Hard by a reduction from the decision version of graph coloring.
- Is the optimization version of graph coloring also NP-Complete?

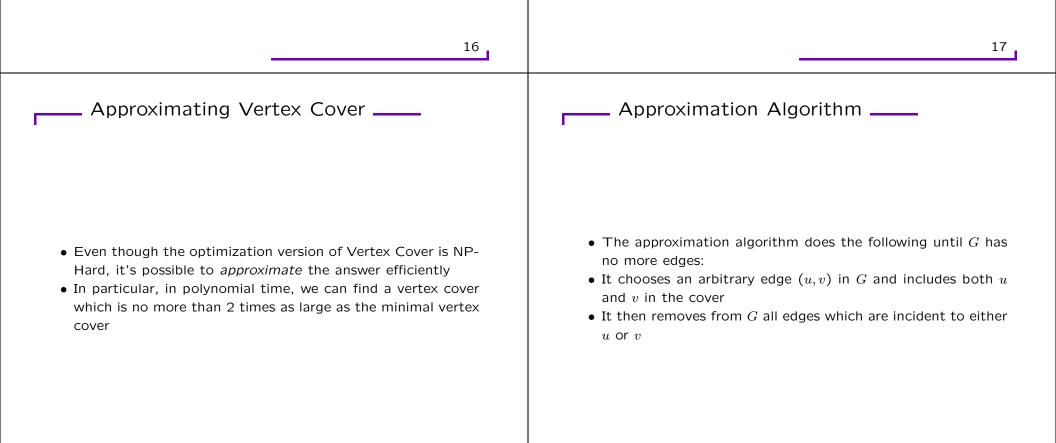
- Consider the problem 4Sat which is: "Is there any assignment of variables to a 4CNF formula that makes the formula evaluate to true?"
- Is this problem NP-Hard? If so, give a reduction from 3Sat that shows this. If not, give a polynomial time algorithm which solves it.

## Challenge Problem \_\_\_\_\_

## \_ Vertex Cover \_\_\_\_

- Consider the following problem: "Does there exist a clique of size 5 in some input graph *G*?"
- Is this problem NP-Hard? If so, prove it by giving a reduction from some known NP-Hard problem. If not, give a polynomial time algorithm which solves it.

- A *vertex cover* of a graph is a set of vertices that touches every edge in the graph
- The decision version of *Vertex Cover* is: "Does there exist a vertex cover of size k in a graph G?".
- We've proven this problem is NP-Hard by an easy reduction from Independent Set
- The *optimization* version of *Vertex Cover* is: "What is the minimum size vertex cover of a graph *G*?"
- We can prove this problem is NP-Hard by a reduction from the decision version of Vertex Cover (left as an exercise).



## Approximation Algorithm

\_\_\_ Analysis \_\_\_\_\_

- Approx-Vertex-Cover(G){ • If we implement the graph with adjacency lists, each edge  $C = \{\};$ need be touched at most once E' = Edges of G;• Hence the run time of the algorithm will be O(|V| + |E|), while(E' is not empty){ which is polynomial time let (u,v) be an arbitrary edge in E'; • First, note that this algorithm does in fact return a vertex add both u and v to C; cover since it ensures that every edge in G is incident to some remove from E' every edge incident to u or v; vertex in C} • Q: Is the vertex cover actually no more than twice the optimal return C; size? } 20 21 \_ TSP \_\_\_\_\_ Analysis \_\_\_\_\_
  - Let A be the set of edges which are chosen in the first line of the while loop
  - $\bullet$  Note that no two edges of A share an endpoint
  - Thus, *any* vertex cover must contain at least one endpoint of each edge in A
  - Thus if C\* is an optimal cover then we can say that  $|C*| \ge |A|$
  - Further, we know that |C| = 2|A|
  - This implies that  $|C| \leq 2|C*|$

Which means that the vertex cover found by the algorithm is no more than twice the size of an optimal vertex cover.

- An optimization version of the TSP problem is: "Given a weighted graph *G*, what is the shortest Hamiltonian Cycle of G?"
- This problem is NP-Hard by a reduction from Hamiltonian Cycle
- However, there is a 2-approximation algorithm for this problem if the edge weights obey the *triangle inequality*

## Triangle Inequality \_\_\_\_\_

#### • In many practical problems, it's reasonable to make the assumption that the weights, *c*, of the edges obey the *triangle inequality*

• The triangle inequality says that for all vertices  $u, v, w \in V$ :

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

- In other words, the cheapest way to get from u to w is always to just take the edge (u, w)
- In the real world, this is usually a pretty natural assumption. For example it holds if the vertices are points in a plane and the cost of traveling between two vertices is just the euclidean distance between them.

- Given a weighted graph G, the algorithm first computes a MST for G, T, and then arbitrarily selects a root node r of T.
- It then lets L be the list of the vertices visited in a depth first traversal of T starting at r.
- Finally, it returns the Hamiltonian Cycle, H, that visits the vertices in the order L.

Approx-TSP(G){
T = MST(G);

L = the list of vertices visited in a depth first traversal
 of T, starting at some arbitrary node in T;

Approximation Algorithm \_\_\_\_\_

return H;

}

#### Example Run a d c b f g c h h a d c b f g c h b f g c h c h c h

The top left figure shows the graph G (edge weights are just the Euclidean distances between vertices); the top right figure shows the MST T. The bottom left figure shows the depth first walk on T, W = (a, b, c, b, h, b, a, d, e, f, e, g, e, d, a); the bottom right figure shows the Hamiltonian cycle H obtained by deleting repeat visits from W, H = (a, b, c, h, d, e, f, g).

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## . Approximation Algorithm \_\_\_\_\_

Analysis \_\_\_\_\_

Analysis _	
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• The first step of the algorithm takes  $O(|E| + |V| \log |V|)$  (if we use Prim's algorithm)

- The second step is O(|V|)
- The third step is O(|V|).
- Hence the run time of the entire algorithm is polynomial

An important fact about this algorithm is that: the cost of the MST is less than the cost of the shortest Hamiltonian cycle.

- To see this, let T be the MST and let H\* be the shortest Hamiltonian cycle.
- Note that if we remove one edge from  $H\ast,$  we have a spanning tree, T'
- Finally, note that  $w(H*) \ge w(T') \ge w(T)$
- Hence  $w(H*) \ge w(T)$

29 28 \_ Analysis \_\_\_\_\_ Analysis \_\_\_\_\_ • Unfortunately, W is not a Hamiltonian cycle since it visits some vertices more than once • Now let W be a depth first walk of T which traverses each However, we can delete a visit to any vertex and the cost will edge exactly twice (similar to what you did in the hw) not increase because of the triangle inequality. (The path • In our example, W = (a, b, c, b, h, b, a, d, e, f, e, g, e, d, a)without an intermediate vertex can only be shorter) • Note that c(W) = 2c(T)• By repeatedly applying this operation, we can remove from • This implies that c(W) < 2c(H\*)W all but the first visit to each vertex, without increasing the cost of W. • In our example, this will give us the ordering H = (a, b, c, h, d, e, f, g) Analysis \_\_\_\_\_

\_ Take Away \_\_\_\_

- By the last slide,  $c(H) \leq c(W)$ .
- So  $c(H) \leq c(W) = 2c(T) \leq 2c(H*)$
- Thus,  $c(H) \leq 2c(H*)$
- In other words, the Hamiltonian cycle found by the algorithm has cost no more than twice the shortest Hamiltonian cycle.

- Many real-world problems can be shown to not have an efficient solution unless P = NP (these are the NP-Hard problems)
- However, if a problem is shown to be NP-Hard, all hope is not lost!
- In many cases, we can come up with an provably good approximation algorithm for the NP-Hard problem.

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