Analysis of Algorithms

CSCI 570

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Lecture 13

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NP-Completeness

Reading: chapter 9

P and NP complexity classes

P = set of problems that can be solved in polynomial time by a deterministic TM.

NP = set of problems that can be solved in polynomial time by a nondeterministic TM.

NP = set of problems for which solution can be verified in polynomial time by a deterministic TM.

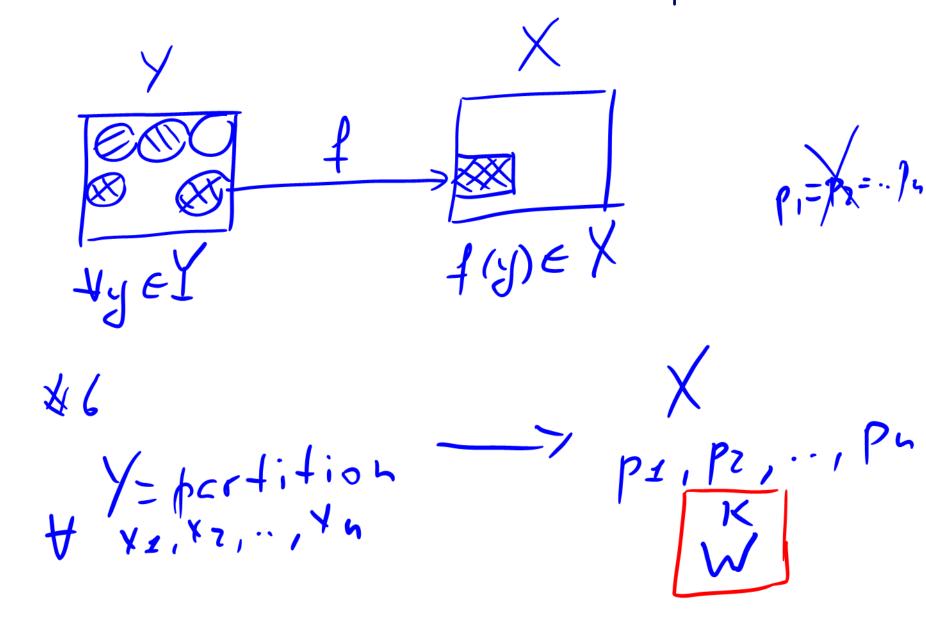
Polynomial Reduction: Y ≤_p X

To reduce a <u>decision</u> problem Y to a <u>decision</u> problem X (we write $Y \leq_p X$) we want a function f that maps Y to X such that:

- 1) f is a polynomial time computable
- \rightarrow 2) $\forall y \in Y (y \text{ is instance of } Y) \text{ is } YES$ if and only if $f(y) \in X \text{ is } YES$.
 - > If we cannot solve Y, we cannot solve X.

We use this to prove NP completeness: knowing that Y is hard, we prove that X is at least as hard as Y.

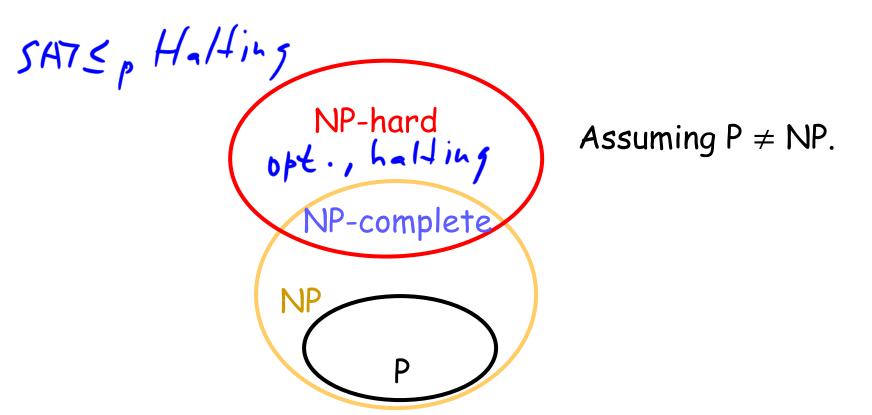
Polynomial Reduction: Y ≤_p X



NP-Hard and NP-Complete

X is NP-Hard, if $\forall Y \in NP$ and $Y \leq_p X$.

X is NP-Complete, if X is NP-Hard and $X \in NP$.



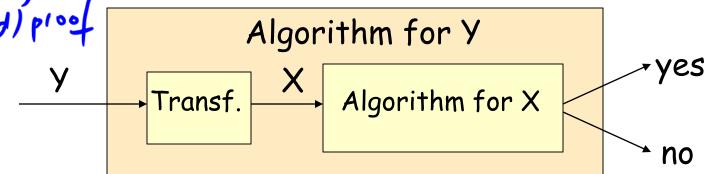
NP-Completeness Proof Method

$$Y \in_{p} X$$

To show that X is NP-Complete:

- 1) Show that X is in NP
- >> 2) Pick a problem Y, known to be an NP-Complete
 - 3) Prove $Y \leq_p X$ (reduce Y to X)
 a) construction, wap 1

 - b) fis polynomial



Boolean Satisfiability Problem (SAT)

A propositional logic formula is built from variables, operators AND (conjunction, \land), OR (disjunction, \lor), NOT (negation, \neg), and parentheses:

CNF
$$(X_1 \lor \neg X_3) \land (X_1 \lor \neg X_2 \lor X_4 \lor X_5) \land ... = \text{True}$$

A formula is said to be satisfiable if it can be made TRUE by assigning appropriate logical values (TRUE, FALSE) to its variables.

ECHTHC 7 Enlar Cycle EP Think . Hamiltonian Cycle Problem



A Hamiltonian cycle (HC) in a graph is a cycle that visits each vertex exactly once.

Problem Statement:

Given a directed or undirected graph G = (V,E). Find if the graph contains a Hamiltonian cycle.

We can prove it that HC problem is NP-complete by reduction from SAT, but we won't.



Traveling Salesman Problem

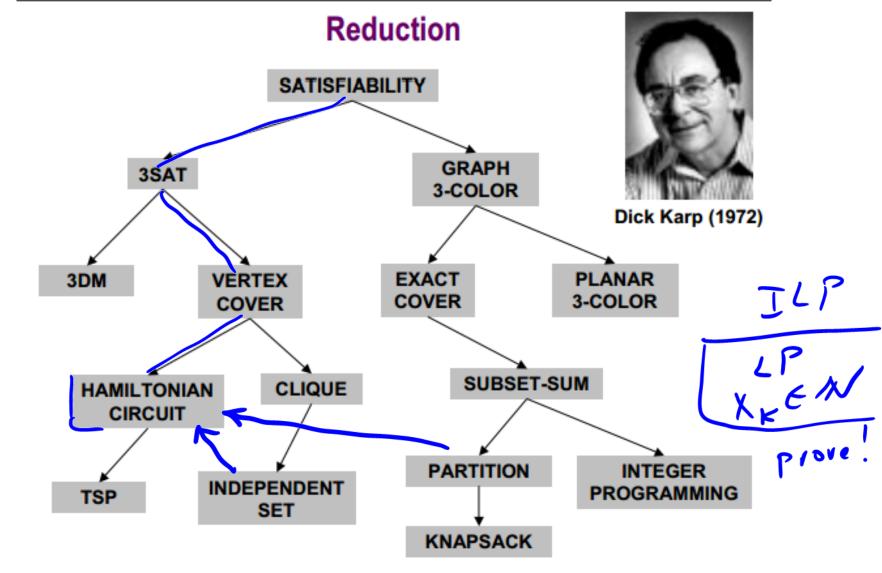


Given a list of cities and the distances between each pair of cities, what is the shortest possible route that visits each city and returns to the origin city?

Problem Statement:

Given a weighted graph G=(V,E) with positive edge costs, is there a Hamiltonian cycle that has total $cost \le k$?

We can prove that it is a NP-compete problem.



Karp introduced the now standard methodology for proving problems to be NP-Complete.

He received a Turing Award for his work (1985).

Discussion Problem 1

Given SAT in Conjunctive Normal Form (CNF)

$$(X_1 \lor \neg X_3) \land (X_1 \lor \neg X_2 \lor X_4 \lor X_5) \land \dots$$

with any number of clauses and any number of literals in each clause. Prove that SAT is polynomial time reducible to 3SAT.

PHSIST $aubucud = (aubux) \Lambda(\overline{X}vcud)$ Proo. TX=F aub: T (aubux) N(xucud) 1, d = + cud=T 9,6 =+ (aubux) 1 (xucvd) a=b=c=d=F hever occurs

(arbucu(due) = (arbucuz) = (aubux) $\Lambda(\overline{x}ucudue)$ = (aubux) $\Lambda(\overline{x}ucudue)$ = (aubux) $\Lambda(\overline{x}ucudue)$ = (aubux) $\Lambda(\overline{x}ucudue)$ (aubucudueuf)=...h-literals, m-clauses isit polynomial? SAT: h-literals, m-clauses 35AT: Citerals > (h-3)m, clauses > (h-2)m Claim. SAT is satisfiable itt 35AT is satis Proof => by construction => 3SATE SAT

Discussion Problem 2

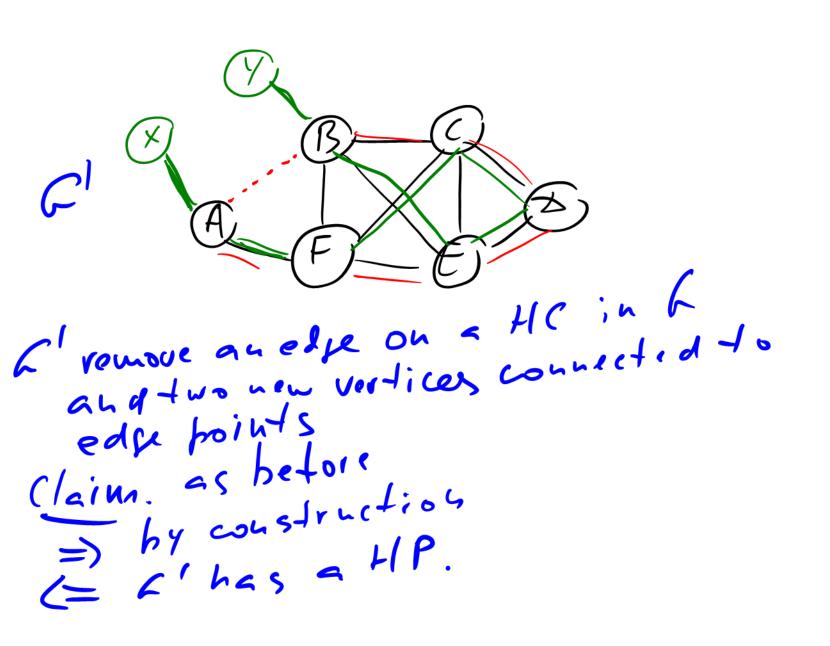
Assuming that finding a Hamiltonian Cycle (HC) in a graph is NP-complete, prove that finding a Hamiltonian Path is also NP-complete. HP is a path that visits each vertex exactly once and isn't required to return to its starting point.

required to return to its starting point.

$$H(ENPC, prove HPENPC)$$

 $HPENP, easy$
 HCE_PHP

remove an edge from a KIC Chas a HC iff G'has a HP by construction C= F'has a HP wrong veduction



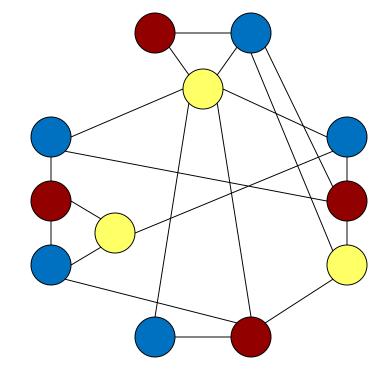
Graph Coloring

Given a graph, can you color the nodes with < k colors such that the endpoints of every edge are colored differently?

O Planar graph Coloring EP

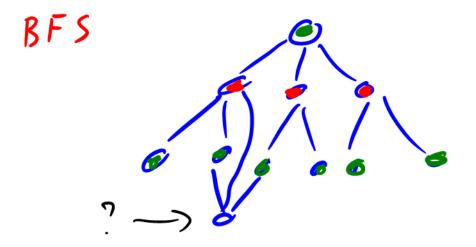
Oit K=Z, in P

Theorem. (k>2) k-Coloring is NP-complete.



Graph Coloring: k = 2

How can we test if a graph has a 2-coloring?

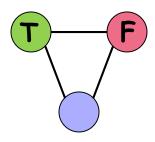


We construct a graph G that will be 3-colorable iff the 3-SAT instance is satisfiable.

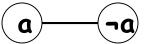
Graph G consists of the following gadgets.

onlyone

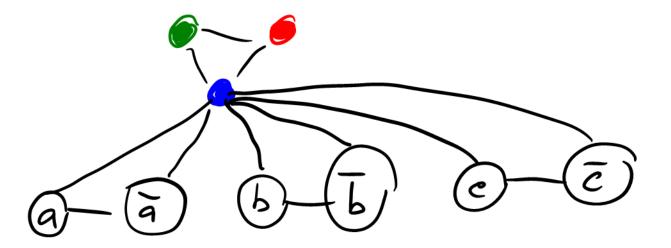
A truth gadget:



A gadget for each variable:



Combining those gadgets together (for three literals)



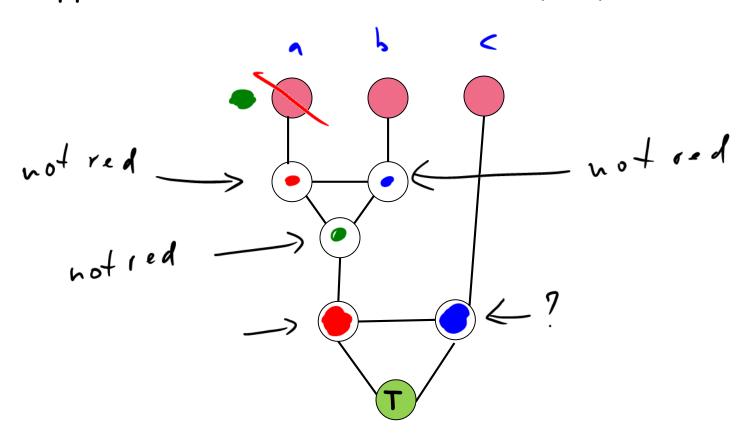
A special gadget for each clause

This gadget connects a truth gadget with variable gadgets.

The hottom vertex is T/green)
iff one of a,b,c is true.

avbuc = F ita=b=c=F galget for lilerals

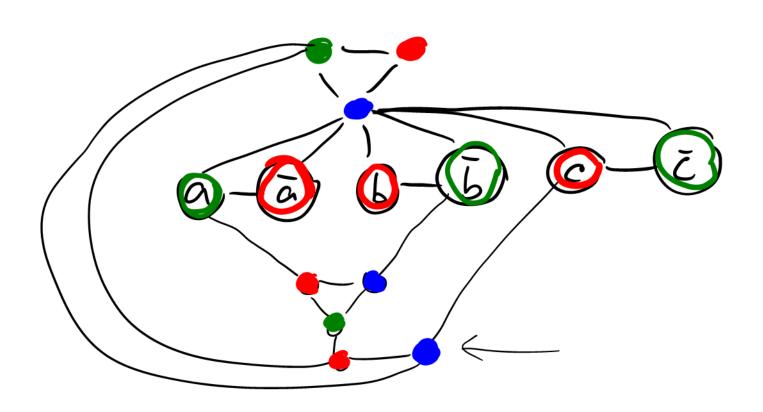
Suppose all a, b and c are all False (red).



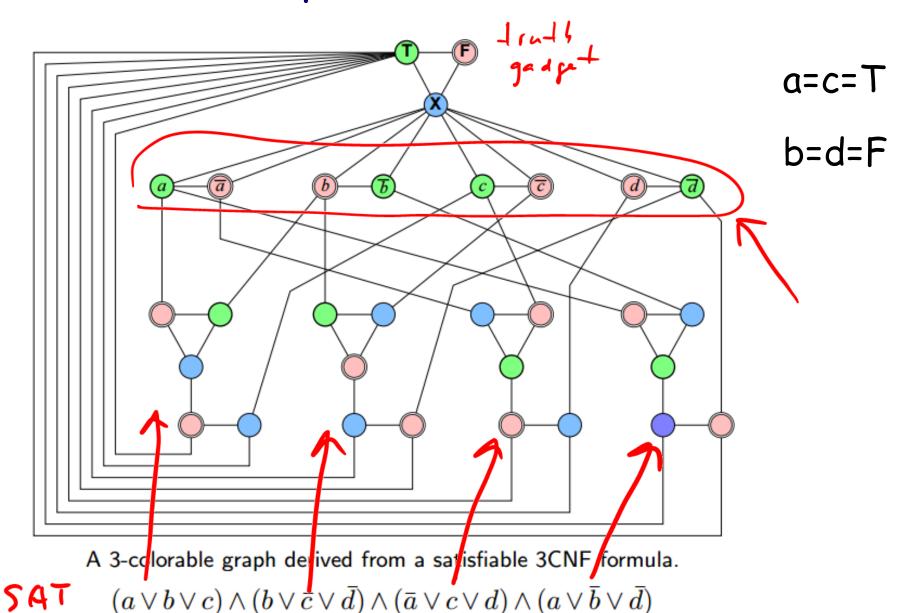
We have showed that if all the variables in a clause are false, the gadget cannot be 3-colored.

Example: a V -b V c

Let == 1, b= (= F



Example with four clauses



Claim: 3-SAT instance is satisfiable if and only if G is 3-colorable.

Proof: =) 354T has truth assimment

by construction

truth gadget -> color

variable gadgets -> color, T= ween

variable gadgets -> coloring is forced

clause radgets -> coloring is

Claim: 3-SAT instance is satisfiable if and only if G is 3-colorable.

Sudoku: n²×n²

NP-hard?

9-coloring

Sudoku grath:

Vertex: cell, 81

Edge: tw vertices is the seme row or estunctor mini-grid

2			3		8		5	
		3		4	5	9	8	
		8			9	7	3	4
6		7		9				
9	8						1	7
				5		6		9
3	1	9	7			2		
	4	6	5	2		8		
	2		9		3			1

Sudoku Graph

how many edges? 810

$$8/.20/2 = 810$$

0	•	•	3	•	8	•	5	,
•	X	X		4	5	9	8	
•	X	*			9	7	3	4
6		7		9				
9	8						1	7
•				5		6		9
3	1	9	7			2		
•	4	6	5	2		8		
•	2		9		3			1

Sudoku

Constructing a Sudoku graph we have proved:

Y to solve problem Y.

Don't be afraid of NP-hard problems.

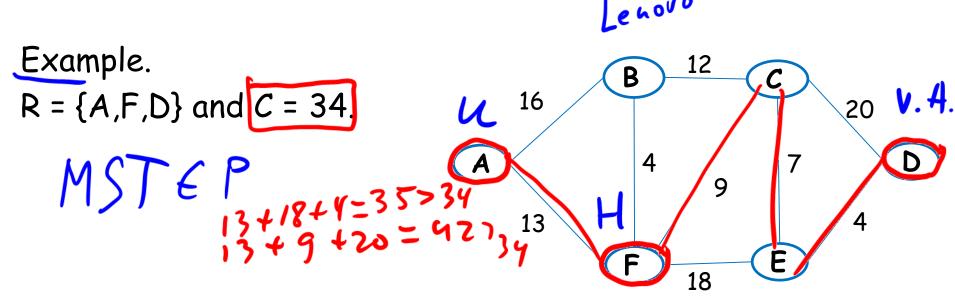
Many reasonable instances (of practical interest) of problems in class NP can be solved!



The largest solved TSP an 85,900-vertex route calculated in 2006. The graph corresponds to the design of a customized computer chip created at Bell Laboratories, and the solution exhibits the shortest path for a laser to follow as it sculpts the chip.

Discussion Problem 3

The Steiner Tree problem is as follows. Given an undirected weighted graph G=(V,E) with positive edge costs, a subset of vertices $R\subseteq V$, and a number C Is there a tree in G that spans all vertices in R (and possibly some other in V) with a total edge cost of at most C? Prove that this problem is NP-complete by reduction from Vertex Cover.



G = (V, E) $VC(G) \leq K$ 6, we have to define Rand C.
assign weights to each edge vertices: V+E R=red vertices SteinerTree in green.

Claim. Chasa VC of size
and wost Kiff C/has a Steinor I rec with R= E(6) and C \left E+K-1. Proof. has all of size & steins free

By construction, we create a E+K-1

with R=E(1) and CIE+K-1 E) L'hasa Steiner Tree with Rand (EE+14) Find UCtor 6. VC(f) = ST-R a) Prove +44+ this is = VC=57-R 1vc1=15Tofvertices | - R(vertices) = cosf(s7)+1-E $\leq (E+k-1)+1-E=K.$

Approximation Algorithms



ML

Suppose we are given an NP-hard problem to solve.

sub optimal

Can we develop a polynomial-time algorithm that produces a "good enough" solution?

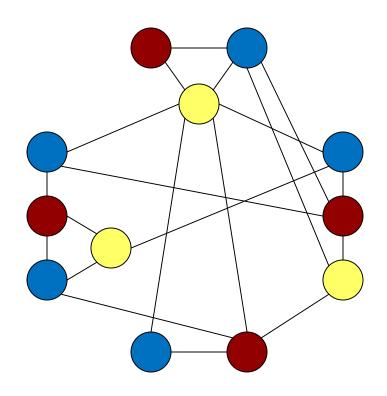
An algorithm that returns near-optimal solutions is called an approximation algorithm.

Graph Coloring

Given a graph G=(V,E), find the minimum number of colors required to color vertices, so no two adjacent vertices have the same color.

This is NP-hard problem.

Let us develop a solution that is close enough to the optimal solution.



Greedy Approximation

Given G=(V,E) with n vertices.

Use the integers {1,2,3, ..., n} to represent colors.

Order vertices by degree in descending order.

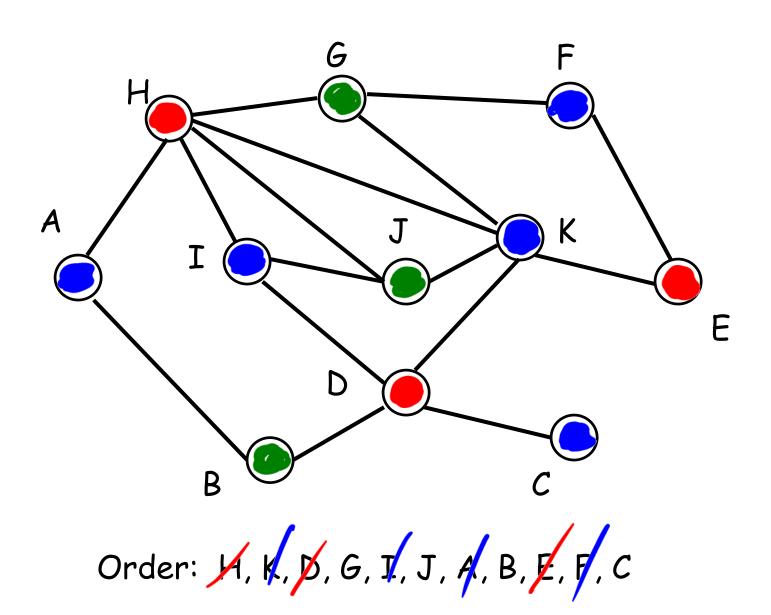
Color the first vertex (highest degree) with color 1.

Go down the vertex list and color every vertex not adjacent to it with color 1.

Remove all colored vertices from the list.

Repeat for uncolored vertices with color 2.

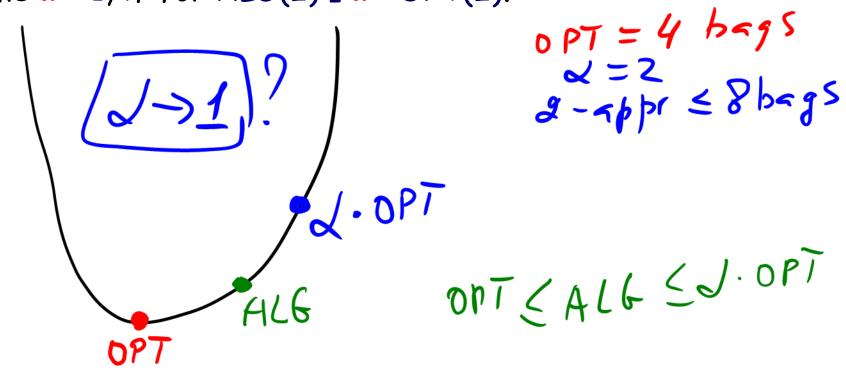
Example



Formal Definition

Let P be a <u>minimization</u> problem, and I be an instance of P. Let ALG(I) be a solution returned by an algorithm, and let OPT(I) be the optimal solution.

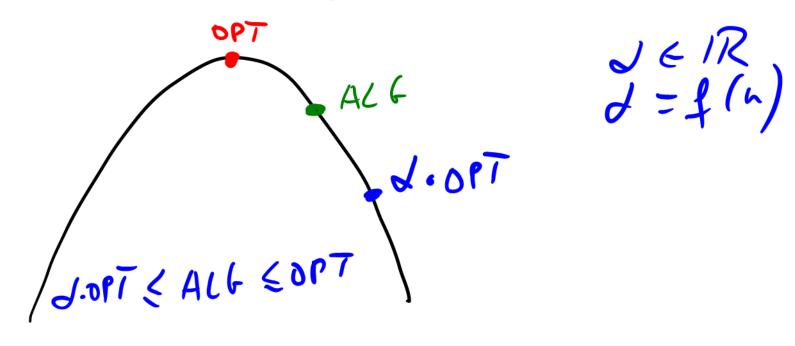
Then ALG(I) is said to be a α -approximation algorithm for some $\alpha > 1$, if for $ALG(I) \leq \alpha \cdot OPT(I)$.



Maximization Problem

Let P be a <u>maximization</u> problem, and I be an instance of P. Let ALG(I) be a solution returned by an algorithm, and let OPT(I) be the optimal solution.

Then ALG(I) is said to be a α -approximation algorithm for some $0 < \alpha < 1$, if for $ALG(I) \ge \alpha \cdot OPT(I)$.

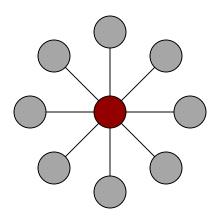


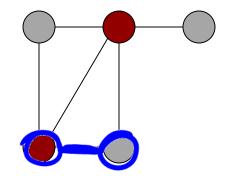
Vertex cover < NPH

Given G=(V,E), find the smallest $S\subset V$ s.t. every edge is incident on a vertex in S.

random

Let us design a greedy algorithm for vertex cover.



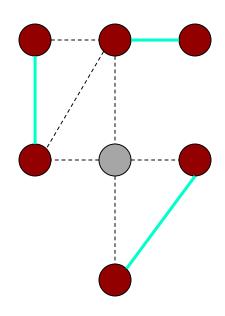


Example

2 - appox. tomaked smaller (isit possible 222 Claim. [Alb] \(2./2P7/ Proof. 1Alb= 1matching/.9. 1A(4)=1matching/.2

Vertex cover

Lemma. Let M be a matching in G, and S be a vertex cover, then $|S| \ge |M|$



2-Approximation Vertex Cover

Approx-VC(G):

M - maximal matching on G

S - take both endpoints of edges in M

Return S

Theorem. Let OPT(G) be the size of the optimal vertex cover and S = ApproxVC(G). Then $|S| \le 2 \cdot OPT(G)$.

Proof.

Can we do better than 2-Approximation?

$$|DPT(K_{n,n})| = ? = n$$

$$|Maximal Madching| = ? = n$$

$$ALF(K_{n,n}) = 2n$$

$$J = 2$$

$$The End$$

