

# Announcements

# Announcements

- Final HW is out
- Exam 3 will be posted on Sunday
- Timeslot signup will be announced this week- keep an eye on your e-mail

# The Theory of Computational Complexity

# Search vs. Optimization

- Search problem: Find solution satisfying some requirement. *User specifies requirement / constraint*
  - Find spanning tree with weight at most  $b$
  - Find bipartite matching with flow greater than  $b$
  - Etc.
- Optimization problem: Find *best* solution
  - Find minimum spanning tree
  - Find heaviest bipartite matching
  - Etc.
- Why are these basically equivalent?

# Search vs. Optimization

not algorithms!

→  $\textcircled{P}$  NP

- $\textcircled{P}$  (Polynomial): The class of all problems that can be solved in polynomial running time ~~\*~~
- $\textcircled{NP}$  (Non-deterministic Polynomial Time): The class of search problems with solutions that can be verified in polynomial time.

deterministic

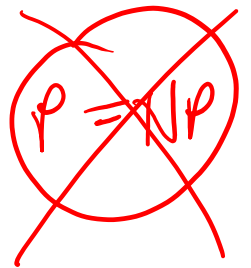
→ checked

~~\*~~ there exists an algorithm for which the worst-case running time is a polynomial function of input size

# Is $P = NP$ ?

$(x \vee y \vee z) \dots (w \vee \bar{a} \vee \bar{b})$

- Whether  $P = NP$  is an open question!
- Most computer scientists think that  $P \neq NP$ .
- But we don't have a proof...
- One of the most important open questions in computer science

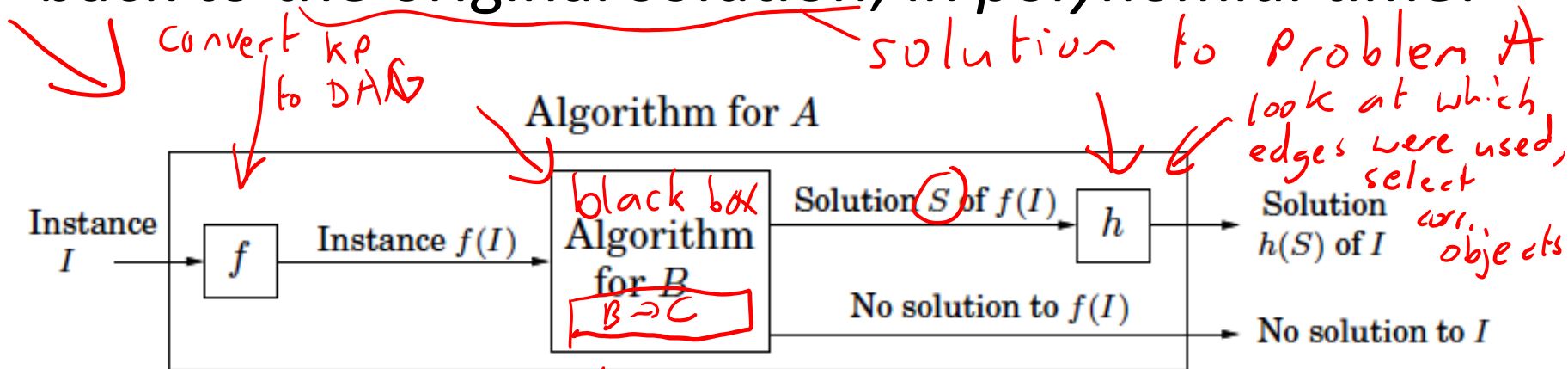


verification: there is some problem and a black box algorithm that claims to solve it. We give the BB input and it produces output. Can we check whether the output is correct in polynomial time?

# Reductions

dynamic programming  
input KP  $\rightarrow$  longest path on DAG  
 $P_A \rightarrow P_B$

- **Problem A reduces to Problem B** if you can convert every instance of Problem A to an instance of Problem B, and convert the solution to Problem B back to the original solution, in *polynomial time*.



- big factory / warehouses  $\rightarrow$  network flow
- Is reduction transitive?  
 $A \rightarrow B$      $B \rightarrow C$      $A \rightarrow C$  yes

# Reductions

① ① ② ... (w-1) (w)

- Why are reductions useful?
- If we have <sup>a polynomial-time</sup> ~~an~~ algorithm for Problem B, and can <sub>f, h</sub> convert back and forth in polynomial time, then we can solve Problem A in polynomial time!
- Example: Bipartite matching reduces to network flows

— many DP problems  $\rightarrow$  longest/shortest path



# NP-Completeness

- Tricky use of reductions: If Problem A *does not* have a polynomial time solution, and it *reduces* to Problem B, then Problem B *also does not* have a polynomial time solution!  $A \rightarrow B$   
*F, h are polynomial-time*
- A search problem is NP-complete if *every other problem in NP reduces to it*
- In other words: if you can solve an NP-complete problem quickly, you can solve anything in NP quickly!  
*polynomial time  $\Rightarrow P = NP$*

# Reduction Strategy

$A \rightarrow B$   ~~$B \rightarrow A$~~  | Problem B is at least as hard as Problem A

## 1. Direction of reduction depends on your goal:

— If you have a polynomial algorithm for Problem ~~B~~<sup>B</sup>, and want to find one for Problem ~~B~~<sup>A</sup>, show that ~~B~~<sup>A</sup> reduces to ~~B~~<sup>B</sup>

( $x \vee y$ )  
2SAT

$A \rightarrow B$

— If you want to show that Problem B has no polynomial solution, show that some NP-complete problem ~~A~~ reduces to Problem B

( $x \vee y \vee z$ )  
3SAT

Suppose we try  
 $B \rightarrow A$  instead  
↓  
NP-C

2SAT  $\rightarrow$  SAT

$A \rightarrow B$

NP-complete

A is hard! (no polynomial time alg for A)

by contradiction,  
means that B  
does not have  
poly-time algorithm

# Reduction Strategy

2. When reducing Problem A reduces to Problem B, show the following: *0. Find the reduction (f,h)*
1. Any instance of Problem A can be converted to an instance of Problem B in polynomial time *(f runs in poly-time)*
  2. A solution to the converted instance can be converted back to a solution for Problem A in polynomial time *(h)*
  3. If Algorithm B finds a solution to the converted instance, it corresponds to an actual solution to the Problem A instance *→ solution is correct*
  4. If the Problem A instance has a solution, then Algorithm B is able to find a solution to the converted instance  
*If Algorithm B says "no solution", then there is no solution to prob. A input*

# In-Class Exercise: Hamiltonian Cycles and Paths

- Hamiltonian Cycle: in an undirected graph, ~~is there~~ <sup>Find</sup> a cycle that passes through every node exactly once?
- Hamiltonian Path: Given vertices  $s$  and  $t$ , ~~is there~~  <sup>$s \neq t$  in undirected graph</sup> a path from  $s$  to  $t$  that passes through every node exactly once? <sub>Find</sub>
- Show how to reduce Hamiltonian Path to Hamiltonian Cycle. (i.e., if we have an algorithm for the Hamiltonian Cycle problem, can we use that to solve the Hamiltonian Path problem?)

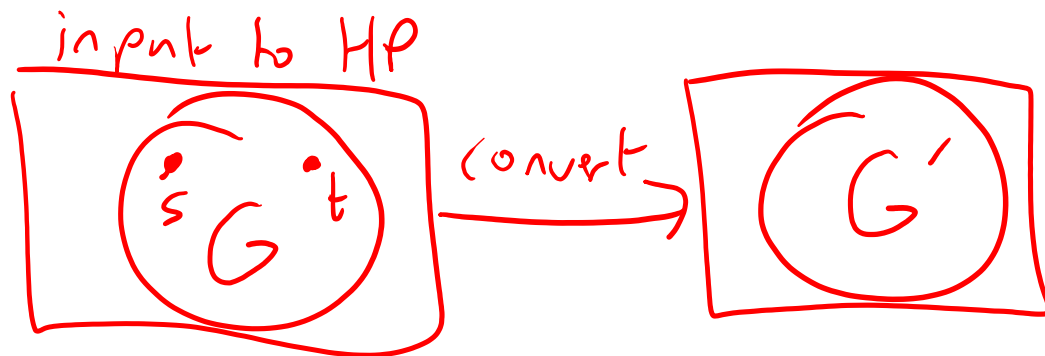
# In-Class Exercise: Hamiltonian Cycles and Paths

What does it mean to reduce  $HP \rightarrow HC$ ?

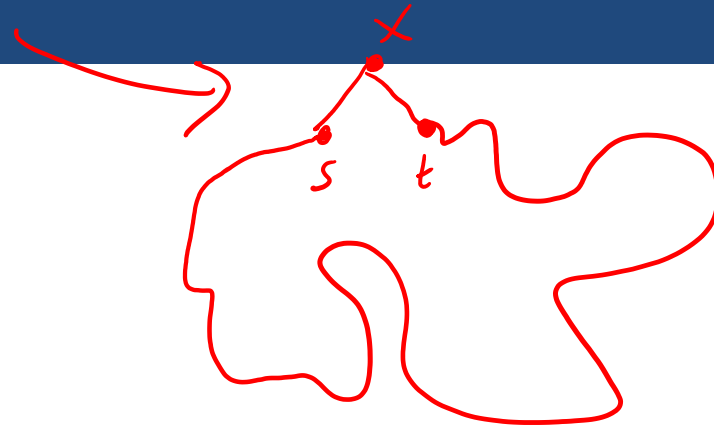
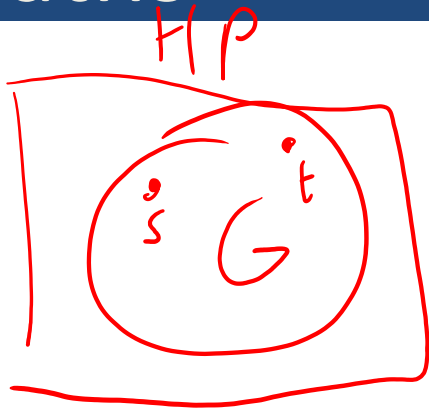
- Assume we have a black box algo for HC

- Construct an algorithm for HP

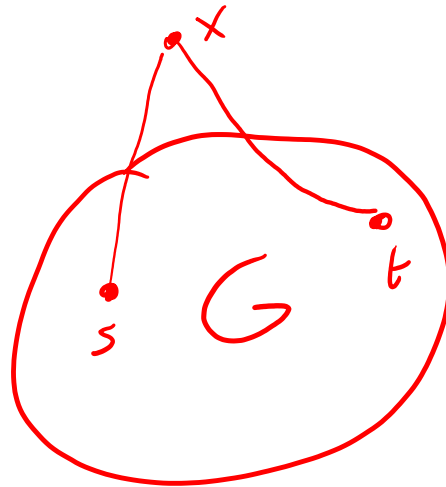
- input is graph  $G$ , two nodes  $s$  &  $t$



# In-Class Exercise: Hamiltonian Cycles and Paths



$G' =$



MC  
Black box

a HC on  $G'$  touches  $x$  exactly once.  
So it must enter  $x$  from  $s$  or  $t$  and leave by the other edge.

# In-Class Exercise: Hamiltonian Cycles and Paths

f: add node  $x$ , ~~and~~ <sup>add</sup> edges  $(s, x)$   $(t, x)$

h: ~~so~~ remove  $x$ ,  $(s, x)$ ,  $(t, x)$  from the HC output by black box, what's left will be a HP in  $G$  from  $s \rightarrow t$

1. Show that  $f$  runs in poly-time.

All we do is add a row/column, change 4 0s to 1 in adj. matrix.

Obv. poly-time.

2. Show that  $h$  runs in poly time, Obvious, all we do is remove 2 edges from HC.

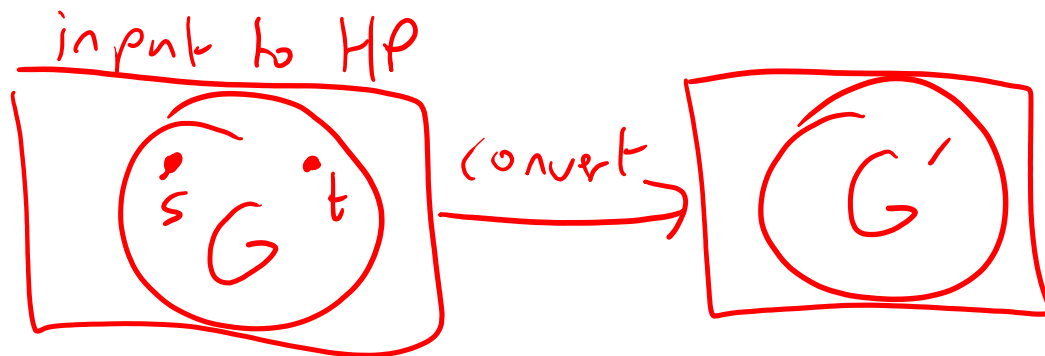
# In-Class Exercise: Hamiltonian Cycles and Paths

What does it mean to reduce  $HP \rightarrow HC$ ?

- Assume we have a black box algo for HC

- Construct an algorithm for HP

- input is graph  $G$ , two nodes  $s$  &  $t$





# In-Class Exercise: Hamiltonian Cycle and Traveling Salesman Problem

- Hamiltonian Cycle: Does the graph have a cycle that visits every node?
- Traveling Salesman Problem: Given a graph with integer weights on the edges, and a starting node  $s$ , and an integer budget  $b$ , is it possible to find a cycle beginning and ending at  $s$  such that the total weight of edges in the cycle is less than  $b$ ?  
*that touches every node*
- Reduce Hamiltonian Cycle to TSP

# In-Class Exercise: Hamiltonian Cycle and Traveling Salesman Problem

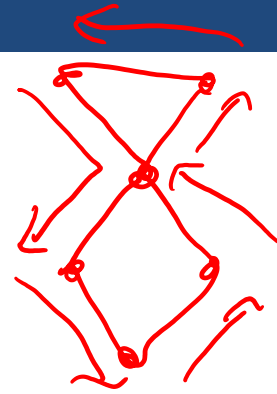
unweighted  
 $G$

input to HC

weighted  
 $G'$

nodes

budget  $b$



$f$ : adds weights to  $G$   
sets  $s = \text{some node}$   
sets  $b = \# \text{ nodes}$

$h$ : returns the same  
output

$s = \text{arbitrary node}$

$b = \# \text{ nodes}$

weight in  $G' = 1$

cycle visits every node  
at least once, but  
total  $\# \text{ edges} / \text{nodes}$   
in the cycle  $\leq b$ ,  
so each node seen  
exactly once