## CS 473: Algorithms, Fall 2019

## Reductions

Lecture 22
November 14, 2019

## Part I

## Reductions

## Reductions

A reduction from Problem $\boldsymbol{X}$ to Problem $\boldsymbol{Y}$ means (informally) that if we have an algorithm for Problem $\boldsymbol{Y}$, we can use it to find an algorithm for Problem $\boldsymbol{X}$.

## Using Reductions

(1) We use reductions to find algorithms to solve problems.
(2) We also use reductions to show that we can't find algorithms for some problems. (We say that these problems are hard.)

## Example 1: Bipartite Matching and Flows

## How do we solve the Bipartite Matching Problem?

Given a bipartite graph $G=(U \cup V, E)$ and number $k$, does $G$ have a matching of size $\geq k$ ?

## Solution

Reduce it to Max-Flow. $G$ has a matching of size $\geq \boldsymbol{k}$ iff there is a flow from $s$ to $t$ of value $\geq k$ in the auxiliary graph $G^{\prime}$.

## Types of Problems

## Decision, Search, and Optimization

(1) Decision problem. Example: given $\boldsymbol{n}$, is $\boldsymbol{n}$ prime?
(2) Search problem. Example: given $\boldsymbol{n}$, find a factor of $\boldsymbol{n}$ if it exists.
(3) Optimization problem. Example: find the smallest prime factor of $n$.

## Optimization and Decision problems

 For max flow...
## Problem (Max-Flow optimization version)

Given an instance $G$ of network flow, find the maximum flow between $s$ and $t$.

## Problem (Max-Flow decision version)

Given an instance $G$ of network flow and a parameter $K$, is there a flow in $G$, from $s$ to $t$, of value at least $K$ ?

While using reductions and comparing problems, we typically work with the decision versions. Decision problems have Yes/No answers. This makes them easy to work with.

## Problems vs Instances

(1) A problem $\boldsymbol{\Pi}$ consists of an infinite collection of inputs $\left\{I_{1}, I_{2}, \ldots,\right\}$. Each input is referred to as an instance.
(2) The size of an instance $\boldsymbol{I}$ is the number of bits in its representation.
(3) For an instance $I$, sol(I) is a set of feasible solutions to $I$.

- For optimization problems each solution $s \in \operatorname{sol}(I)$ has an associated value.


## Examples

## Example

An instance of Bipartite Matching is a bipartite graph, and an integer $\boldsymbol{k}$. The solution to this instance is "YES" if the graph has a matching of size $\geq \boldsymbol{k}$, and "NO" otherwise.

## Example

An instance of Max-Flow is a graph $G$ with edge-capacities, two vertices $s, t$, and an integer $\boldsymbol{k}$. The solution to this instance is "YES" if there is a flow from $s$ to $t$ of value $\geq k$, else "NO".

## What is an algorithm for a decision Problem X?

It takes as input an instance of $\boldsymbol{X}$, and outputs either "YES" or "NO".

## Using reductions to solve problems

(1) $\mathcal{R}$ : Reduction $X \rightarrow Y$
(2) $\mathcal{A}_{Y}$ : algorithm for $Y$ :

- $\Longrightarrow$ New algorithm for $\boldsymbol{X}$ :

$$
\begin{aligned}
\mathcal{A}_{X}\left(I_{X}\right): & \\
& / / I_{X}: \text { instance of } X . \\
& I_{Y} \Leftarrow \mathcal{R}\left(I_{X}\right) \\
& \text { return } \mathcal{A}_{Y}\left(I_{Y}\right)
\end{aligned}
$$



If $\mathcal{R}$ and $\mathcal{A}_{\boldsymbol{Y}}$ polynomial-time $\Longrightarrow \mathcal{A}_{\boldsymbol{X}}$ polynomial-time.

## Comparing Problems

(1) "Problem $X$ is no harder to solve than Problem $Y$ ".
(2) If Problem $X$ reduces to Problem $Y$ (we write $X \leq Y$ ), then $X$ cannot be harder to solve than $Y$.
(0) Bipartite Matching $\leq$ Max-Flow. Bipartite Matching cannot be harder than Max-Flow.

- Equivalently, Max-Flow is at least as hard as Bipartite Matching.
(0) $X \leq Y$ :
(1) $\boldsymbol{X}$ is no harder than $\boldsymbol{Y}$, or
(0) $\boldsymbol{Y}$ is at least as hard as $\boldsymbol{X}$.


## Polynomial-time reductions

We say that an algorithm is efficient if it runs in polynomial-time.
To find efficient algorithms for problems, we are only interested in polynomial-time reductions. Reductions that take longer are not useful.

If we have a polynomial-time reduction from problem $X$ to problem $Y$ (we write $X \leq_{p} Y$ ), and a poly-time algorithm $\mathcal{A}_{\boldsymbol{Y}}$ for $Y$, we have a polynomial-time/efficient algorithm for $\boldsymbol{X}$.


## Polynomial-time Reduction

A polynomial time reduction from a decision problem $X$ to a decision problem $\boldsymbol{Y}$ is an algorithm $\mathcal{A}$ that has the following properties:
(1) given an instance $\boldsymbol{I}_{\boldsymbol{X}}$ of $\boldsymbol{X}, \mathcal{A}$ produces an instance $\boldsymbol{I}_{\boldsymbol{Y}}$ of $\boldsymbol{Y}$
(2) $\mathcal{A}$ runs in time polynomial in $\left|\boldsymbol{I}_{\boldsymbol{X}}\right|$.
(3) Answer to $\boldsymbol{I}_{\boldsymbol{X}}$ YES iff answer to $\boldsymbol{I}_{\boldsymbol{Y}}$ is YES.

## Proposition

If $\boldsymbol{X} \leq_{P} \boldsymbol{Y}$ then a polynomial time algorithm for $\boldsymbol{Y}$ implies a polynomial time algorithm for $\boldsymbol{X}$.

Such a reduction is called a Karp reduction. Most reductions we will need are Karp reductions.

## Reductions again...

Let $X$ and $Y$ be two decision problems, such that $X$ can be solved in polynomial time, and $X \leq_{P} Y$. Then
(A) $Y$ can be solved in polynomial time.
(B) $Y$ can NOT be solved in polynomial time.
(C) If $Y$ is hard then $X$ is also hard.
(D) None of the above.
(E) All of the above.

## Polynomial-time reductions and hardness

For decision problems $X$ and $Y$, if $X \leq_{P} Y$, and $Y$ has an efficient algorithm, $\boldsymbol{X}$ has an efficient algorithm.

If $\boldsymbol{X} \leq_{P} \boldsymbol{Y}$ and $\boldsymbol{X}$ does not have an efficient algorithm, $\boldsymbol{Y}$ cannot have an efficient algorithm!

## Polynomial-time reductions and instance sizes

## Proposition

Let $\mathcal{R}$ be a polynomial-time reduction from $\boldsymbol{X}$ to $\boldsymbol{Y}$. Then for any instance $\boldsymbol{I}_{\boldsymbol{X}}$ of $\boldsymbol{X}$, the size of the instance $\boldsymbol{I}_{\boldsymbol{Y}}$ of $\boldsymbol{Y}$ produced from $\boldsymbol{I}_{\boldsymbol{X}}$ by $\mathcal{R}$ is polynomial in the size of $\boldsymbol{I}_{\boldsymbol{X}}$.

## Proof.

$\mathcal{R}$ is a polynomial-time algorithm and hence on input $\boldsymbol{I}_{\boldsymbol{X}}$ of size $\left|\boldsymbol{I}_{\boldsymbol{X}}\right|$ it runs in time $\boldsymbol{p}\left(\left|I_{X}\right|\right)$ for some polynomial $\boldsymbol{p}()$.
$\boldsymbol{I}_{\boldsymbol{Y}}$ is the output of $\mathcal{R}$ on input $\boldsymbol{I}_{\boldsymbol{X}}$.
$\mathcal{R}$ can write at most $p\left(\left|I_{X}\right|\right)$ bits and hence $\left|I_{Y}\right| \leq p\left(\left|I_{X}\right|\right)$.
Note: Converse is not true. A reduction need not be polynomial-time even if output of reduction is of size polynomial in its input.

## Polynomial-time Reduction

A polynomial time reduction from a decision problem $X$ to a decision problem $\boldsymbol{Y}$ is an algorithm $\mathcal{A}$ that has the following properties:
(1) Given an instance $\boldsymbol{I}_{\boldsymbol{X}}$ of $\boldsymbol{X}, \mathcal{A}$ produces an instance $\boldsymbol{I}_{\boldsymbol{Y}}$ of $\boldsymbol{Y}$.
(2) $\mathcal{A}$ runs in time polynomial in $\left|I_{\boldsymbol{X}}\right|$. This implies that $\left|I_{\boldsymbol{Y}}\right|$ (size of $I_{Y}$ ) is polynomial in $\left|I_{\boldsymbol{X}}\right|$.
(3) Answer to $I_{X}$ YES iff answer to $I_{Y}$ is YES.

## Proposition

If $\boldsymbol{X} \leq_{P} \boldsymbol{Y}$ then a polynomial time algorithm for $\boldsymbol{Y}$ implies a polynomial time algorithm for $\boldsymbol{X}$.

Such a reduction is called a Karp reduction. Most reductions we will need are Karp reductions

## Transitivity of Reductions

## Proposition <br> $X \leq_{p} Y$ and $Y \leq_{p} Z$ implies that $X \leq_{p} Z$.

Note: $X \leq_{P} Y$ does not imply that $Y \leq_{P} X$ and hence it is very important to know the FROM and TO in a reduction.

To prove $\boldsymbol{X} \leq_{p} \boldsymbol{Y}$ you need to show a reduction FROM $\boldsymbol{X}$ TO $\boldsymbol{Y}$ In other words show that an algorithm for $\boldsymbol{Y}$ implies an algorithm for $X$.

## Using Reductions to show Hardness

Here, we say that a problem is "hard" if there is no polynomial-time algorithm known for it (and it is believed that such an algorithm does not exist)

- Start with an existing "hard" problem $\boldsymbol{X}$
- Prove that $X \leq_{P} Y$
- Then we have shown that $Y$ is a "hard" problem


## Examples of hard problems

## Problems

(1) SAT
(2) 3SAT
( Independent Set and Clique

- Vertex Cover
© Set Cover
- Hamilton Cycle
© Knapsack and Subset Sum and Partition
© Integer Programming
- ...


## Part II

## Examples of Reductions

## Independent Sets and Cliques

Given a graph $G$, a set of vertices $V^{\prime}$ is:
(1) independent set: no two vertices of $V^{\prime}$ connected by an edge.
(2) clique: every pair of vertices in $V^{\prime}$ is connected by an edge of G

## Independent Sets and Cliques

Given a graph $G$, a set of vertices $V^{\prime}$ is:
(1) independent set: no two vertices of $V^{\prime}$ connected by an edge.
(2) clique: every pair of vertices in $V^{\prime}$ is connected by an edge of G.


## Independent Sets and Cliques

Given a graph $G$, a set of vertices $V^{\prime}$ is:
(1) independent set: no two vertices of $V^{\prime}$ connected by an edge.
(2) clique: every pair of vertices in $V^{\prime}$ is connected by an edge of G.


## Independent Sets and Cliques

Given a graph $G$, a set of vertices $V^{\prime}$ is:
(1) independent set: no two vertices of $V^{\prime}$ connected by an edge.
(2) clique: every pair of vertices in $V^{\prime}$ is connected by an edge of G.


## The Independent Set and Clique Problems

## Problem: Independent Set

Instance: A graph G and an integer $\boldsymbol{k}$.
Question: Does $G$ has an independent set of size $\geq \boldsymbol{k}$ ?
Problem: Clique
Instance: A graph G and an integer $k$.
Question: Does $G$ has a clique of size $\geq k$ ?

## Reducing Independent Set to Clique

An instance of Independent Set is a graph $\boldsymbol{G}$ and an integer $\boldsymbol{k}$.

## Reducing Independent Set to Clique

An instance of Independent Set is a graph $\boldsymbol{G}$ and an integer $\boldsymbol{k}$.


## Reducing Independent Set to Clique

An instance of Independent Set is a graph $\boldsymbol{G}$ and an integer $\boldsymbol{k}$.
Convert $\boldsymbol{G}$ to $\overline{\boldsymbol{G}}$, in which $(\boldsymbol{u}, \boldsymbol{v})$ is an edge iff $(\boldsymbol{u}, \boldsymbol{v})$ is not an edge of $\boldsymbol{G}$. ( $\bar{G}$ is the complement of $\boldsymbol{G}$.)
We use $\bar{G}$ and $k$ as the instance of Clique.


## Reducing Independent Set to Clique

An instance of Independent Set is a graph $\boldsymbol{G}$ and an integer $\boldsymbol{k}$.
Convert $\boldsymbol{G}$ to $\bar{G}$, in which $(\boldsymbol{u}, \boldsymbol{v})$ is an edge iff $(\boldsymbol{u}, \boldsymbol{v})$ is not an edge of $\boldsymbol{G}$. ( $\overline{\mathcal{G}}$ is the complement of $\boldsymbol{G}$.)
We use $\bar{G}$ and $k$ as the instance of Clique.


## Reducing Independent Set to Clique

An instance of Independent Set is a graph $\boldsymbol{G}$ and an integer $\boldsymbol{k}$.
Convert $\boldsymbol{G}$ to $\overline{\boldsymbol{G}}$, in which $(\boldsymbol{u}, \boldsymbol{v})$ is an edge iff $(\boldsymbol{u}, \boldsymbol{v})$ is not an edge of $\boldsymbol{G}$. ( $\bar{G}$ is the complement of $\boldsymbol{G}$.)
We use $\bar{G}$ and $k$ as the instance of Clique.


## Reducing Independent Set to Clique

An instance of Independent Set is a graph $\boldsymbol{G}$ and an integer $\boldsymbol{k}$.
Convert $\boldsymbol{G}$ to $\bar{G}$, in which $(\boldsymbol{u}, \boldsymbol{v})$ is an edge iff $(\boldsymbol{u}, \boldsymbol{v})$ is not an edge of $\boldsymbol{G}$. ( $\bar{G}$ is the complement of $\boldsymbol{G}$.)
We use $\bar{G}$ and $k$ as the instance of Clique.


## Independent Set and Clique

(1) Independent Set $\leq$ Clique.

What does this mean?
(2) If have an algorithm for Clique, then we have an algorithm for Independent Set.
(3) Clique is at least as hard as Independent Set.
(4) Also... Independent Set is at least as hard as Clique.

## Vertex Cover

Given a graph $G=(V, E)$, a set of vertices $S$ is:
(1) A vertex cover if every $e \in E$ has at least one endpoint in $S$.

## Vertex Cover

Given a graph $G=(V, E)$, a set of vertices $S$ is:
(1) A vertex cover if every $e \in E$ has at least one endpoint in $S$.


## Vertex Cover

Given a graph $G=(V, E)$, a set of vertices $S$ is:
(1) A vertex cover if every $e \in E$ has at least one endpoint in $S$.


## Vertex Cover

Given a graph $G=(V, E)$, a set of vertices $S$ is:
(1) A vertex cover if every $e \in E$ has at least one endpoint in $S$.


## The Vertex Cover Problem

## Problem (Vertex Cover)

Input: A graph $G$ and integer $k$.
Goal: Is there a vertex cover of size $\leq k$ in $G$ ?

Can we relate Independent Set and Vertex Cover?

## Relationship between...

## Vertex Cover and Independent Set

## Proposition

Let $G=(V, E)$ be a graph. $S$ is an independent set if and only if $\boldsymbol{V} \backslash \boldsymbol{S}$ is a vertex cover.

## Proof.

$(\Rightarrow)$ Let $S$ be an independent set
(1) Consider any edge $\boldsymbol{u} \boldsymbol{v} \in \boldsymbol{E}$.
(2) Since $\boldsymbol{S}$ is an independent set, either $\boldsymbol{u} \notin \boldsymbol{S}$ or $\boldsymbol{v} \notin \boldsymbol{S}$.
(3) Thus, either $\boldsymbol{u} \in \boldsymbol{V} \backslash \boldsymbol{S}$ or $\boldsymbol{v} \in \boldsymbol{V} \backslash \boldsymbol{S}$.
(0) $\boldsymbol{V} \backslash \boldsymbol{S}$ is a vertex cover.
$(\Leftarrow)$ Let $V \backslash S$ be some vertex cover:
(1) Consider $\boldsymbol{u}, \boldsymbol{v} \in \boldsymbol{S}$
(2) $\boldsymbol{u v}$ is not an edge of G , as otherwise $\boldsymbol{V} \backslash \boldsymbol{S}$ does not cover $\boldsymbol{u} \boldsymbol{v}$.
(3 $\Longrightarrow S$ is thus an independent set.

## Independent Set $\leq_{\mathrm{p}}$ Vertex Cover

(1) $\boldsymbol{G}$ : graph with $\boldsymbol{n}$ vertices, and an integer $\boldsymbol{k}$ be an instance of the Independent Set problem.
(2) $G$ has an independent set of size $\geq k$ iff $G$ has a vertex cover of size $\leq \boldsymbol{n}-\boldsymbol{k}$
(0) $(G, k)$ is an instance of Independent Set, and $(G, n-k)$ is an instance of Vertex Cover with the same answer.
(0) Therefore, Independent Set $\leq_{P}$ Vertex Cover. Also Vertex Cover $\leq_{p}$ Independent Set.

## The Set Cover Problem

## Problem (Set Cover)

Input: Given a set $\boldsymbol{U}$ of $\boldsymbol{n}$ elements, a collection $S_{1}, S_{2}, \ldots S_{m}$ of subsets of $\boldsymbol{U}$, and an integer $\boldsymbol{k}$.
Goal: Is there a collection of at most $k$ of these sets $S_{i}$ whose union is equal to $U$ ?

## The Set Cover Problem

## Problem (Set Cover)

Input: Given a set $\boldsymbol{U}$ of $\boldsymbol{n}$ elements, a collection $S_{1}, S_{2}, \ldots S_{m}$ of subsets of $U$, and an integer $k$.
Goal: Is there a collection of at most $k$ of these sets $S_{i}$ whose union is equal to $U$ ?

## Example

Let $U=\{1,2,3,4,5,6,7\}, k=2$ with

$$
\begin{array}{ll}
S_{1}=\{3,7\} & S_{2}=\{3,4,5\} \\
S_{3}=\{1\} & S_{4}=\{2,4\} \\
S_{5}=\{5\} & S_{6}=\{1,2,6,7\}
\end{array}
$$

## The Set Cover Problem

## Problem (Set Cover)

Input: Given a set $\boldsymbol{U}$ of $\boldsymbol{n}$ elements, a collection $S_{1}, S_{2}, \ldots S_{m}$ of subsets of $U$, and an integer $k$.
Goal: Is there a collection of at most $k$ of these sets $S_{i}$ whose union is equal to $U$ ?

## Example

Let $U=\{1,2,3,4,5,6,7\}, k=2$ with

$$
\begin{array}{ll}
S_{1}=\{3,7\} & S_{2}=\{3,4,5\} \\
S_{3}=\{1\} & S_{4}=\{2,4\} \\
S_{5}=\{5\} & S_{6}=\{1,2,6,7\}
\end{array}
$$

$\left\{S_{2}, S_{6}\right\}$ is a set cover

## Vertex Cover $\leq_{\text {p }}$ Set Cover

Given graph $G=(V, E)$ and integer $k$ as instance of Vertex Cover, construct an instance of Set Cover as follows:

## Vertex Cover $\leq_{\mathrm{p}}$ Set Cover

Given graph $G=(V, E)$ and integer $k$ as instance of Vertex Cover, construct an instance of Set Cover as follows:
(1) Number $k$ for the Set Cover instance is the same as the number $k$ given for the Vertex Cover instance.

## Vertex Cover $\leq_{\mathrm{p}}$ Set Cover

Given graph $G=(V, E)$ and integer $k$ as instance of Vertex Cover, construct an instance of Set Cover as follows:
(1) Number $\boldsymbol{k}$ for the Set Cover instance is the same as the number $k$ given for the Vertex Cover instance.
(2) $U=E$.

## Vertex Cover $\leq_{\mathrm{p}}$ Set Cover

Given graph $G=(V, E)$ and integer $k$ as instance of Vertex Cover, construct an instance of Set Cover as follows:
(1) Number $k$ for the Set Cover instance is the same as the number $k$ given for the Vertex Cover instance.
(2) $U=E$.
(3) We will have one set corresponding to each vertex; $S_{v}=\{e \mid e$ is incident on $v\}$.

## Vertex Cover $\leq_{\text {p }}$ Set Cover

Given graph $G=(V, E)$ and integer $k$ as instance of Vertex Cover, construct an instance of Set Cover as follows:
(1) Number $\boldsymbol{k}$ for the Set Cover instance is the same as the number $k$ given for the Vertex Cover instance.
(2) $U=E$.
(3) We will have one set corresponding to each vertex; $S_{v}=\{e \mid e$ is incident on $v\}$.

Observe that $G$ has vertex cover of size $\boldsymbol{k}$ if and only if $\boldsymbol{U},\left\{S_{v}\right\}_{v \in \boldsymbol{v}}$ has a set cover of size $\boldsymbol{k}$. (Exercise: Prove this.)

## Vertex Cover $\leq_{\mathrm{p}}$ Set Cover: Example



## Vertex Cover $\leq_{\mathrm{p}}$ Set Cover: Example



Let $U=\{a, b, c, d, e, f, g\}$, $k=2$ with

$$
\begin{array}{ll}
S_{1}=\{c, g\} & S_{2}=\{b, d\} \\
S_{3}=\{c, d, e\} & S_{4}=\{e, f\} \\
S_{5}=\{a\} & S_{6}=\{a, b, f, g\}
\end{array}
$$

## Vertex Cover $\leq_{\mathrm{p}}$ Set Cover: Example



Let $U=\{a, b, c, d, e, f, g\}$, $k=2$ with

$$
\begin{array}{ll}
S_{1}=\{c, g\} & S_{2}=\{b, d\} \\
S_{3}=\{c, d, e\} & S_{4}=\{e, f\} \\
S_{5}=\{a\} & S_{6}=\{a, b, f, g\}
\end{array}
$$

$\left\{S_{3}, S_{6}\right\}$ is a set cover
$\{3,6\}$ is a vertex cover

## Proving Reductions

To prove that $X \leq_{P} Y$ you need to give an algorithm $\mathcal{A}$ that:
(1) Transforms an instance $I_{X}$ of $X$ into an instance $I_{\boldsymbol{Y}}$ of $Y$.
(2) Satisfies the property that answer to $\boldsymbol{I}_{\boldsymbol{X}}$ is YES iff $I_{Y}$ is YES.
(1) typical easy direction to prove: answer to $I_{Y}$ is YES if answer to $I_{\boldsymbol{X}}$ is YES
(2) typical difficult direction to prove: answer to $\boldsymbol{I}_{\boldsymbol{X}}$ is YES if answer to $\boldsymbol{I}_{\boldsymbol{Y}}$ is YES (equivalently answer to $\boldsymbol{I}_{\boldsymbol{Y}}$ is NO if answer to $\boldsymbol{I}_{\boldsymbol{X}}$ is NO).
(3) Runs in polynomial time.

## Example of incorrect reduction proof

Try proving Matching $\leq_{p}$ Bipartite Matching via following reduction:
(1) Given graph $G=(V, E)$ obtain a bipartite graph $G^{\prime}=\left(V^{\prime}, E^{\prime}\right)$ as follows.
(0) Let $\boldsymbol{V}_{1}=\left\{\boldsymbol{u}_{1} \mid \boldsymbol{u} \in \boldsymbol{V}\right\}$ and $\boldsymbol{V}_{2}=\left\{\boldsymbol{u}_{2} \mid \boldsymbol{u} \in \boldsymbol{V}\right\}$. We set $\boldsymbol{V}^{\prime}=\boldsymbol{V}_{1} \cup \boldsymbol{V}_{2}$ (that is, we make two copies of $\boldsymbol{V}$ )
(2) $E^{\prime}=\left\{\boldsymbol{u}_{1} \boldsymbol{v}_{2} \mid \boldsymbol{u} \neq \boldsymbol{v}\right.$ and $\left.\boldsymbol{u} \boldsymbol{v} \in E\right\}$
(2) Given $G$ and integer $\boldsymbol{k}$ the reduction outputs $\boldsymbol{G}^{\prime}$ and $\boldsymbol{k}$.

## "Proof"

## Claim

Reduction is a poly-time algorithm. If $G$ has a matching of size $k$ then $G^{\prime}$ has a matching of size $\boldsymbol{k}$.

## Proof.

## Exercise.

## Claim

If $\boldsymbol{G}^{\prime}$ has a matching of size $\boldsymbol{k}$ then $\boldsymbol{G}$ has a matching of size $\mathbf{k}$.

## "Proof"

## Claim

Reduction is a poly-time algorithm. If $\mathbf{G}$ has a matching of size $\boldsymbol{k}$ then $G^{\prime}$ has a matching of size $k$.

## Proof.

## Exercise.

## Claim

If $G^{\prime}$ has a matching of size $k$ then $G$ has a matching of size $k$.
Incorrect! Why?

## "Proof"

## Claim

Reduction is a poly-time algorithm. If $\mathbf{G}$ has a matching of size $\boldsymbol{k}$ then $G^{\prime}$ has a matching of size $\boldsymbol{k}$.

## Proof.

## Exercise.

## Claim

If $\boldsymbol{G}^{\prime}$ has a matching of size $\boldsymbol{k}$ then $\boldsymbol{G}$ has a matching of size $\boldsymbol{k}$.
Incorrect! Why? Vertex $\boldsymbol{u} \in \boldsymbol{V}$ has two copies $\boldsymbol{u}_{1}$ and $\boldsymbol{u}_{2}$ in $\boldsymbol{G}^{\prime}$. A matching in $G^{\prime}$ may use both copies!

## Subset sum and Partition?

## Problem: Subset Sum

Instance: S - set of positive integers, $t$ : - an integer number (target).
Question: Is there a subset
$X \subseteq S$ such that $\sum_{x \in X} x=$ $t$ ?

## Problem: Partition

Instance: A set $S$ of $n$ numbers.
Question: Is there a subset $T \subseteq S$ s.t. $\sum_{t \in T} t=$ $\sum_{s \in S \backslash T} s ?$

Assume that we can solve Subset Sum in polynomial time, then we can solve Partition in polynomial time. This statement is
(A) True.
(B) Mostly true.
(C) False.
(D) Mostly false.

## II: Partition and subset sum?

## Problem: Partition

Instance: A set $S$ of $\boldsymbol{n}$ numbers.
Question: Is there a subset $T \subseteq S$ s.t. $\sum_{t \in T} t=$ $\sum_{s \in S \backslash T} s ?$

## Problem: Subset Sum

Instance: S - set of positive integers, $t$ : - an integer number (target).
Question: Is there a subset $X \subseteq S$ such that $\sum_{x \in X} x=t$ ?

Assume that we can solve Partition in polynomial time, then we can solve Subset Sum in polynomial time. This statement is
(A) True.
(B) Mostly true.
(C) False.
(D) Mostly false.

