## CS173

Introduction to Graph Algorithms

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## Today's material

- Exhaustive search strategies
- Greedy search
- Decision, Optimization, and Construction Problems
- Proving theorems about graphs


## Solving MAX CLIQUE using exhaustive search

Suppose you want to solve MAX CLIQUE.
Given input graph $G=(V, E)$ :

- Enumerate all subsets of $V$
- For each one, determine if it is a clique; if so, record size
- Return size of largest clique found.

Obviously correct, but too expensive. (How expensive?)
This is an example of Exhaustive Search

## Solving MAX CLIQUE using greedy search

Given input graph $G=(V, E)$ :

- Order the vertices $v_{1}, v_{2}, \ldots, v_{n}$
- $A:=\left\{v_{1}\right\}$
- For $i=2$ up to $n$ DO:
- If $A \cup\left\{v_{i}\right\}$ is a clique, then $A:=A \cup\left\{v_{i}\right\}$

Return $A$
Obviously $A$ is a clique, but it may not be maximum.
This is a fast algorithm, but it may not find an optimal solution.
(Class: show such a graph.)
This is an example of a greedy algorithm.

## Solving MAX CLIQUE

How can we solve MAX CLIQUE?

- The exhaustive search strategy is not polynomial time.
- The greedy algorithm is fast but not guaranteed to find an optimal solution.
- What should we do?

The problem is MAX CLIQUE is NP-hard!

## In Class Problem

Suppose you have an algorithm $\mathcal{A}$ that solves the decision problem for MATCHING:

- Input: Graph $G=(V, E)$ and positive integer $k$
- Question: $\exists E_{0} \subseteq E$ such that $\left|E_{0}\right|=k$ and $E_{0}$ is a matching?

Can we make a polynomial number of calls to $\mathcal{A}$ (and a polynomial amount of other operations) to

- find the size of the maximum matching in $G$ ?
- find the largest matching in $G$ ?


## Relationship between decision, optimization, and construction problems

To solve the optimization problem, we define Algorithm $\mathcal{B}$ as follows.
The input is graph $G=(V, E)$. If $E=\emptyset$, we return 0 . Else, we do the following:

- For $k=|E|$ down to $1, \mathrm{DO}$
- If $\mathcal{A}(G, k)=Y E S$, then Return $(k)$

It is easy to see that

- $\mathcal{B}$ is correct,
- that $\mathcal{B}$ calls $\mathcal{A}$ at most $m$ times
- that $\mathcal{B}$ does at most $O(m)$ additional steps.

Hence $\mathcal{B}$ satisfies the desired properties.

## Relationship between decision, optimization, and construction problems

We define Algorithm $\mathcal{C}$ to find a maximum matching, as follows. The input is graph $G=(V, E)$. If $E=\emptyset$, we return $\emptyset$. Otherwise, let $E=\left\{e_{1}, e_{2}, \ldots, e_{m}\right\}$, and let $k=\mathcal{B}(G)$.

- Let $G^{*}$ be a copy of $G$
- For $i=1$ up to $m$ DO
- Let $G^{\prime}$ be the graph obtained by deleting edge $e_{i}$ (but not the endpoints of $e_{i}$ ) from $G^{*}$.
- If $\mathcal{A}\left(G^{\prime}, k\right)=Y E S$, then set $G^{*}:=G^{\prime}$.
- Return the edge set $E\left(G^{*}\right)$ of $G^{*}$.

It is easy to see that $\mathcal{C}$ calls $\mathcal{B}$ once, calls $\mathcal{A}$ at most $m$ times, and does at most $O(m)$ other operations. Hence the running time satisifes the required bounds.
What about accuracy?

## Finding the largest matching in a graph

- Let $G^{*}$ be a copy of $G$
- For $i=1$ up to $m$ DO
- Let $G^{\prime}$ be the graph obtained by deleting edge $e_{i}$ (but not the endpoints of $e_{i}$ ) from $G^{*}$.
- If $\mathcal{A}\left(G^{\prime}, k\right)=Y E S$, then set $G^{*}=G^{\prime}$.

Return the edge set of $G^{*}$.
Notes:

- The edge set returned at the end is a matching (we'll look at this carefully in the next slide).
- We never reduce the size of the maximum matching when we delete edges. Hence, $\mathcal{B}\left(G^{*}\right)=\mathcal{B}(G)$.
- Therefore this algorithm returns a maximum matching.


## Finding the largest matching in a graph

Recall $k$ is the size of a maximum matching in input graph $G$, with edge set $\left\{e_{1}, e_{2}, \ldots, e_{m}\right\}, m \geq 1$.

- Let $G^{*}$ be a copy of $G$
- For $i=1$ up to $m$ DO
- Let $G^{\prime}$ be the graph obtained by deleting edge $e_{i}$ (but not the endpoints of $e_{i}$ ) from $G^{*}$.
- If $\mathcal{A}\left(G^{\prime}, k\right)=Y E S$, then set $G^{*}=G^{\prime}$.

Return the edge set of $G^{*}$.
Theorem: The edge set $E^{*}$ of $G^{*}$ is a matching in $G$.
Proof (by contradiction): If not, then $E^{*}$ has at least two edges $e_{i}$ and $e_{j}$ (both from $E$ ) that share an endpoint. Let $E_{0}$ be a maximum matching in $G^{*}$; hence $E_{0}$ is a maximum matching for $G$. Note that $E_{0}$ cannot include both $e_{i}$ and $e_{j}$. Suppose (w.l.o.g.) $e_{i} \notin E_{0}$. During the algorithm, we checked whether a graph $G^{\prime}$ that did not contain $e_{i}$ had a matching of size $k$. Since we did not delete $e_{i}$, this means the answer was NO. But the edge set of that $G^{\prime}$ contains the matching $E_{0}$, which means $G^{\prime}$ has a matching of size $k$, yielding a contradiction.

## Reductions

- We used an algorithm $\mathcal{A}$ for decision problem $\pi$ to solve an optimization or construction problem $\pi^{\prime}$ on the same input. We also required that we call $\mathcal{A}$ at most a polynomial number of times, and that we do at most a polynomial number of other operations.
- This means that if $\mathcal{A}$ runs in polynomial time, then we have a polynomial time algorithm for both $\pi$ and $\pi^{\prime}$. Note that we use two things here: $\mathcal{A}$ is polynomial, and the input did not change in size.
- What we did isn't really a Karp reduction, because Karp reductions are only for decision problems... but the ideas are very related.
- If you can understand why this works, you will understand why Karp reductions have to satisfy what they satisfy.

Just try to understand the ideas. This is not about memorization.

## Complements of graphs

Let $G=(V, E)$ be a graph
The graph $G^{c}$ contains the same vertex set, but only contains the missing edges (though not the self-loops), and is referred to as the complement of $G$.

- Suppose $V_{0}$ is a clique in $G$. What can you say about $V_{0}$ in $G^{c}$ ?
- Suppose $V_{0}$ is an independent set in $G$. What can you say about $V_{0}$ in $G^{c}$ ?


## Complements of sets

Suppose $V_{0}$ is an independent set in $G$. What can you say about $G \backslash V_{0}$ ?

Suppose $V_{0}$ is a clique in $G$. What can you say about $V \backslash V_{0}$ ?

Suppose $V_{0}$ is a vertex cover in $G$. What can you say about $V \backslash V_{0}$ ?

## Manipulating Graphs

## Adding vertices to graphs:

Suppose you add a vertex $v$ to $G$ and make $v$ adjacent to every vertex in $V$.

Let the new graph be called $G^{\prime}$.
How do these values change between $G$ and $G^{\prime}$ ?

- the size of the maximum clique,
- the size of the maximum independent set,
- the chromatic number


## Things to think about

- Suppose $G$ is a simple graph that has a maximum matching of size $k$ and a minimum vertex cover of size $k^{\prime}$. Prove that $k^{\prime} \geq k$.
- Prove that every tree can be 2-colored.
- Prove that every tree with at least two three vertices has a sibling pair of leaves (where two leaves are siblings if they share a neighbor).
- Come up with a simple algorithm to find a maximal matching (i.e., a matching that cannot be enlarged by adding another edge) in a graph, and analyze its running time.
- Show how having an algorithm to compute the chromatic number in a graph can be used to find an optimal vertex coloring for a graph, with only a polynomial number of calls to the algorithm.

