Chapter 99

Review session

CS 473: Fundamental Algorithms, Spring 2013

February 19, 2013

99.0.0.1 Why Graphs?

- (A) Graphs help model networks which are ubiquitous: transportation networks (rail, roads, airways), social networks (interpersonal relationships), information networks (web page links) etc etc.
- (B) Fundamental objects in Computer Science, Optimization, Combinatorics
- (C) Many important and useful optimization problems are graph problems
- (D) Graph theory: elegant, fun and deep mathematics

99.0.0.2 Basic Graph Search

Given G = (V, E) and vertex $u \in V$:

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Explore(u):

Initialize S = \{u\}

while there is an edge (x, y) with x \in S and y \notin S do

add y to S
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99.0.0.3 DFS in Directed Graphs

$\mathbf{DFS}(G)$	D	$\mathbf{FS}(u)$
Mark all nodes u as unvisited		Mark u as visited
T is set to \emptyset		$\operatorname{pre}(u) = + + time$
time = 0		for each edge (u,v) in $Out(u)$ do
\mathbf{while} there is an unvisited node u (lo	$\mathbf{if}\;v$ is not marked
$\mathbf{DFS}(u)$		add edge (u,v) to T
		$\mathbf{DFS}(v)$
Output T		post(u) = + + time

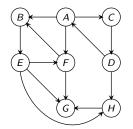
99.0.0.4 pre and post numbers

Node u is **active** in time interval [pre(u), post(u)]

Proposition 99.0.1. For any two nodes u and v, the two intervals [pre(u), post(u)] and [pre(v), post(v)] are disjoint or one is contained in the other.

99.0.0.5 Connectivity and Strong Connected Components

Definition 99.0.2. Given a directed graph G, u is strongly connected to v if u can reach v and v can reach u. In other words $v \in rch(u)$ and $u \in rch(v)$.



99.0.0.6 Directed Graph Connectivity Problems

- (A) Given G and nodes u and v, can u reach v?
- (B) Given G and u, compute rch(u).
- (C) Given G and u, compute all v that can reach u, that is all v such that $u \in \operatorname{rch}(v)$.
- (D) Find the strongly connected component containing node u, that is SCC(u).
- (E) Is G strongly connected (a single strong component)?
- (F) Compute all strongly connected components of G.

First four problems can be solve in O(n + m) time by adapting **BFS/DFS** to directed graphs. The last one requires a clever **DFS** based algorithm.

99.0.0.7 DFS Properties

Generalizing ideas from undirected graphs:

- (A) DFS(u) outputs a directed out-tree T rooted at u
- (B) A vertex v is in T if and only if $v \in \operatorname{rch}(u)$
- (C) For any two vertices x, y the intervals [pre(x), post(x)] and [pre(y), post(y)] are either disjoint are one is contained in the other.
- (D) The running time of DFS(u) is O(k) where $k = \sum_{v \in rch(u)} |Adj(v)|$ plus the time to initialize the Mark array.
- (E) **DFS**(G) takes O(m + n) time. Edges in T form a disjoint collection of out-trees. Output of DFS(G) depends on the order in which vertices are considered.

99.0.0.8 DFS Tree

Edges of G can be classified with respect to the **DFS** tree T as:

- (A) **Tree edges** that belong to T
- (B) A *forward edge* is a non-tree edges (x, y) such that pre(x) < pre(y) < post(y) < post(x).
- (C) A **backward edge** is a non-tree edge (x, y) such that pre(y) < pre(x) < post(x) < post(y).
- (D) A cross edge is a non-tree edges (x, y)such that the intervals [pre(x), post(x)]and [pre(y), post(y)] are disjoint.

99.0.0.9 Algorithms via DFS

 $SC(G, u) = \{v \mid u \text{ is strongly connected to } v\}$

(A) Find the strongly connected component containing node u. That is, compute SCC(G, u). $SCC(G, u) = rch(G, u) \cap rch(G^{rev}, u)$

Hence, SCC(G, u) can be computed with two **DFS**es, one in G and the other in G^{rev} . Total O(n+m) time.

99.0.1 Linear Time Algorithm

99.0.1.1 ... for computing the strong connected components in G

```
do DFS(G^{rev}) and sort vertices in decreasing post order.
Mark all nodes as unvisited
for each u in the computed order do
if u is not visited then
DFS(u)
Let S_u be the nodes reached by u
Output S_u as a strong connected component
Remove S_u from G
```

Analysis Running time is O(n+m). (Exercise)

Example: Makefile

99.0.1.2 BFS with Distances

```
\begin{aligned} \textbf{BFS}(s) \\ & \text{Mark all vertices as unvisited and for each } v \text{ set } \operatorname{dist}(v) = \infty \\ & \text{Initialize search tree } T \text{ to be empty} \\ & \text{Mark vertex } s \text{ as visited and set } \operatorname{dist}(s) = 0 \\ & \text{set } Q \text{ to be the empty queue} \\ & \textbf{enq}(s) \\ & \textbf{while } Q \text{ is nonempty } \textbf{do} \\ & u = \textbf{deq}(Q) \\ & \textbf{for each vertex } v \in \operatorname{Adj}(u) \text{ do} \\ & \textbf{if } v \text{ is not visited } \textbf{do} \\ & \text{ add edge } (u, v) \text{ to } T \\ & \text{Mark } v \text{ as visited, } \textbf{enq}(v) \\ & \text{ and set } \operatorname{dist}(v) = \operatorname{dist}(u) + 1 \end{aligned}
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Proposition 99.0.3. BFS(s) runs in O(n+m) time.

99.0.1.3 BFS with Layers

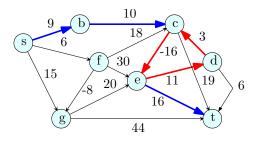
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\begin{aligned} \textbf{BFSLayers}(s): \\ & \text{Mark all vertices as unvisited and initialize } T \text{ to be empty} \\ & \text{Mark } s \text{ as visited and set } L_0 = \{s\} \\ & i = 0 \\ & \textbf{while } L_i \text{ is not empty } \textbf{do} \\ & \text{ initialize } L_{i+1} \text{ to be an empty list} \\ & \textbf{for each } u \text{ in } L_i \text{ do} \\ & \textbf{for each edge } (u, v) \in \text{Adj}(u) \text{ do} \\ & \text{ if } v \text{ is not visited} \\ & \text{ mark } v \text{ as visited} \\ & \text{ add } (u, v) \text{ to tree } T \\ & \text{ add } v \text{ to } L_{i+1} \\ & i = i+1 \end{aligned}
```

Running time: O(n+m)

99.0.2 Checking if a graph is bipartite...

99.0.2.1 Linear time algorithm

Corollary 99.0.4. There is an O(n+m) time algorithm to check if G is bipartite and output an odd cycle if it is not.



99.0.2.2 Dijkstra's Algorithm

Initialize for each node v,
$$\operatorname{dist}(s, v) = \infty$$

Initialize $S = \{s\}$, $\operatorname{dist}(s, s) = 0$
for $i = 1$ to $|V|$ do
Let v be such that $\operatorname{dist}(s, v) = \min_{u \in V-S} \operatorname{dist}(s, u)$
 $S = S \cup \{v\}$
for each u in $\operatorname{Adj}(v)$ do
 $\operatorname{dist}(s, u) = \min(\operatorname{dist}(s, u), \operatorname{dist}(s, v) + \ell(v, u))$

- (A) Using Fibonacci heaps. Running time: $O(m + n \log n)$.
- (B) Can compute shortest path tree.

99.0.2.3 Single-Source Shortest Paths with Negative Edge Lengths

Single-Source Shortest Path Problems Input: A directed graph G = (V, E)with arbitrary (including negative) edge lengths. For edge $e = (u, v), \ \ell(e) = \ell(u, v)$ is its length.

- Given nodes *s*, *t* find shortest path from *s* to *t*.
- Given node *s* find shortest path from *s* to all other nodes.

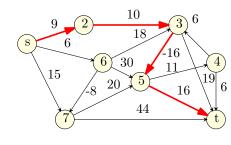
99.0.2.4 Negative Length Cycles

Definition 99.0.5. A cycle C is a negative length cycle if the sum of the edge lengths of C is negative.

99.0.2.5 A Generic Shortest Path Algorithm

Dijkstra's algorithm does not work with negative edges.

 $\begin{aligned} & \textbf{Relax} \left(e = (u, v) \right) \\ & \textbf{if} \ \left(d(s, v) > d(s, u) + \ell(u, v) \right) \ \textbf{then} \\ & d(s, v) = d(s, u) + \ell(u, v) \end{aligned}$



GenericShortestPathAlg: d(s,s) = 0for each node $u \neq s$ do $d(s,u) = \infty$ while there is a tense edge do Pick a tense edge eRelax(e) Output d(s, u) values

99.0.2.6 Bellman-Ford to detect Negative Cycles

```
for each u \in V do

d(s, u) = \infty

d(s, s) = 0

for i = 1 to |V| - 1 do

for each edge e = (u, v) do

Relax(e)

for each edge e = (u, v) do

if e = (u, v) is tense then

Stop and output that s can reach

a negative length cycle

Output for each u \in V: d(s, u)
```

- (A) Total running time: O(mn).
- (B) Can detect negative cycle reachable from s.
- (C) Appropriate construction detect any negative cycle in a graph.

99.0.3 Shortest paths in DAGs

99.0.3.1 Algorithm for DAGs

```
ShorestPathInDAG(G, s):

s = v_1, v_2, v_{i+1}, \dots, v_n be a topological sort of G

for i = 1 to n do

d(s, v_i) = \infty

d(s, s) = 0

for i = 1 to n - 1 do

for each edge e in Adj(v_i) do

Relax(e)

return d(s, \cdot) values computed
```

Running time: O(m+n) time algorithm! Works for negative edge lengths and hence can find *longest* paths in a **DAG**.

99.0.3.2 Reduction

Reducing problem A to problem B:

- (A) Algorithm for A uses algorithm for B as a *black box*.
- (B) Example: Uniqueness (or distinct element) to sorting.

99.0.3.3 Recursion

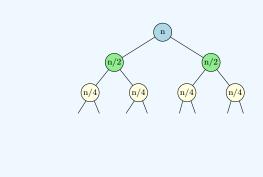
- (A) Recursion is a very powerful and fundamental technique.
- (B) Basis for several other methods.
 - (A) Divide and conquer.
 - (B) Dynamic programming.
 - (C) Enumeration and branch and bound etc.
 - (D) Some classes of greedy algorithms.
- (C) Recurrences arise in analysis.

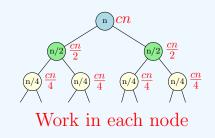
Examples seen:

- (A) Recursion: Tower of Hanoi, Selection sort, Quick Sort.
- (B) Divide & Conquer:
 - (A) Merge sort.
 - (B) Multiplying large numbers.

99.0.4 Solving recurrences using recursion trees

99.0.4.1 An illustrated example: Merge sort...

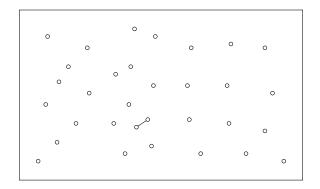




99.0.5 Solving recurrences

99.0.5.1 The other "technique" - guess and verify

- (A) Guess solution to recurrence.
- (B) Verify it via induction. Solved in class:



(A) $T(n) = 2T(n/2) + n/\log n$. (B) $T(n) = T(\sqrt{n}) + 1$. (C) $T(n) = \sqrt{n}T(\sqrt{n}) + n$. (D) T(n) = T(n/4) + T(3n/4) + n

99.0.5.2 Closest Pair - the problem

Input Given a set S of n points on the plane

Goal Find $p, q \in S$ such that d(p, q) is minimum

Algorithm:

One can compute closest pair points in the plane in $O(n \log n)$ time using divide and conquer.

99.0.5.3 Median selection

Problem

Given list L of n numbers, and a number k find kth smallest number in n.

- (A) Quick Sort can be modified to solve it (but worst case running time is quadratic (if lucky linear time).
- (B) Seen divide & conquer algorithm... Involved, but linear running time.

99.0.6 Recursive algorithm for Selection

99.0.6.1 A feast for recursion

```
 \begin{array}{l} \textbf{select}(A, \ j):\\ n = |A|\\ \textbf{if} \ n \leq 10 \ \textbf{then}\\ \quad & \text{Compute } j\textbf{th} \ \textbf{smallest element in } A \ \textbf{using brute force.}\\ \text{Form lists } L_1, L_2, \ldots, L_{\lceil n/5\rceil} \ \textbf{where } L_i = \{A[5i-4], \ldots, A[5i]\}\\ \text{Find median } b_i \ \textbf{of each } L_i \ \textbf{using brute-force}\\ B \ \textbf{is the array of } b_1, b_2, \ldots, b_{\lceil n/5\rceil}.\\ b = \textbf{select}(B, \ \lceil n/10\rceil)\\ \text{Partition } A \ \textbf{into } A_{\texttt{less or equal}} \ \textbf{and } A_{\texttt{greater}} \ \textbf{using } b \ \textbf{as pivot}\\ \textbf{if } |A_{\texttt{less or equal}}| = j \ \textbf{then}\\ \ \textbf{return } b\\ \textbf{if } |A_{\texttt{less or equal}}| > j) \ \textbf{then}\\ \ \textbf{return select}(A_{\texttt{less or equal}}, \ j)\\ \textbf{else}\\ \ \textbf{return select}(A_{\texttt{greater}}, \ j - |A_{\texttt{less or equal}}|) \end{aligned}
```

99.0.6.2 Back to Recursion

Seen some simple recursive algorithms:

- (A) Binary search.
- (B) Fast exponentiation.
- (C) Fibonacci numbers.
- (D) Maximum weight independent set.