

The background features a large, light blue watermark logo of the Technical University of Cluj-Napoca. The logo consists of a shield with a stylized 'T' and 'U' inside, and the text 'TECHNICAL UNIVERSITY' at the top, 'OF CLUJ-NAPOCA' in the middle, and 'Computer Science' at the bottom.

Fundamental Algorithms

Lecture #11

Cluj-Napoca

December, 11, 2019

Agenda

- **Graphs**
 - **BFS - conclusions**
 - **DSF**
 - **Single Source Shortest Path Dijkstra**

Graphs - BFS

- Partition of the vertex set:
 - Unvisited - white (all in the beginning; white- \rightarrow grey at EnQueue)
 - Under visitation - grey; forms the boundary; are all in the Queue
 - Visited - black (all in the end; grey \rightarrow black at DeQueue)
 - Builds a BFS tree
 - By storing the parent
 - static representation,
 - NOT a dynamic one; π is an array)
- $G_{\pi}(V_{\pi}, E_{\pi})$
- $$V_{\pi} = \{v \mid v \in V, \pi[v] \neq \text{nil}\} \cup \{s\}$$
- $$E_{\pi} = \{(v, \pi[v]) \mid (v, \pi[v]) \in E, v \in V_{\pi} - \{s\}\}$$
- It is a forest of trees (in case the graph is not connected)
 - The algorithm needs to restart from every connected component (wrapper over BFS)
 - Runs linearly in the size of the graph ($O(|V| + |E|)$)

BFS – the algorithm

bfs (G, s)

```

for each  $u \in V(G) - \{s\}$                                 //initialization step
  do color  $[u] \leftarrow \text{white}$ 
    d $[u] \leftarrow \infty$                                 //distance from source
     $\pi[u] \leftarrow \text{nil}$                                 //parent node
color $[s] \leftarrow \text{grey}$                                 //initialize source
d $[s] \leftarrow 0$ 
 $\pi[s] \leftarrow \text{nil}$ 
Q  $\leftarrow \{s\}$                                           //initialize Queue (FIFO policy)
while Q  $\neq \emptyset$                                     //as long as still have discovered vertices
  do u  $\leftarrow \text{head } [Q]$                             //u – first from the Q; was NOT removed from Q
    for each  $v \in \text{Adj}[u]$                               //take all the neighbor vertices
      do if color $[v] = \text{white}$  //only from outside the frontier
        then color $[v] \leftarrow \text{grey}$ 
          d $[v] \leftarrow d[u] + 1$ 
           $\pi[v] \leftarrow u$ 
          EnQ (Q, v)
    DeQ (Q)
color $[u] \leftarrow \text{black}$ 

```

Graphs - DFS

- Similar to BFS, the queue is replaced by a **stack**
- (or FIFO policy by LIFO policy) – yet the recursive version more powerful – to discuss
- Keeps the same **color representation** (with similar meaning of colors)
- They (the color clusters) define a vertical boundary between visited/unvisited vertices
- Partition of the vertex set:
 - Unvisited - white (all in the beginning; white->grey at Push)
 - Under visitation - grey; forms the boundary; are all in the Stack
 - Visited - black (all in the end; grey -> black at Pop)

Graphs – **DFS** – additional data

- Keeps the same π - parent node \Rightarrow produces a tree as output, **df-tree** with S as root (also with static representation)
- Add 2 new items representing **time stamps**:
 - $d[u]$ = discovery time
 - = moment when dfs **first reaches** the vertex u
 - $f[u]$ = finishing time
 - = moment when dfs **last releases** the vertex u, after visiting ALL its (unvisited) neighbors
- Both types (π and d/f) informative. Lots of properties



DFS – the algorithm

dfs (**G**, **s**)

```
for each  $u \in V(G)$  //initialization step
  do color[u] <- white
     $\pi[u]$  <- nil //parent node
time <- 0 //global variable; keeps track of time evol.
for each vertex  $u \in V[G]$  //take all the neighbor vertices
  do if color[u] = white //take only unvisited nodes
    then dfs_visit(u)

dfs_visit (u)
color[u] <- grey //say u is under visiting process
d[u] <- time //just discovered
time <- time+1
for each  $v \in \text{Adj}[u]$  //take all the neighbor vertices
  do if color[v] = white //only unvisited
    then  $\pi[v]$  <- u //rec call
      dfs_visit(v)
color[v] <- black //release the node
f[u] <- time
time <- time+1
```

DFS - analysis

- Initialization steps: $O(V)$
- The adjacency list of each vertex is scanned only once (when the vertex reaches the top of the stack): $O(E)$
- The dfs alg: $O(V+E)$ (means $O(|V|+|E|)$)
- The predecessor subgraph ($\pi[s]$) forms a tree /forest of trees (what is forest? When tree/forest?)

$$G_{\pi}(V_{\pi}, E_{\pi})$$

$$V_{\pi} = \{v \mid v \in V, \pi[v] \neq \text{nil}\} \cup \{s\}$$

$$E_{\pi} = \{(v, \pi[v]) \mid (v, \pi[v]) \in E, v \in V_{\pi} - \{s\}\}$$

DFS – parenthesis theorem

Th: $G(V, E), \forall u, v \in V$

In any dfs of a directed or undirected graph, time intervals of u, v are:

(1) either entirely disjoint

$$d[u] < f[u] < d[v] < f[v]$$

(2) or one is entirely contained in the other one

$$d[u] < d[v] < f[v] < f[u]$$

v descendant of u

v discovered while u was grey

DFS – edges characterization

- Edge characterization – basis for many other processes (ex. Topological sort, cycles detection, ...)
- Edges (u, v) visited in the dfs
 - \Rightarrow u always grey (already on the stack)
 - v various colors
- How could we create 4 categories out of 3 colors?
- **Tree edges**
- **Back edges**
- **Forward edges**
- **Cross edges**

DFS – edges characterization

- **Tree edges**
 - $u = \text{grey}; v = \text{white};$
 - $d[u] < d[v] < f[v] < f[u]$ (intervals included)
- **Back edges**
 - $u = \text{grey}; v = \text{grey}$
 - $d[v] < d[u] < f[u] < f[v]$ (**u inner v**)
 - link the same branch of the tree backwards =>
 - **They produce cycle**
- **Forward edges**
 - $u = \text{grey}; v = \text{black}$
 - $d[u] < d[v] < f[v] < f[u]$ (**v inner u**)
 - link the same branch of the tree forward =>
 - Does NOT produce cycles

DFS – edges characterization

- **Forward edges**
 - $u = \text{grey}; \mathbf{v = black}$
 - $d[u] < d[v] < f[v] < f[u]$ (**v inner u**)
 - link the same branch of the tree forward \Rightarrow
 - Does NOT produce cycle
- **Cross edges**
 - $u = \text{grey}; \mathbf{v = black}$ (i.e. visit of v finished)
 - $d[v] < f[v] < d[u] < f[u]$ (external times: **() ()**)
 - Does NOT produce cycles
 - Link (Q: how can we differentiate among them?)
 - Different branches of the same tree
 - Different trees of the dfs forest (Q: what's the forest?)

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DFS – Benefits

- Produces the *tree structure* (or forest of trees; what's this? Oral discussion) as output in linear time (consider further discussions /seminars for reaching a linear structure as output also in linear time)
- *Time stamps* – have the ability to identify other properties
- *Edge characterization* (many benefits; the immediate one = cycle detection)
 - Cycle detection how? In directed graphs? In undirected graphs? Oral discussion.
- *Parentheses properties*
- DFS may be used as:
 - Skeleton for other algorithms
 - Preprocessing step for other strategies

Single Source Shortest Path the problem - Dijkstra

- $G = (V, E), w(u, v), \forall u, v \in V, w: E \rightarrow \mathbb{R}$
- Define a path from source (v_0) to **all** other vertices in the graph: $v_0 \rightarrow v_k$
 - Path: $p = \langle v_0, v_1, \dots, v_k \rangle$
 - The weight of the path: $w(p) = \sum w(v_{i-1}, v_i), i=1, k$
 - For each p from v_0 , find $\min w(p)$, denoted δ
- Notation:

$$\delta(u, v) = \begin{cases} \min \{w(p): \exists p \text{ } u \rightarrow v\} \\ \infty \quad \sim \exists p \text{ } u \rightarrow v \end{cases}$$
- Shortest path from u to v : $w(p) = \delta(u, v)$

Single Source Shortest Path technique

- Greedy
 - Does NOT guarantee optimality (known fact). Same as in MST.
 - Often used for optimization problems (with additional, specific constraints added + proof)
 - Here, the **optimal structure** template
 - Remember Optimal Structure

SSSP- technique

Optimal Structure discussion

Dynamic programming - Based on the **optimal structure**

- Reusing instead of recomputing (it pays space to gain time)
- Optimization problems
- Solution to the problem – composition of solutions (of identical problems) on subdomains
- Is it divide et impera? Why not? Oral explanation
- Structure of the problem (Optimal structure): a problem exhibits **optimal substructure** if an optimal solution to the problem contains within it optimal solutions to subproblems
- Solution optimal \Rightarrow solution to subproblems **are** optimal (proof by contradiction)
- solution to subproblems optimal \Rightarrow Solution optimal ?
- NO! WHY? (counterexample)

Notes on \Rightarrow

- Conditional statement $(p \Rightarrow q)$ is equivalent to its contrapositive $(\sim q \Rightarrow \sim p)$
- Conditional statement $(p \Rightarrow q)$ is **NOT** equivalent to its inverse $(q \Rightarrow p)$
- $(p \Rightarrow q) \Leftrightarrow (\sim q \Rightarrow \sim p)$

BUT

- $(p \Rightarrow q) \not\Leftrightarrow (q \Rightarrow p)$
- Check with truth tables!

SSSP- technique

Optimal Structure discussion

- Dynamic programming - Based on the **optimal structure**
 - Solution to the problem – composition of solutions (of identical problems) on subdomains
 - Solution optimal \Rightarrow solution to subproblems **are** optimal

SSSP

data and functions used

- $d[v]$ = distance from source to v
- $\pi[v]$ = parent of v on the shortest path from source to v

initialize_single_source (G, s)

for each $v \in V(G)$

do $d[v] \leftarrow \infty$

$\pi[v] \leftarrow \text{nil}$

$d[s] \leftarrow 0$

SSSP

Relaxation – the technique

- Process that tries to *improve* the shortest path found so far
- Done by taking into account along the path a **new node u** , not yet considered
- Update (**relax**) the path (with parent), if necessary (blackboard)
- Update occurs **only** in case
 - the cost of the path to v so far ($d[v]$) is $>$
 - than the cost of path to u and (u,v) 'weight: $d[u] + w(u, v)$

SSSP

Relax – the function

Relax (**u**, **v**, **w**) //relaxes the cost of the path
//from source to v by considering the **new** vertex u

if $d[v] > d[u] + w(u, v)$

then $d[v] \leftarrow d[u] + w(u, v)$

$\pi[v] \leftarrow u$

Update the shortest path from source to v

- if former path better, leave it as it is
- else, in case the new vertex (u) makes any improvement, take the route containing u

Note: a similar techniques was used by Prim's

SSSP Dijkstra's strategy (more than a simple algorithm)

Dijkstra (G, w, s) //1959

initialize_single_source (G, s)

$S \leftarrow \emptyset$

$Q \leftarrow V[G]$ //distances updated to edges, where exist

while $Q \neq \emptyset$

do $u \leftarrow \text{extract_min}(Q)$

$S \rightarrow S \cup \{u\}$

for each vertex $v \in \text{Adj}[u]$

do relax (u, v, w)

Single Source Shortest Path

Dijkstra's efficiency

- The original (initial version, in 59) alg. does not use a priority Q, just a regular Q; Analysis for priority Qs
- Q = linear array $O(V^2)$
 - Initial step $O(V)$
 - Extract min $O(V)$, V times, so $O(V^2)$
 - Each u is added once to S; each edge is inspected once => for loop takes $O(E)$
- Q = binary heap $O(E \lg V)$
 - Extract min $O(\lg V)$, V times, so $O(V \lg V)$
 - Build the initial heap (initial Q) $O(V)$
 - Relax = decrease key takes $O(\lg V)$ applied once to each edge, so $O(E \lg V)$
- Q = Fibonacci heap $O(V \lg V + E)$
 - homework

Single Source Shortest Path

Dijkstra's – trace + proof

- Trace – blackboard
- Proof of correctness
 - Informal discussions
 - Relies on the property that subpaths of shortest path are shortest paths as well (optimal substructure)
 - Total Correctness: termination of the algorithm (empty queue)
+ Partial correctness with Induction on $\delta(u, v)$.
 - Check the textbook (for the formal approach)

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