

MA/CSSE 473

Day 39

Disjoint Set
Implementation

Dijkstra



MA/CSSE 473 Day 39

- HW 15 Due Friday (it's a little different!)
- Fill out the Course evaluation form
 - If everybody in a section does it, everyone in that section gets 10 bonus points on the Final Exam
 - I can't see **who** has completed the evaluation, but I can see **how many (current: 5/16, 5/23)**
- Final Exam Wednesday evening, Nov 17
- **Student Questions**
- Prim data structures and detailed algorithm
- Kruskal data structures and detailed algorithm



Final Exam Format

- **Part 1:** You may use textbook, two 2-sided 8.5"x11" sheets of paper, calculator
- **Part 2:** You may also use your homework assignments, notes, and quizzes, things I have posted for the course, programs that you have written
- Receive both parts at the beginning of the exam; work on questions in any order, turn in part 1 before using additional part 2 resources



Disjoint Set Representation

- Each disjoint set is a tree, with the "marked" element as its root
- Efficient representation of the trees:
 - an array called *parent*
 - $parent[i]$ contains the index of i 's parent.
 - If i is a root, $parent[i]=i$



Using this representation

- `makeiset(i):`

```
def makeiset1(i):  
    parent[i] = i
```
 - `findset(i):`

```
def findset1(i):  
    while i != parent[i]:  
        i = parent[i]  
    return i
```
 - `mergetrees(i,j):`
 - assume that i and j are the marked elements from different sets.
 - `union(i,j):`

```
def mergetrees1(i, j):  
    parent[i] = j
```

 - assume that i and j are elements from different sets
- ```
def union1(i, j):
 mergetrees1(findset1(i), findset1(j))
```



## Analysis

- Assume that we are going to do n makeset operations followed by m union/find operations
- time for makeset?
- worst case time for findset?
- worst case time for union?
- Worst case for all m union/find operations?
- worst case for total?
- What if  $m < n$ ?
- Write the formula using the min function



## Can we keep the trees from growing so fast?

- Make the shorter tree the child of the taller one
- What do we need to add to the representation?
- rewrite makeset, mergetrees.

```
def makeset2(i):
 parent[i] = i
 height[i] = 0
```

```
def mergetrees2(i, j):
 if height[i] < height[j]:
 parent[i] = j
 elif height[i] > height[j]:
 parent[j] = i
 else:
 parent[i] = j
 height[j] = height[j] + 1
```

- findset & union are unchanged.
- What can we say about the maximum height of a k-node tree?



## Theorem: max height of a k-node tree T produced by these algorithms is $\lfloor \lg k \rfloor$

- Base case...
- Induction hypothesis...
- Induction step:
  - Let T be a k-node tree
  - T is the union of two trees:
    - $T_1$  with  $k_1$  nodes and height  $h_1$
    - $T_2$  with  $k_2$  nodes and height  $h_2$
  - What can we say about the heights of these trees?
  - Case 1:  $h_1 \neq h_2$ . Height of T is
  - Case 2:  $h_1 = h_2$ . WLOG Assume  $k_1 \geq k_2$ . Then  $k_2 \leq k/2$ . Height of tree is  $1 + h_2 \leq \dots$



Q2

## Worst-case running time

- Again, assume  $n$  makeset operations, followed by  $m$  union/find operations.
- If  $m > n$
- If  $m < n$



## Speed it up a little more

- **Path compression:** Whenever we do a findset operation, change the parent pointer of each node that we pass through on the way to the root so that it now points directly to the root.
- Replace the **height** array by a **rank** array, since it now is only an upper bound for the height.
- Look at makeset, findset, mergetrees (on next slides)



Q3

## Makeset

This algorithm represents the set  $\{i\}$  as a one-node tree and initializes its rank to 0.

```
def makeset3(i):
 parent[i] = i
 rank[i] = 0
```



## Findset

- This algorithm returns the root of the tree to which  $i$  belongs and makes every node on the path from  $i$  to the root (except the root itself) a child of the root.

```
def findset(i):
 root = i
 while root != parent[root]:
 root = parent[root]
 j = parent[i]
 while j != root:
 parent[i] = root
 i = j
 j = parent[i]
 return root
```



## Mergetrees

This algorithm receives as input the roots of two distinct trees and combines them by making the root of the tree of smaller rank a child of the other root. If the trees have the same rank, we arbitrarily make the root of the first tree a child of the other root.

```
def mergetrees(i, j) :
 if rank[i] < rank[j]:
 parent[i] = j
 elif rank[i] > rank[j]:
 parent[j] = i
 else:
 parent[i] = j
 rank[j] = rank[j] + 1
```



## Analysis

- It's complicated!
- R.E. Tarjan proved (1975)\*:
  - Let  $t = m + n$
  - Worst case running time is  $\Theta(t \alpha(t, n))$ , where  $\alpha$  is a function with an *extremely* slow growth rate.
  - Tarjan's  $\alpha$ :
    - $\alpha(t, n) \leq 4$  for all  $n \leq 10^{19728}$
- Thus the amortized time for each operation is essentially constant time.

\* According to *Algorithms* by R. Johnsonbaugh and M. Schaefer, 2004, Prentice-Hall, pages 160-161



Q4

## Dijkstra's Algorithm

- For a selected vertex, (call it **start**) in a connected, weighted graph  $G$ , find a shortest (minimum total weight) path from *start* to each other vertex
- Assumption: All weights are positive



Q5

## Dijkstra's algorithm approach

- The shortest path from vertex  $v$  to vertex  $w$  is either
  - empty, if  $v = w$ , or
  - obtained by adding an edge  $(u, w)$  to the end of a shortest path from  $v$  to  $u$ .



## Dijkstra's algorithm start-up

- Begin with a subgraph that consists of only the starting vertex.
- Among all edges  $(start, v)$ , choose the one with the smallest weight, and add it (along with the edge that connects it) to our subgraph. Clearly  $(start, v)$  is the shortest path from  $start$  to  $v$ .



## Dijkstra's algorithm continues

- Loop:  
Among all vertices not in a path yet, and which are also adjacent to a vertex that is in the path, choose one with minimal path length from start, and add it to the path, along with the edge that connects it to a path.
- Which algorithm that we have seen does this closely resemble?

Example:

```
g = AdjacencyListGraph(
 [[1, [(2, 4), (3, 2), (5, 3)]],
 [2, [(1, 4), (4, 5)]],
 [3, [(1, 2), (5, 6), (4, 1), (6, 3)]],
 [4, [(2, 5), (6, 6), (3, 1)]],
 [5, [(1, 3), (3, 6), (6, 2)]],
 [6, [(5, 2), (3, 3), (4, 6)]]
])
```



Q6

## Dijkstra's algorithm data structures

- A parent array as in Prim's algorithm.
- A minheap that stores "unpathed" nodes as keys and minimum path lengths (as extensions of a minimum path that has already been discovered) as weights.
  - Use an indirect binary heap, just as we did for Prim.
- A cost array that stores the minimum path length to each vertex
  - Not strictly necessary; could reconstruct from adjacency list and parent array.



## Dijkstra's algorithm details

```
def dijkstra(adj, start):
 """ parent[v] = parent of v in shortest path to v rooted at start """
 n = adj.length() # vertices in graph
 key = [None] + [INFINITY]*n # later they will be decreased
 parent = [None] + [0]*n
 cost = [None] + [0]*n
 key[start] = 0
 heap = MinHeap(key) # non-infinity value in heap represents fringe vertex
 for i in range(1, n+1):
 minCost = heap.minKey()
 v = heap.delete()
 edges = adj.getList(v) # all vertices adjacent to v
 for edge in edges: # an edge is a list of: other vertex and weight
 w = edge[VERTEX]
 if heap.isIn(w) and minCost + edge[WEIGHT] < heap.keyVal(w):
 parent[w] = v
 cost[w] = minCost + edge[WEIGHT]
 heap.decrease(w, cost[w])

 return parent, cost
```



Q7-8