

Lecture 8 | Part 1

**Today's Lecture** 

## **Disjoint Sets**

- Often need to keep a collection of disjoint sets.
   Example: {{4, 6, 2, 0}, {1, 3}, {5}}
- May need to union disjoint sets.
- May need to check if two items are in same set.

#### **Use Case**

- We are given a stream of nodes, edges.
- Want to keep track of CCs at every step.
- BFS/DFS take O(V + E) time; efficient to compute CCs once, but then need to recompute.

#### **Use Cases**

- Used in Kruskal's algorithm for MST.
- Used in single linkage clustering.
- Used in Tarjan's algorithm to find LCA in a tree.

### **Disjoint Sets, Abstractly**

- A disjoint sets ADT represents a collection of disjoint sets.
  - Example: {{4, 6, 2, 0}, {1, 3}, {5}}
- Supports three operations:
   .make\_set(),.find\_set(x),.union(x, y)
- Sometimes called a Union-Find data type.

## Assumption

Elements are consecutive integers.
 Example: {{4, 6, 2, 0}, {1, 3}, {5}}

Not really a limitation.

Keep dictionary mapping, e.g., string ids to integers.

### .make\_set()

Create a new singleton set.

 Element "id" automatically inferred, returned.

```
>> ds = DisjointSet()
>> ds.make_set()
>> ds.make_set()
1
>> ds.make_set()
2
```

### .union(x, y)

- Union sets containing x and y.
- Updates data structure in-place.

>> ds = DisjointSet()
>> ds.make\_set()
>> ds.make\_set()
1
>> ds.make\_set()
2
>> ds.union(0, 2)

#### .find\_set(x)

- Find **representative** of set containing x.
- Representative is arbitrary, but same for all items in same set.
- Used to test if two nodes in same set.
- Guaranteed to not change unless a union is performed.

```
>> # ds is {{0}, {1}, {2}}
\gg ds.union(0, 2)
>> ds.find_set(0)
\odot
>> ds.find set(2)
\odot
\gg ds.union(0. 1)
>> ds.find set(0)
1
>> ds.find set(1)
1
>> ds.find set(2)
1
```

#### **Today's Lecture**

- How do we implement a disjoint set?
- We'll introduce the disjoint set forest data structure.
- Talk about two heuristics that make it very efficient.



Lecture 8 | Part 2

**Disjoint Set Forests** 

## **Implementing Disjoint Sets**

First idea: a list of sets.

 $[\{2, 4, 3\}, \{1, 5\}, \{0\}]$ 

Problem: unioning two sets takes time linear in size of smaller.

# Looking Ahead

 We'll design data structure so that all operations, including union, take (practically) Θ(1) time.

#### The Idea

- Represent collection as a forest of trees, called a Disjoint Set Forest.
- Example: {{2, 4, 3, 6}, {1, 5}, {0}}
- Not unique!

#### **Tree Structure**

- Each node has reference to **parent**.
- ▶ Not a binary tree!

## **Representing Forests**

- We have several choices:
- ▶ 1) Each node is own **object** with parent attribute.
- > 2) Keep a **list** containing parent of each element.

#### Approach #1

#### class DSFNode:

- make\_set becomes DSFNode()
- find\_set and union are functions, not methods.
- They accept DSFNode objects.

## Assumption

- Now, assume that the elements of the disjoint set are consecutive integers 0, 1, ..., n – 1.
- ▶ We can store the tree in a single list of size *n*.
- arr[i] is parent of element i:

[2, 0, None, 2, 4]

#### Approach #2

#### class DisjointSetForest:

```
def __init__(self):
    # self._parent[i] is
    # parent of element i
    self._parent = []
```

```
def make_set(self):
```

• • •

```
def find_set(self, x):
```

• • •

```
def union(self, x, y):
```

. . .

#### **Implementation Notes**

- We'll use the second approach.
- We can use second representation because elements are consecutive integers.
- For cache locality, use numpy array, not list.

#### .make\_set

```
def make_set(self):
    # infer new element's "id"
    x = len(self._parent)
    self._parent.append(None)
    return x
```

```
>> dsf = DisjointSetForest()
>> dsf.make_set()
>> dsf.make_set()
1
>> dsf.make_set()
2
>> dsf._parent
[None, None, None]
```

.find\_set(x)

Idea: use the "root" as the representative.

#### Exercise

#### Implement.find\_set(x) recursively.

```
.find_set
```

```
def find_set(self, x):
    if self._parent[x] is None:
        return x
    else:
        return self.find set(self. parent[x])
```

## .union(x, y)

Idea: make one root the parent of the other.

## .union(x, y)

```
def union(self, x, y):
    x_rep = self.find_set(x)
    y_rep = self.find_set(y)
    if x_rep != y_rep:
        self._parent[y_rep] = x_rep
```

```
>> # dsf is {{0}, {1}, {2}}
>> dsf._parent
[None, None, None]
>> dsf.union(0, 1)
>> dsf._parent
[None, 0, None]
>> dsf.union(1, 2)
>> dsf._parent
[None, 0, 0]
```

# Analysis

- .make\_set: Θ(1) time<sup>1</sup>
- .union: depends on .find\_set
- .find\_set: O(h), where h is height of tree

<sup>&</sup>lt;sup>1</sup>Amortized, since we're using a dynamic array. But truly  $\Theta(1)$  with an over-allocated static array or in the object representation.

### **Tree Height**

Trees can be very deep, with h = O(n).
find\_set and .union can take O(n) time!

#### Example: # dsf is {{0}, {1}, {2}, {3}, {4}} >> dsf.union(1, 0) >> dsf.union(2, 1) >> dsf.union(3, 2) >> dsf.union(4, 3)

## **Tree Height**

But trees can also be shallow, with h = O(1).

```
Example:
# dsf is {{0}, {1}, {2}, {3}, {4}}
>> dsf.union(0, 1)
>> dsf.union(1, 2)
>> dsf.union(2, 3)
>> dsf.union(3, 4)
```



#### Lecture 8 | Part 3

#### Path Compression and Union-by-Rank

#### **The Bad News**

- We saw that the tree can become very deep.
- In worst case, .find\_set and thus .union take O(n) time.

## Heuristics

- Now: two heuristics helping trees stay shallow.
- Union-by-Rank and Path Compression
- Together, these result in a massive speed up.

## **Path Compression**

Idea: if we find a long path during .find\_set, "compress" it to (possibly) reduce height.

#### .find\_set

```
def find_set(self, x):
    if self._parent[x] is None:
        return x
    else:
        root = self.find_set(self._parent[x])
        self._parent[x] = root
        return root
```

## Union-by-Rank

Should we .union(x, y) or .union(y, x)?

## Union-by-Rank

- Placing deeper tree under shallower tree increases height by one.
- But placing shallower tree under deeper tree doesn't increase height.
- **Idea**: always place shallower tree under deeper.

## Rank

- We need to keep track of height (rank) of each tree.
- Store rank attribute.
- rank[i] is height<sup>2</sup> of tree rooted at node i.

<sup>&</sup>lt;sup>2</sup>Exactly the height if path compression isn't used, but upper bound if it is.

#### Rank

```
class DisjointSetForest:
```

```
def __init__(self):
    self._parent = []
    self._rank = []
```

```
def make_set(self):
    # infer new element's "id"
    x = len(self._parent)
    self._parent.append(None)
    self._rank.append(0)
    return x
```

#### .union

```
def union(self, x, y):
   x rep = self.find set(x)
   v rep = self.find set(v)
    if x rep == v rep:
        return
    if self. rank[x rep] > self. rank[v rep]:
        self. parent[v rep] = x rep
    else:
        self. parent[x rep] = y rep
        if self. rank[x rep] == self. rank[v rep]:
            self. rank[y rep] += 1
```

#### Note

- With path compression, rank is no longer exactly the height – it is an upper bound.
- But this is good enough.



Lecture 8 | Part 4

Analysis

## Analysis of DSF

- A DSF with path compression and union-by-rank ensures trees are shallow.
- How does this affect runtime?

#### Answer

- Assuming union-by-rank and path compression...
- In a sequence of *m* operations, *n* of which are .make\_sets...
- Amortized cost of a single operation is  $O(\alpha(n))$ .

α is the inverse Ackermann function, and it is essentially constant.

#### Inverse Ackermann

 $\begin{array}{ll} \alpha(n) & n \\ 0 & n \in [0, 1, 2] \\ 1 & n = 3 \\ 2 & n \in [4, \dots, 7] \\ 3 & n \in [8, \dots, 2047] \\ 4 & n \in [2048, \dots, 2^{2048}] \text{ and beyond} \end{array}$ 

## Proof

- The formal analysis is quite involved.
- But we'll provide some intuition.

## **Union-by-rank Alone**

Union-by-rank alone ensures height is O(log n).

# dsf is {{0}, {1}, {2}, {3}}
>> dsf.union(0, 1)
>> dsf.union(2, 3)
>> dsf.union(0, 2)

## **Union-by-rank Alone**

Union-by-rank alone ensures .find\_set is O(log n).

## Path Compression + U-by-R

- With path compression, individual .find\_set calls can take O(log n).
- But they massively improve subsequent calls.
   For other nodes, too!



Lecture 8 | Part 5

**Epilogue: pytest** 

## **Testing Your Code**

- Testing code is essential (for homework and real life).
- Consider it to be part of the problem.
- How do we test Python code?

## Approach #1: Run it by hand

► Write your code.

Open up a Python interpreter.

Type in a few examples, see if code works.

It doesn't work. Repeat.

#### **Downsides**

You often run the same test over and over again.

- > You have to type it in every time.
- This is annoying.

#### Main Idea

If something is annoying, you'll avoid doing it. Spend the time to make things less annoying.

#### **Approach #2: Doctests**

def add(x, y):
 """Add two numbers.

>> add(1, 2)
3
>> add(2, 2)
4
>> add(1, -1)
0
"""

return x + y

#### Doctests

# Useful, but brittle.Relies on string comparison.

#### Approach #2: Unit Testing Frameworks

Create a file that only includes tests.

- Write test for each way that code will be used.
   Example: for a stack, write test for push, pop, peek.
- Try to anticipate "corner cases".
- Write the tests **before** you write the code.

# **Unit Testing in Python**

- unittest: built-in module for unit testing
- pytest: nicer to use, more "modern"

```
import stack
import pytest
def test_push_then_peek():
    s = stack_Stack(10)
    s.push(1)
    s.push(5)
    s.push(3)
    assert s.peek() == 3
def test_push_then_pop():
    s = stack.Stack(10)
    s.push(1)
    s.push(5)
    s.push(3)
    assert s.pop() == 3
```

# Debugging

- Testing and debugging go hand-in-hand.
- Should know how to use the Python debugger.

# **Unit Testing Guidelines**

- Should test "public" interface, not "private" implementation details.
- Should "exercise" all of the code (coverage).
- Write the tests before the code.