

DSC 190

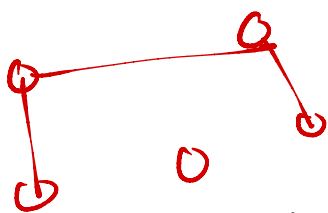
DATA STRUCTURES & ALGORITHMS

Lecture 8 | Part 1

Today's Lecture

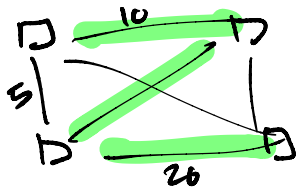
Disjoint Sets

- ▶ 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 $\Theta(V + E)$ 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()`

$\{\{0\}, \{1\}, \{2\}\}$

- Create a new singleton set.
- Element “id” automatically inferred, returned.

```
»» ds = DisjointSet()
»» ds.make_set()
0
»» ds.make_set()
1
»» ds.make_set()
2
```

`.union(x, y)`

`.union(2, 1)`

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

```
»> ds = DisjointSet()
»> ds.make_set()
0
»> ds.make_set()
1
»> ds.make_set()
2
»> ds.union(0, 2)
```

$\{\{0\}, \{1\}, \{2\}\} \rightarrow \{\{0, 2\}, \{1\}\}$

.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}}
>>> ds.union(0, 2)
>>> ds.find_set(0)
0
>>> ds.find_set(2)
0
>>> ds.union(0, 1)
>>> ds.find_set(0)
1
>>> ds.find_set(1)
1
>>> ds.find_set(2)
1
```

$\{\{0, 2\}, \{1\}\}$

$\{\{0, 1, 2\}\}$

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.

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DATA STRUCTURES & ALGORITHMS

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.

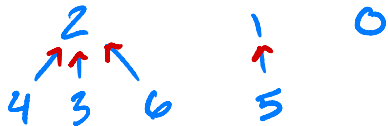
Looking Ahead

$\Theta(\alpha(n))$

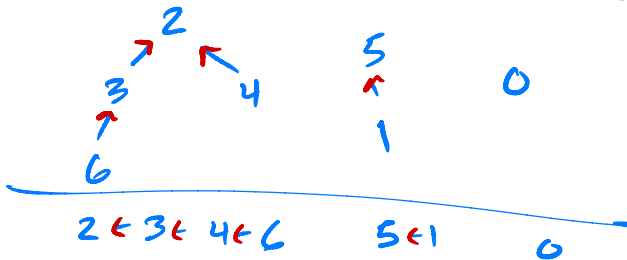
- We'll design data structure so that all operations, including union, take (practically) $\Theta(1)$ time.

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:
```

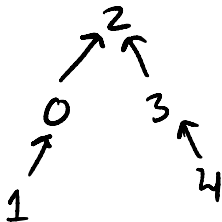
```
    def __init__(self, key, parent=None):  
        self.parent = parent  
        self.key = key
```

- ▶ 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** $0, 1, \dots, n - 1$.
- ▶ We can store the tree in a single list of size n .
- ▶ $\text{arr}[i]$ 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):
        ...
```

def __len__

__foo

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`.

0 1 2

.make_set

$\{\{0\}, \{1\}, \{2\}\}$

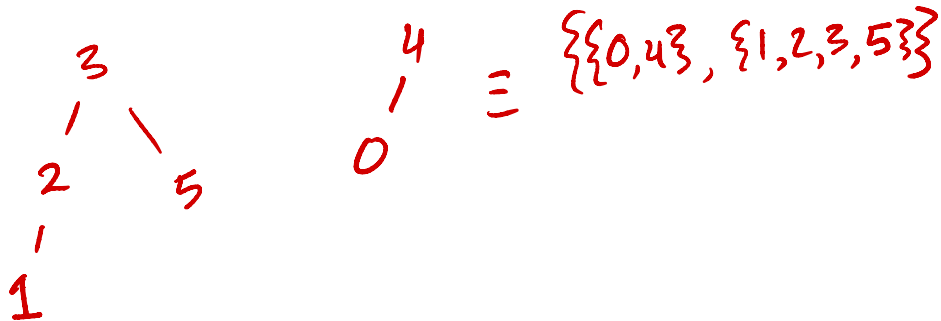
```
def make_set(self):  
    # infer new element's "id"  
    x = len(self._parent)  
    self._parent.append(None)  
    return x
```

$[None, None, None]$

```
>>> dsf = DisjointSetForest()  
>>> dsf.make_set()  
0  
>>> dsf.make_set()  
1  
>>> dsf.make_set()  
2  
>>> dsf._parent  
[None, None, None]
```

.find_set(x)

- Idea: use the “root” as the representative.



def find_set(x)

Exercise

Implement `.find_set(x)` recursively.

.find_set

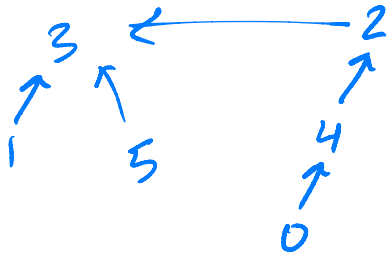
```
def find_set(self, x):  
    if self._parent[x] is None:  
        return x  
    else:  
        return self.find_set(self._parent[x])
```

3
/
2
/
5

.union(x, y)

- Idea: make one root the parent of the other.

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



0 ↗
1 ↖
2 ↖

(x_rep) ← (y_rep)

.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`: $\Theta(1)$ time¹
- ▶ `.union`: depends on `.find_set`
- ▶ `.find_set`: $O(h)$, where h is height of tree

¹Amortized, 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 $\Theta(n)$ 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)
```

3
↑
2
↑
1
↑
0

4

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)
```



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Lecture 8 | Part 3

Path Compression and Union-by-Rank

The Bad News

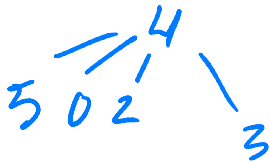
- ▶ We saw that the tree can become very deep.
- ▶ In worst case, `.find_set` and thus `.union` take $\Theta(n)$ time.

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.

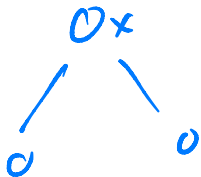


.find_set

```
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.
- ▶ $\text{rank}[i]$ is height^2 of tree rooted at node i .

²Exactly 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)  
    y_rep = self.find_set(y)  
  
    if x_rep == y_rep:  
        return  
  
    if self._rank[x_rep] > self._rank[y_rep]:  
        self._parent[y_rep] = x_rep  
    else:  
        self._parent[x_rep] = y_rep  
        if self._rank[x_rep] == self._rank[y_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.

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DATA STRUCTURES & ALGORITHMS

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

| $\alpha(n)$ | n |
|-------------|-----|
|-------------|-----|

| | |
|---|-------------------|
| 0 | $n \in [0, 1, 2]$ |
|---|-------------------|

| | |
|---|---------|
| 1 | $n = 3$ |
|---|---------|

| | |
|---|-----------------------|
| 2 | $n \in [4, \dots, 7]$ |
|---|-----------------------|

| | |
|---|--------------------------|
| 3 | $n \in [8, \dots, 2047]$ |
|---|--------------------------|

| | |
|---|--|
| 4 | $n \in [2048, \dots, 2^{2048}]$ and beyond |
|---|--|

$\tilde{O}(n)$

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!

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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
```

python -m
doctest
file.py

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
```

pytest --pdb

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.