

Data structures

Computer Science Enrichment Club - Algorithms Division October 27, 2017

- Review the Union-Find data structure, and look at applications
- Study range queries
- Quick look at Square Root Decomposition
- Learn about Segment Trees

- We have *n* items
- Maintains a collection of disjoint sets
- Each of the *n* items is in exactly one set
- *items* = $\{1, 2, 3, 4, 5, 6\}$
- collections = $\{1, 4\}, \{3, 5, 6\}, \{2\}$
- collections = $\{1\}, \{2\}, \{3\}, \{4\}, \{5\}, \{6\}$
- Supports two operations efficiently: find(x) and union(x,y).

Union-Find

- *items* = $\{1, 2, 3, 4, 5, 6\}$
- collections = $\{1,4\}, \{3,5,6\}, \{2\}$
- find(x) returns a representative item from the set that x is in
 - find(1) = 1
 - find(4) = 1
 - find(3) = 5
 - find(5) = 5
 - find(6) = 5
 - find(2) = 2
- a and b are in the same set if and only if find(a) == find(b)

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- union(x, y) merges the set containing x and the set containing y together.
 - union(4, 2)
 - collections = $\{1, 2, 4\}, \{3, 5, 6\}$
 - union(3, 6)
 - collections = $\{1, 2, 4\}, \{3, 5, 6\}$
 - union(2, 6)
 - collections = $\{1, 2, 3, 4, 5, 6\}$

Union-Find implementation

- Quick Union with path compression
- Extremely simple implementation
- Extremely efficient

```
struct union_find {
    vector<int> parent;
    union find(int n) {
        parent = vector<int>(n);
        for (int i = 0; i < n; i++) {</pre>
            parent[i] = i;
        }
    }
    // find and union
};
```

// find and union

}

```
int find(int x) {
    if (parent[x] == x) {
        return x;
    } else {
        parent[x] = find(parent[x]);
        return parent[x];
    }
}
void unite(int x, int y) {
    parent[find(x)] = find(y);
```

```
• If you're in a hurry...
```

```
#define MAXN 1000
int p[MAXN];
```

```
int find(int x) {
    return p[x] == x ? x : p[x] = find(p[x]); }
void unite(int x, int y) { p[find(x)] = find(y); }
```

```
for (int i = 0; i < MAXN; i++) p[i] = i;</pre>
```

- Union-Find maintains a collection of disjoint sets
- When are we dealing with such collections?
- Most common example is in graphs







• *items* =
$$\{1, 2, 3, 4, 5, 6, 7\}$$





- *items* = $\{1, 2, 3, 4, 5, 6, 7\}$
- collections = $\{1, 4, 7\}, \{2\}, \{3, 5, 6\}$





- *items* = $\{1, 2, 3, 4, 5, 6, 7\}$
- collections = $\{1, 4, 7\}, \{2\}, \{3, 5, 6\}$
- union(2, 5)



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- union(6, 2)



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• https://open.kattis.com/problems/wheresmyinternet

- We have an array A of size n
- Given *i*, *j*, we want to answer:
 - $\max(A[i], A[i+1], ..., A[j-1], A[j])$
 - $\min(A[i], A[i+1], ..., A[j-1], A[j])$
 - sum(A[i], A[i+1], ..., A[j-1], A[j])
- We want to answer these queries efficiently, i.e. without looking through all elements
- Sometimes we also want to update elements

• sum(0,6)

• sum(0, 6) = 33

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- sum(0, 6) = 33
- sum(2,5) = 29
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- How do we support these queries efficiently?

Range sum on a static array

- Simplification: only support queries of the form sum(0, j)
- Notice that sum(i,j) = sum(0,j) sum(0,i-1)



- So we're only interested in prefix sums
- But there are only *n* of them...
- Just compute them all once in the beginning

1	0	7	8	5	9	3

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1						

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1	1					

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- O(1) time each query
- Can we support updating efficiently?
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- Can we support updating efficiently? No, at least not without modification

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---	---	---	---	---	---	---

- sum(0, 6) = 33
- update(3, -2)

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1	0	7	-2	5	9	3
---	---	---	----	---	---	---

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- E.g. *k* = 2:

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- Easy to do in O(1), but doesn't really matter (we'll see why)





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 - Only have to go inside at most two buckets (each end)
 - Have to consider at most n/k buckets
 - In total roughly n/k + 2k
 - Time complexity: O(n/k + k)

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- Time complexity is minimized for $k = \sqrt{n}$:
 - Updating in $O(\sqrt{n})$
 - Querying in $O(n/\sqrt{n} + \sqrt{n}) = O(\sqrt{n})$
- Also known as square root decomposition, and is a very powerful technique

• https://open.kattis.com/problems/supercomputer
- Now we know how to do these queries in $O(\sqrt{n})$
- May be too slow if *n* is large and many queries
- Can we do better?













• Each vertex contains the sum of some segment of the array

Segment Tree - Code

```
struct segment tree {
    segment tree *left. *right:
    int from, to, value;
    segment_tree(int from, int to)
        : from(from), to(to), left(NULL), right(NULL), value(0) { }
}:
segment_tree* build(const vector<int> &arr, int 1, int r) {
   if (1 > r) return NULL;
    segment_tree *res = new segment_tree(1, r);
    if (1 == r) {
        res->value = arr[1];
   } else {
        int m = (1 + r) / 2;
        res->left = build(arr, 1, m);
        res->right = build(arr, m + 1, r);
        if (res->left != NULL) res->value += res->left->value;
        if (res->right != NULL) res->value += res->right->value;
    3
   return res:
}
```























• sum(0,5) = 16 + 14 = 30



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- We only need to consider a few vertices to get the entire range



- sum(0,5) = 16 + 14 = 30
- We only need to consider a few vertices to get the entire range
- But how do we find them?













```
int query(segment_tree *tree, int 1, int r) {
    if (tree == NULL) return 0;
    if (1 <= tree->from && tree->to <= r) return tree->value;
    if (tree->to < 1) return 0;
    if (r < tree->from) return 0;
    return query(tree->left, 1, r) + query(tree->right, 1, r);
}
```























```
int update(segment_tree *tree, int i, int val) {
    if (tree == NULL) return 0;
    if (tree->to < i) return tree->value;
    if (i < tree->from) return tree->value;
    if (tree->from == tree->to && tree->from == i) {
        tree->value = val;
    } else {
        tree->value = update(tree->left, i, val) + update(tree->right, i, val);
    }
    return tree->value;
}
```

- Now we can
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 - query a range
 - update a single value
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- Now we can
 - build a Segment Tree in O(n)
 - query a range in $O(\log n)$
 - update a single value in $O(\log n)$
- But how efficient are these operations?
- Trivial to use Segment Trees for min, max, gcd, and other similar operators, basically the same code
- Also possible to update a range of values in $O(\log n)$ (Google for Segment Trees with Lazy Propagation if you want to learn more)

• https://open.kattis.com/problems/supercomputer