Disjoint Sets: Naive Implementations

Alexander S. Kulikov

Steklov Institute of Mathematics at St. Petersburg Russian Academy of Sciences

Data Structures Data Structures and Algorithms





2 Naive Implementations











A disjoint-set data structure supports the following operations:

• MakeSet(x) creates a singleton set $\{x\}$

A disjoint-set data structure supports the following operations:

MakeSet(x) creates a singleton set {x}
Find(x) returns ID of the set containing x:

A disjoint-set data structure supports the following operations:

- MakeSet(x) creates a singleton set {x}
 Find(x) returns ID of the set containing x:
 - if x and y lie in the same set, then
 Find(x) = Find(y)

A disjoint-set data structure supports the following operations:

- MakeSet(x) creates a singleton set $\{x\}$
- Find(x) returns ID of the set containing x:
 - if x and y lie in the same set, then Find(x) = Find(y)
 otherwise, Find(x) ≠ Find(y)

A disjoint-set data structure supports the following operations:

- MakeSet(x) creates a singleton set $\{x\}$
- Find(x) returns ID of the set containing x:
 - if x and y lie in the same set, then
 Find(x) = Find(y)
 - otherwise, $Find(x) \neq Find(y)$
- Union(x, y) merges two sets containing x and y

Preprocess(*maze*)

for each cell c in maze: MakeSet(c) for each cell c in maze: for each neighbor n of c: Union(c, n)

Preprocess(*maze*)

for each cell c in maze: MakeSet(c) for each cell c in maze: for each neighbor n of c: Union(c, n)

IsReachable(A, B)
return Find(A) = Find(B)



MakeSet(1)

































 $\texttt{Find}(1) = \texttt{Find}(2) \rightarrow \texttt{False}$







































2 Naive Implementations

For simplicity, we assume that our n objects are just integers $1, 2, \ldots, n$.

Using the Smallest Element as ID

Use the smallest element of a set as its ID

Using the Smallest Element as ID

- Use the smallest element of a set as its ID
- Use array smallest[1...n]:
 smallest[i] stores the smallest element in the set i belongs to

Example

$$\{9,3,2,4,7\} \quad \{5\} \quad \{6,1,8\}$$
smallest 1 2 2 2 5 1 2 1 2

$$\texttt{smallest}[i] \leftarrow i$$

Find(i)

return smallest[i]

$$\texttt{smallest}[i] \gets i$$

Find(i)

return smallest[i]

Running time: O(1)

 $i_i d \leftarrow Find(i)$ $j_id \leftarrow Find(j)$ if $i_i = j_i$: return $m \leftarrow \min(i_i, j_i)$ for k from 1 to n: if smallest[k] in $\{i_i, j_i, j_i\}$: $\texttt{smallest}[k] \leftarrow m$

 $i_i d \leftarrow Find(i)$ $j_i d \leftarrow Find(j)$ if $i_i = j_i$: return $m \leftarrow \min(i_i, j_i, j_i, d)$ for k from 1 to n: if smallest[k] in $\{i_id, j_id\}$: $\texttt{smallest}[k] \leftarrow m$

Running time: O(n)

Current bottleneck: Union

- Current bottleneck: Union
- What basic data structure allows for efficient merging?

- Current bottleneck: Union
- What basic data structure allows for efficient merging?
- Linked list!

- Current bottleneck: Union
- What basic data structure allows for efficient merging?
- Linked list!
- Idea: represent a set as a linked list, use the list tail as ID of the set









Pros: Running time of Union is O(1)

Running time of Union is O(1) Well-defined ID

Running time of Union is O(1) Well-defined ID

Cons:

- Running time of Union is O(1)
 Well-defined ID
- Cons:
 - Running time of Find is O(n) as we need to traverse the list to find its tail

Running time of Union is O(1)
Well-defined ID

Cons:

- Running time of Find is O(n) as we need to traverse the list to find its tail
- Union(x, y) works in time O(1) only if we can get the tail of the list of x and the head of the list of y in constant time!

$9 \rightarrow 3 \rightarrow 2 \rightarrow 4 \rightarrow 7 \rightarrow$

 $6 \rightarrow 1 \rightarrow 8 \rightarrow$





Find(9) goes through all elements



can we merge in a different way?

$9 \rightarrow 3 \rightarrow 2 \rightarrow 4 \rightarrow 7 \rightarrow$

 $6 \rightarrow 1 \rightarrow 8 \rightarrow$

$9 \rightarrow 3 \rightarrow 2 \rightarrow 4 \rightarrow 7 \rightarrow$ $6 \rightarrow 1 \rightarrow 8 \rightarrow$

$9 \rightarrow 3 \rightarrow 2 \rightarrow 4 \rightarrow 7 \rightarrow$ $6 \rightarrow 1 \rightarrow 8 \rightarrow$

instead of a list we get a tree



we'll see that representing sets as trees gives a very efficient implementation: nearly constant amortized time for all operations