# **Sets**

It is assumed that the elements of the sets are the numbers 1, 2, 3, ..., n. These numbers might, in practice, be indices into a symbol table in which the names of the elements are stored.

We assume that the sets being represented are *pairwise disjoint* (that is, if Si and Sj,  $i \neq j$ , are two sets, then there is no element that is in both Si and Sj).

For example, when n = 10, the elements can be partitioned into three disjoint sets:

- $S1 = \{1, 7, 8, 9\}$
- $S2 = \{2, 5, 10\}$
- $S3 = \{3, 4, 6\}$

Figure shows one possible representation for these:

### S1:

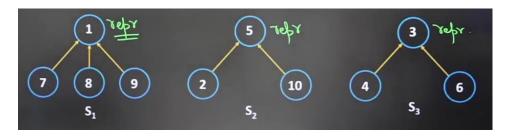
- 1 (representative)
  - 。 7
  - 。 8
  - 。 9

## S2:

- 5 (repr)
  - 。 2
  - 。 10

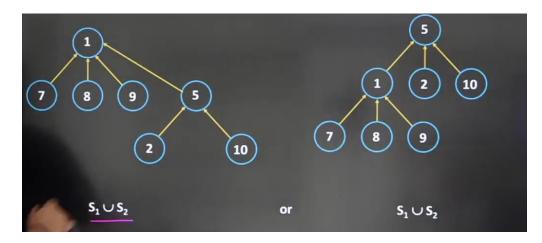
- 3 (repr)
  - 。 4
  - 。 6

Possible tree representation of sets



### The operations we wish to perform on these sets are:

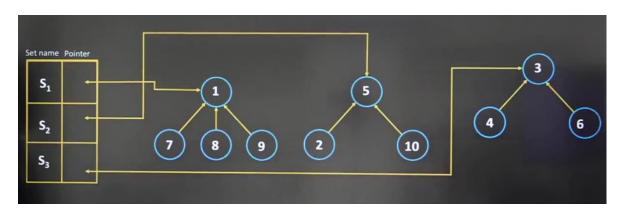
1. **Disjoint set union.** If Si and Sj are two disjoint sets, then their union Si U Sj = all elements x such that x is in Si or Sj. Thus, S1 U S2 = {1, 7, 8, 9, 2, 5, 10}. Since we have assumed that all sets are disjoint, we can assume that following the union of Si and Sj, the sets Si and Sj do not exist independently; that is, they are replaced by Si U Sj in the collection of sets.



2. **Find(i).** Given the element i, find the set containing i. Thus, 4 is in set S3, and 9 is in set S1.

# **Representation of Sets:**

## **Data Representation:**



# **Array-based Representation:**

For n = 10, consider the sets:

- $S1 = \{1, 7, 8, 9\}$
- $S2 = \{2, 5, 10\}$
- $S3 = \{3, 4, 6\}$

This can be represented using an array P such that:

Here, a negative value indicates that the element is the representative (root) of the set, and its magnitude shows the size (optional, not visible

here). A positive value points to the index of its parent (i.e., the representative of the set it belongs to).

This representation allows efficient implementation of **find** and **union** operations.

# **Algorithms:**

```
Algorithm: Union(i, j){
p[i] := j
}
Algorithm: Find(i){
while (p[i] \ge 0) do i := p[i];
return i;
}
```

# **Optimizations**

## **Path Compression (Optimized Find):**

```
function Find(i):
    if p[i] < 0 then
        return i;
    else
        p[i] := Find(p[i]);
    return p[i];</pre>
```

This helps flatten the structure for faster future queries.

## **Union by Size (or Rank):**

```
function Union(i, j):
    iRoot := Find(i);
    jRoot := Find(j);
    if iRoot == jRoot then return;

if size[iRoot] > size[jRoot] then
    p[jRoot] := iRoot;
    size[iRoot] += size[jRoot];

else
    p[iRoot] := jRoot;
    size[jRoot] += size[iRoot];
```

(*Note: The size can be encoded by negative values in p[]*)

# **Time Complexity:**

With both Path Compression and Union by Rank/Size:

• Each operation runs in  $O(\alpha(n))$ , where  $\alpha$  is the **inverse** Ackermann function, which grows extremely slowly.

This makes the disjoint set data structure nearly constant time for practical purposes.