

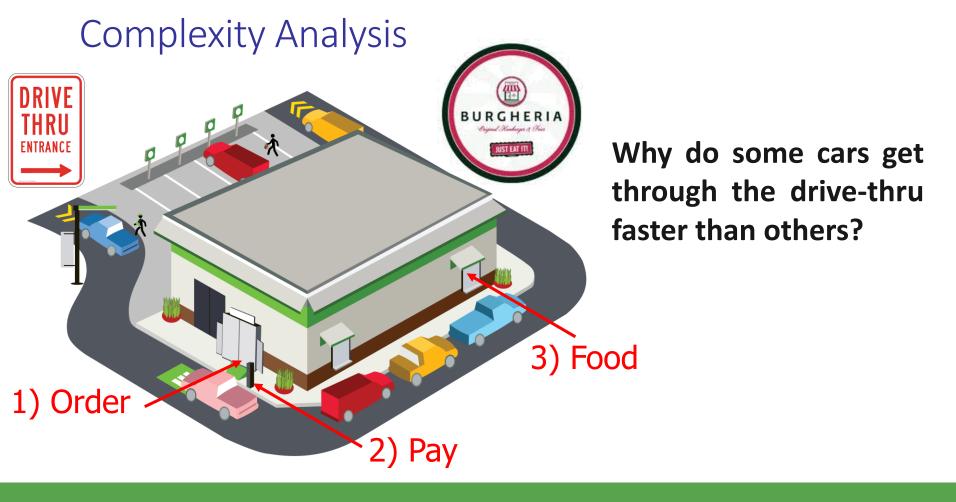
Complexity Analysis Paolo Camurati

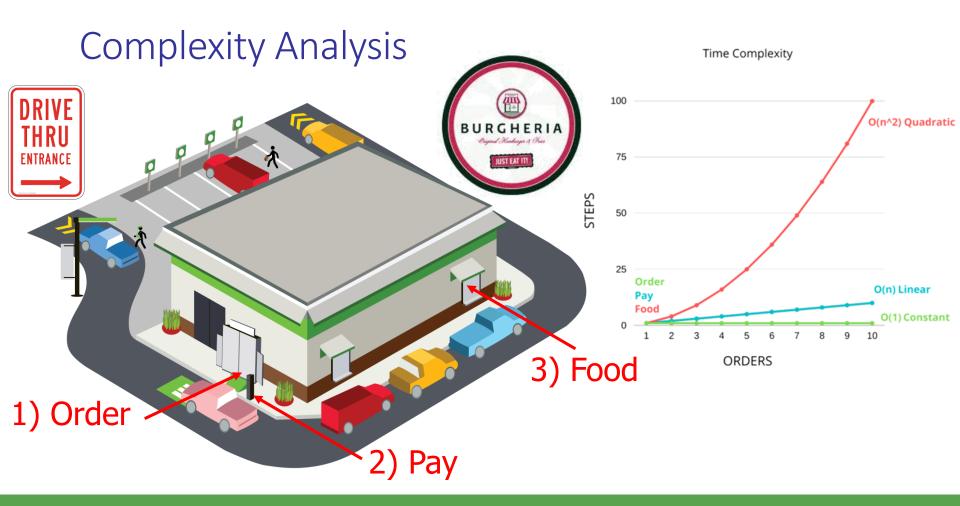
Edited by Josie E. Rodriguez











Definition:

- Forecast of resources (memory, time) needed by the algorithm for execution.
 - Empirical
 - Analytical

Features:

- Machine-independent
- Assumption: Sequential single-processor model (traditional architecture)
- Independent of the input data of a particular instance of the problem.

Example:

- Problem P: sort integer data
- Instance I: data are 45 10 6 7 99
- Size of instance |I|: number of bits needed to encode I,
 in this case 5 x the size of the integer or simply 5

- It depends on the size n of the problem.
- Examples:
 - n number of bits of the operands for integer multiplication
 - n size of the file to sort
 - n number of characters in a string of text
 - n number of data to sort for a sorting algorithm
- Output:
 - S(n): memory occupation (memory footprint)
 - T(n): execution time (Performance)

Algorithm Classification

- 1: constant
- log n: logarithmic
- n: linear
- n log n: *linearithmic*
- n²: **quadratic**
- n³: **cubic**
- 2ⁿ: exponential

Worst-case Asymptotic Analysis

Goal:

 to guess an upper-bound for T(n) (execution time) for an algorithm on n data in the worst possible case

Asymptotic: $n \rightarrow \infty$:

for small **n**, complexity is irrelevant

Worst-case Asymptotic Analysis

Why worst-case analysis?

- Conservative guess
- Worst case is very frequent
- Average case:
 - either it coincides with the worst case
 - or it is not definable, unless we resort to complex assumptions on data.

Importance of complexity analysis

Advantages of a lower complexity:

- it compensates hardware (in)efficiencyExample:
- Algorithm #1:
 - T(n) (execution time) = $2n^2$
 - machine #1: 10⁸ instructions/second
- Algorithm #2:
 - **T(n) (execution time)** = $50n \lg_2 n$
 - machine 2: 10⁶ instructions/second





Importance of complexity analysis

```
If n = 1M = 10^6:
```

- Algorithm #1: $2 \cdot (10^6)^2 / 10^8 = 2 \cdot 10^4 = 20000 \text{ s} = 333,33 \text{ min}$
- Algorithm #2: $50.10^6 \lg_2 10^6 / 10^6 = 50.6 \lg_2 10^6 1000 s = 16,67 min$

An inefficient algorithm rapidly «wastes» the increase in hardware performance!

Examples

Discrete Fourier Transform:

- decomposition of a N-sample waveform into periodic components
- applications: DVD, JPEG, astrophysics,
- trivial algorithm: quadratic (N²)
- FFT (Fast Fourier Transform): N log N

Simulation of N bodies:

- simulates gravity interaction among N bodies
- trivial algorithm: quadratic (N²)
- Barnes-Hut algorithm: N log N

Search Algorithms on Arrays

Let v[N] be an array of N distinct elements, let k be a key:

Decision problem:

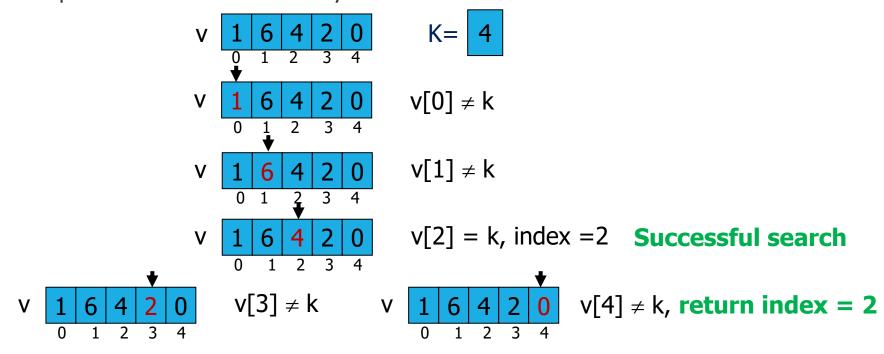
Does key k appear in array v[N]? Yes/No

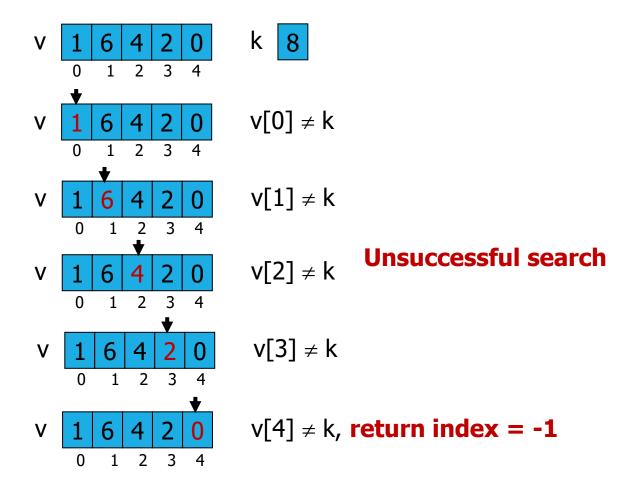
Search problem:

if **k** is in the array **v[N]**, where (**at what index**)?

Algorithm #1: Linear Search

Scan the array v[N] from first to possibly last element, compare at each step current element and key k.





Alternatives:

- Solution #1: scan the array from first to last element: always N operations
- Solution #2: use a flag: early scan stop possible, at most N operations, in the worst case N operations.

The worst-case asymptotic complexity is the same (N), the second alternative improves the average case.

Solution #1

```
int LinearSearch1(int v[], int N, int k)
  int i = 0, index = -1;
 for (i = 0; i < N; i++)
   if (k == v[i])
      index = i;
  return index;
```

Solution #2

```
int LinearSearch2(int v[], int N, int k)
 int i = 0;
 int found = 0;
 while (i < N \&\& found == 0)
    if (k == v[i])
      found = 1;
    else
      1++;
  if (found == 0)
   return -1;
  else
    return i;
```

Algorithm #2: Binary Search

Let **v[N]** be a **sorted** array of **N** distinct elements and let **k** be a key **k**:

- Decision problem: does key k appear in array v[N]? Yes/No
- Search problem: if k is in the array, where (at what index)?

We work on a **subarray** identified by the contiguous elements of **v** whose indices range from a **leftmost one** (I) and a **rightmost one** (r).

Initially array and **subarray** coincide (I = 0 and r = N-1).

The middle element of the subarray is at index m = (l+r)/2.

Example: **k=9**



Algorithm #2: Binary Search

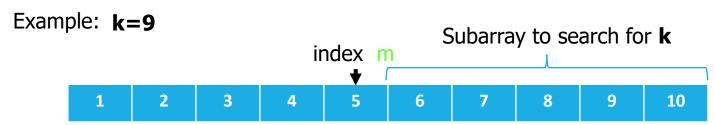
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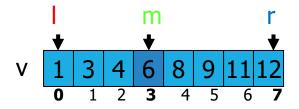


- Loop: at each step compare k to the middle element v[m] of the subarray
- Loop condition: $l \le r \&\&$ found==0: the key has not yet been found and the subarray is meaningful (l doesn't exceed r)
- Body of the loop:
 - if v[m] == k:
 termination with success, found = 1
 - if v[m] < k: search continues in the right subarray: l=m+1, r unchanged
 - if v[m] > k: search continues in the left subarray: | unchanged, r = m-1
- Upon exiting the loop, test found, return -1 for failure or m for success.

1 3 4 6 8 9 11 12 k 4

| = leftmost index, initially | = 0
| r = rightmost index, initially r = N-1
| m = index of middle element
| v[m] = middle element

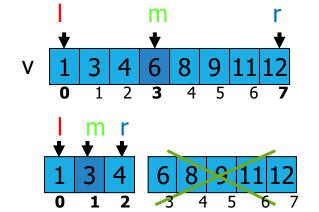




Iteration 1:

$$l=0$$
, r=7, m= ($l+r$)/2=3, v[3]>k, r=m-1



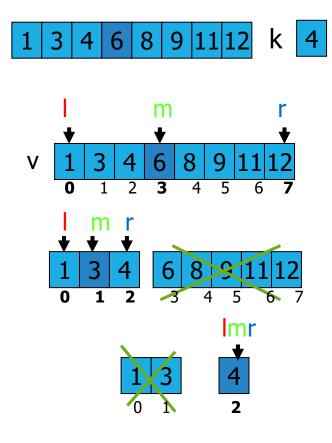


Iteration 1:

$$l=0$$
, $r=7$, $m=(l+r)/2=3$, $v[3]>k$, $r=m-1$

Iteration 2:

$$l=0, r=2, m= (l+r)/2=1, v[1]< k, l=m+1$$



Iteration 1:

$$l=0$$
, $r=7$, $m=(l+r)/2=3$, $v[3]>k$, $r=m-1$

Iteration 2:

$$l=0, r=2, m= (l+r)/2=1, v[1]< k, l=m+1$$

Iteration 3:

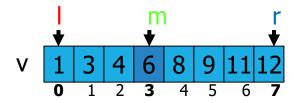
$$l=2, r=2, m= (l+r)/2=2, v[2]=k$$

Successful search, return index =2

1 3 4 6 8 9 11 12 k 10

l = leftmost index, initially l = 0
r = rightmost index, initially r = N-1
m = index of middle element
v[m] = middle element

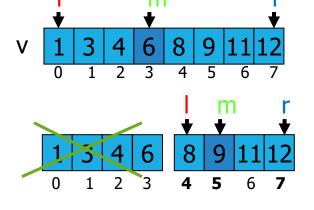




Iteration 1:

$$l=0, r=7, m= (l+r)/2=3, v[3]< k, l=m+1$$



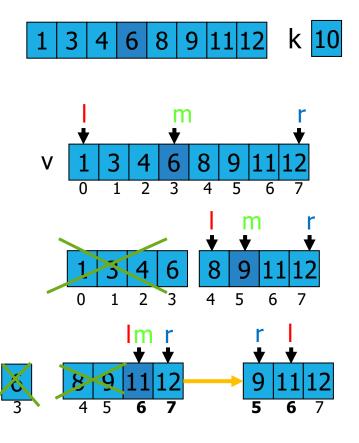


Iteration 1:

$$l=0, r=7, m= (l+r)/2=3, v[3]< k, l=m+1$$

Iteration 2:

$$l=4$$
, $r=7$, $m=(l+r)/2=5$, $v[5]< k$, $l=m+1$



Iteration 1:

$$l=0, r=7, m= (l+r)/2=3, v[3]< k, l=m+1$$

Iteration 2:

$$l=4$$
, $r=7$, $m=(l+r)/2=5$, $v[5]< k$, $l=m+1$

Iteration 3:

```
int BinSearch(int v[], int N, int k) {
  int m, found= 0, 1=0, r=N-1;
while(1 <= r && found == 0){</pre>
    m = (1+r)/2;
    if(v[m] == k)
      found = 1;
    if(v[m] < k)
     1 = m+1;
    else
      r = m-1;
  if (found == 0)
    return -1;
  else
    return m;
```

Analysis of Linear Search

- We consider n numbers for a search miss and in average n/2 for a search hit
- T(n) grows linearly with n.

Analysis of Binary Search

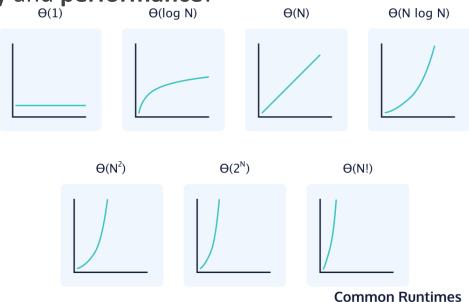
- At the beginning the array to be examined contains n numbers
- At the 2nd iteration the array to be examined contains about n/2 numbers
- •
- At the i-th iteration the array to be examined contains about n/2i numbers
- Termination occurs when the array to be examined contains 1 number, thus n/2ⁱ = 1, i = log₂(n)
- T(n) grows logarithmically with n.

Asymptotic Notations

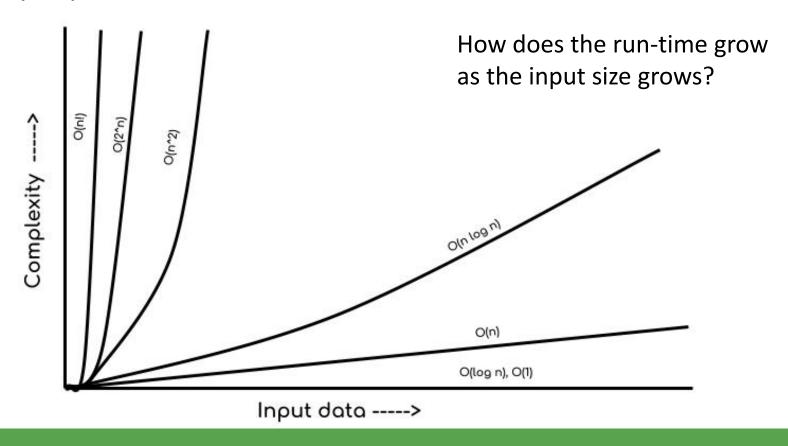
```
O(1) = O(yeah)
O(\log n) = O(\text{nice})
   O(n) = O(ok)
  O(n^2) = O(my)
  O(2^n) = O(no)
  O(n!) = O(mg)
```

Asymptotic Notations

- Mathematical notations to describe the running time of an algorithm (when the input tends towards a particular value or a limiting value).
- We are talking about efficiency and performance!
- Most used:
 - Big-Oh notation
 - Omega notation
 - Theta notation



Asymptotic Notations



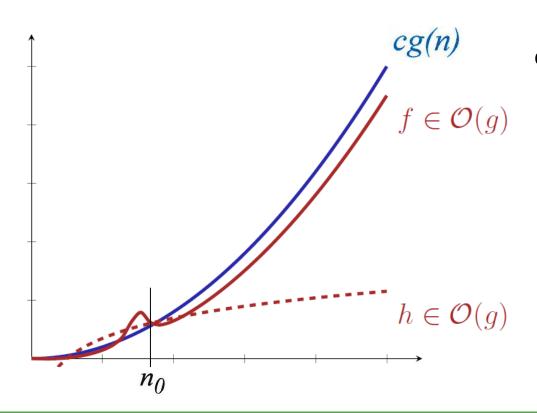
Definition:

```
T(n) = O(g(n)) \Leftrightarrow
 \exists c>0, \exists n_0>0 \text{ such that } \forall n \geq n_0
 0 \leq T(n) \leq cg(n)
```

```
g(n) = upper bound for T(n). The number of steps is at most g(n) (constant c doesn't count in asymptotic analysis).
```

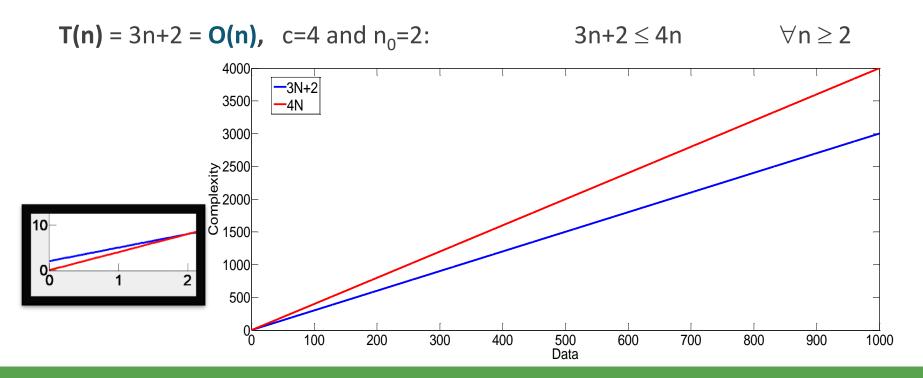
```
c (constant)n (data)
```

Worst case scenario!



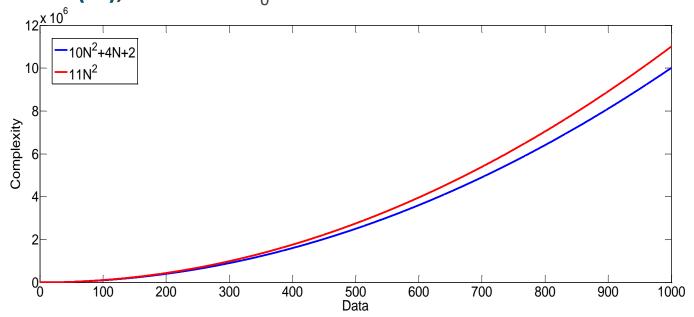
gives the worst-case complexity of an algorithm

Examples:



Examples:

T(n) =
$$10n^2 + 4n + 2 =$$
O(n^2), c=11 and $n_0 = 5$ $10n^2 + 4n + 2 \le 11n^2 \quad \forall n \ge 5$



Examples:

$$T(n) = 3n+2 = O(n^2)$$
, c=3 and $n_0=2$

$$3n+2 \leq 3n^2$$

$$\forall\, n\geq 2$$

Theorem:

if
$$T(n) = a_m n^m + + a_1 n + a_0$$

Then

$$T(n) = O(n^m)$$

Big-Omega (Ω) Asymptotic Notation

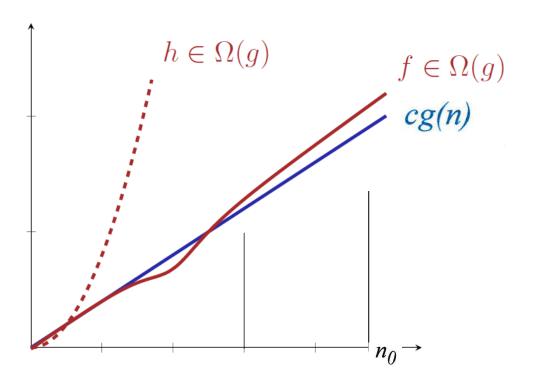
Definition:

$$T(n) = \Omega(g(n)) \Leftrightarrow$$

 $\exists c>0, \exists n_0>0 \text{ such that } \forall n \geq n_0$
 $0 \leq c g(n) \leq T(n)$

g(n) = lower bound for T(n). The number of steps is at least g(n) (constant c doesn't count in asymptotic analysis).

Big-Omega (Ω) Asymptotic Notation



provides the best case complexity of an algorithm

Big-Omega (Ω) Asymptotic Notation

Examples:

$$T(n) = 3n+2 = \Omega(n), \quad \text{c=3 and } n_0 = 1 \qquad \qquad 3n \leq 3n+2 \qquad \forall n \geq 1 \\ T(n) = 10n^2 + 4n + 2 = \Omega(n^2), \quad \text{c=1 and } n_0 = 1 \qquad \qquad n^2 \leq 10n^2 + 4n + 2 \qquad \forall n \geq 1 \\ T(n) = 10n^2 + 4n + 2 = \Omega(n), \quad \text{c=30 and } n_0 = 3 \qquad \qquad 30n \leq 10n^2 + 4n + 2 \qquad \forall n \geq 3$$

Theorem:

if
$$T(n) = a_m n^m + + a_1 n + a_0$$
 then $T(n) = \Omega(n^m)$

Big-Theta (Θ) Asymptotic Notation

Definition:

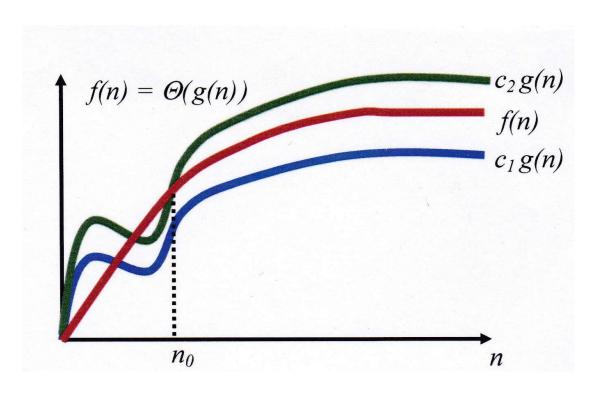
$$T(n) = \Theta(g(n)) \Leftrightarrow$$

$$\exists c_{1,,}c_{2} > 0, \exists n_{0} > 0 \text{ such that } \forall n \geq n_{0}$$

$$0 \leq c_{1} g(n) \leq T(n) \leq c_{2} g(n)$$

g(n) = tight asymptotic bound for <math>T(n). The number of steps is exactly g(n) (constants c_1 and c_2 do not count in asymptotic analysis).

Big-Theta (Θ) Asymptotic Notation



f(n) can be **sandwiched** between $c_1g(n)$ and $c_2g(n)$

Average-case complexity analysis of an algorithm.

Big-Theta (Θ) Asymptotic Notation

Examples:

T(n) =
$$3n+2 = \Theta(n)$$
, $c_1=3$, $c_2=4$ and $n_0=2$ $3n \le 3n+2 \le 4n$ $\forall n \ge 1$ **T(n)** = $3n+2 \ne \Theta(n^2)$, **T(n)** = $10n^2+4n+2 \ne \Theta(n)$

Theorems:

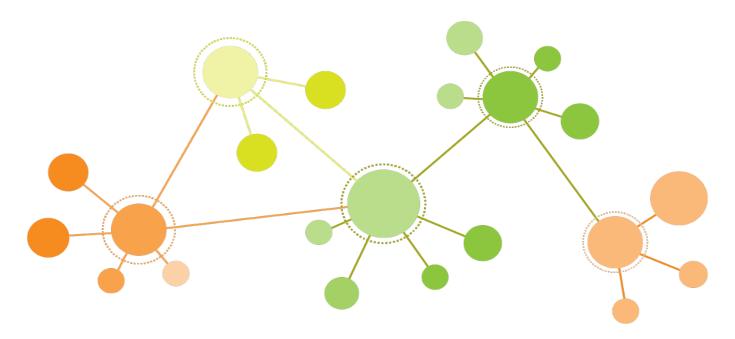
- If $T(n) = a_m n^m + + a_1 n + a_0$, then $T(n) = \Theta(n^m)$
- Let g(n) and T(n) be 2 functions, $T(n) = \Theta(g(n)) \Leftrightarrow T(n) = O(g(n)) \text{ and } T(n) = \Omega(g(n)).$

Big-Theta (Θ) Asymptotic Notation

Exercise:

```
Given f(n), g(n), h(n), k(n):
          if f(n) = \Theta(k(n)) and k(n) = \Theta(g(n)), then g(n) = \Theta(f(n))?
                                   Yes / No
          if f(n) = O(g(n)) and g(n) = O(h(n)), then h(n) = \Omega(f(n))?
                                  Yes / No
```

Real problem to understand the impact of the choice of the **algorithm** and of the **data structure** on **complexity**:



Real problem to understand the impact of the choice of the **algorithm** and of the **data structure** on **complexity**:

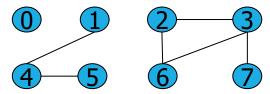
- Undirected graph whose vertices are integers and whose edges are pairs of integers
- Input: sequence of integer pairs (p, q)
- Interpretation: p is connected to q
- Connectivity relation:
 - Reflexive: p is connected to p
 - Symmetrical: if p is connected to q, q is connected to p
 - **Transitive:** if p is connected to q and q is connected to r, then p is connected to r

Thus it is an equivalence relation.

- Output: list of previously unknown connections (or not transitively implied by the previous ones):
 - null if p and q are already connected (directly or indirectly)
 - else (p, q)

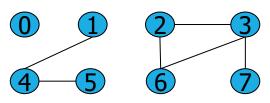
Connected component in an **undirected graph**: maximal subset of mutually reachable nodes

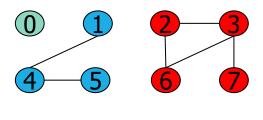
Example:

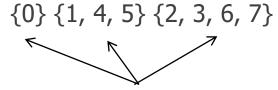


Connected component in an **undirected graph**: maximal subset of mutually reachable nodes

Example:







3 connected components

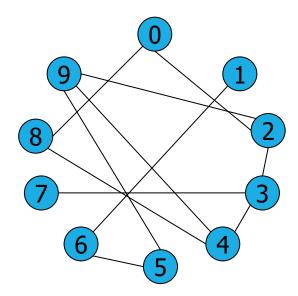
Applications

- Pixels in digital pictures
- Computer networks (computers, links)
- Electrical networks (components, wires)
- Social networks (friends)
- Mathematical sets
- Program variables.

Input sequence:

3-4, 4-9, 8-0, 2-3, 5-6, 2-9, 5-9, 7-3, 4-8, 5-6, 0-2, 6-1

Corrisponding graph:



Let's validate:

Input

3 4

9 1 2 7 6 5 Output

4

Let's validate:

Input

9 4

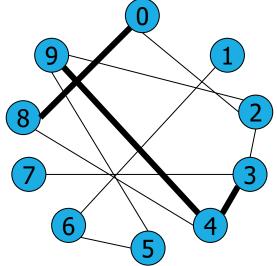
6

Output

) 4

Let's validate:

Input 8 0 Output 8 0

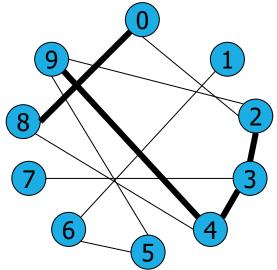


Let's validate:

Input

2 3

Output 2 3



Let's validate:

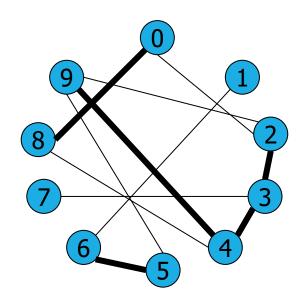
Input 5 6

9 1 2 7 Output 5 6

Let's validate:

Input 2 9

Output



Path 2-3-4-9 already exists

Let's validate:

Input 5 9

Output 5 9

Let's validate:

Input

7 3

9 1 2 7 6 5 Output

3

Let's validate:

Input

4 8

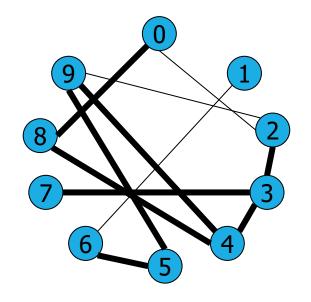
Output 4 8

Let's validate:

Input

5 6

Output



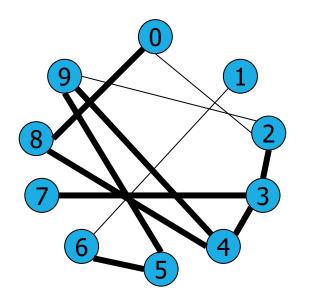
Path 5-6 already exists

Let's validate:

Input

0 2

Output



Path 0-8-4-3-2 already exists

Let's validate:

Input

6 1

Output

1

On-line approach

Assumptions:

- We don't have the graph
- We work online pair by pair, keeping and updating information necessary to find out connectivity.
- Each pair is made of 2 integers in the range from 0 to N-1

Sets S_i of connected pairs, initially as many sets as nodes, each node being connected just to itself.

Abstract operations:

- **find:** find the set an object belongs to
- union: merge two sets

On-line approach

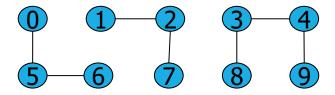
- Algorithm: repeat for all pairs (p, q)
 - read the pair (p, q)
 - execute find on p: find an S_p such that $p \in S_p$
 - execute find on \mathbf{q} : find an S_q such that $\mathbf{q} \in S_q$
 - if S_p and S_q coincide, consider the next pair, otherwise execute union on S_p and S_q

Quick find

Represent sets S_i of connected pairs with array id:

- initially id[i] = i (no connection)
- if p and q are connected, id[p] = id[q]

Example: the following **graph**



is represented like this:



Quick find

Algorithm:

- repeat for all pairs (p, q):
 - read pair (p, q)
 - if pair is connected (id[p] = id[q]),
 - do nothing and move to the next pair,
 - else
 - scan the array, replacing id[p] values with id[q] values

Quick find

- find: simple reference to cell in array id[index], unit cost O(1)
- union: scan array to replace id[p] values with id[q] values, cost linear in array size O(n)
- overall **number of operations** related to # pairs * array size

Tree representation

- Some objects represent the set they belong to
- Other objects point to the the object that represents the set they belong to.



(5)

1

6

(2)

(3)

4

9

Initially

$$S_0 = \{0\}, S_1 = \{1\}, S_2 = \{2\}, S_3 = \{3\}, S_4 = \{4\}$$

 $S_5 = \{5\}, S_6 = \{6\}, S_7 = \{7\}, S_8 = \{8\}, S_9 = \{9\}$

- 0
- $\widehat{1}$
- **(2**)
- 3-4

- **(5)**
- 6
- (7)
- 8
- 9

Input:
$$pq = 34$$

$$id[\mathbf{p}]=3 \neq id[\mathbf{q}]=4$$

$$S_0 = \{0\}, S_1 = \{1\}, S_2 = \{2\}, S_{3-4} = \{3,4\},$$

 $S_5 = \{5\}, S_6 = \{6\}, S_7 = \{7\}, S_8 = \{8\}, S_9 = \{9\}$



$$\widehat{7}$$

Input: pq = 49





$$id[\mathbf{p}]=4 \neq id[\mathbf{q}]=9$$

$$S_0 = \{0\}, S_1 = \{1\}, S_2 = \{2\}, S_{3-4-9} = \{3,4,9\},$$

$$S_5 = \{5\}, S_6 = \{6\}, S_7 = \{7\}, S_8 = \{8\}$$





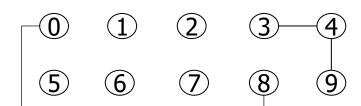












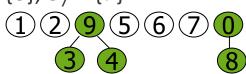
Input:
$$pq = 80$$

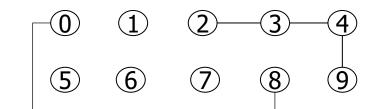


$$id[\mathbf{p}]=8 \neq id[\mathbf{q}]=0$$

$$S_{0-8} = \{0,8\}, S_1 = \{1\}, S_2 = \{2\}, S_{3-4-9} = \{3,4,9\},$$

 $S_5 = \{5\}, S_6 = \{6\}, S_7 = \{7\}$



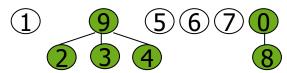


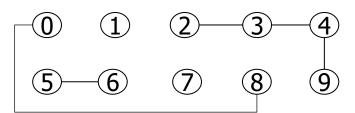
Input:
$$pq = 2$$
 3



$$id[\mathbf{p}]=2 \neq id[\mathbf{q}]=9$$

$$S_{0-8} = \{0,8\}, S_1 = \{1\}, S_{2-3-4-9} = \{2,3,4,9\}, S_5 = \{5\}, S_6 = \{6\}, S_7 = \{7\}$$





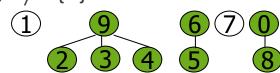
Input:
$$pq = 56$$

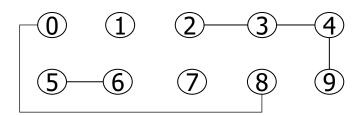


$$id[\mathbf{p}]=5 \neq id[\mathbf{q}]=6$$

$$S_{0-8} = \{0,8\}, S_1 = \{1\}, S_{2-3-4-9} = \{2,3,4,9\},$$

$$S_{5-6} = \{5,6\}, S_7 = \{7\}$$





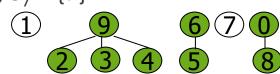
Input:
$$p q = 2 9$$

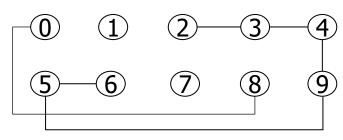
$$id[p]=9 = id[q]=9$$

no change

$$S_{0-8} = \{0,8\}, S_1 = \{1\}, S_{2-3-4-9} = \{2,3,4,9\},$$

$$S_{5-6} = \{5,6\}, S_7 = \{7\}$$



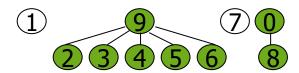


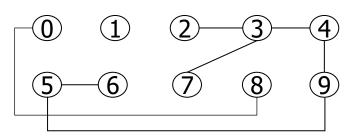
Input: p q = 5 9

id	0	1	9	9	9	6	6	7	0	9
	0	1	2	3	4	5	6	7	8	9

$$id[\mathbf{p}]=6 \neq id[\mathbf{q}]=9$$

$$S_{0-8} = \{0,8\}, S_1 = \{1\}, S_{2-3-4-5-6-9} = \{2,3,4,5,6,9\}, S_7 = \{7\}$$

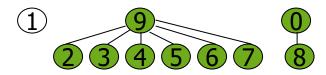


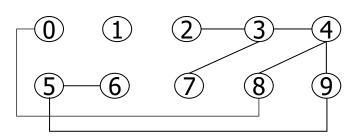


Input: p q = 7 3

$$id[\mathbf{p}]=7 \neq id[\mathbf{q}]=9$$

$$S_{0-8} = \{0,8\}, S_1 = \{1\}, S_{2-3-4-5-6-7-9} = \{2,3,4,5,6,7,9\}$$

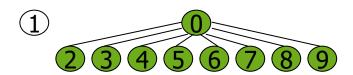


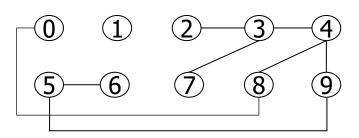


Input: pq = 48

$$id[\mathbf{p}]=9 \neq id[\mathbf{q}]=0$$

$$S_1 = \{1\}, S_{0-2-3-4-5-6-7-8-9} = \{0,2,3,4,5,6,7,8,9\}$$



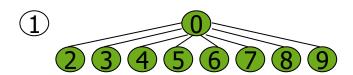


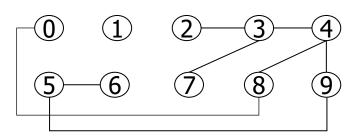
Input: pq = 56

$$id[\mathbf{p}]=0=id[\mathbf{q}]=0$$

no change

$$S_1 = \{1\}, S_{0-2-3-4-5-6-7-8-9} = \{0,2,3,4,5,6,7,8,9\}$$



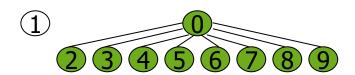


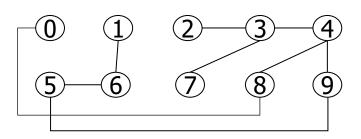
Input: $\mathbf{p} \mathbf{q} = 0 2$

$$id[\mathbf{p}]=0=id[\mathbf{q}]=0$$

no change

$$S_1 = \{1\}, S_{0-2-3-4-5-6-7-8-9} = \{0,2,3,4,5,6,7,8,9\}$$

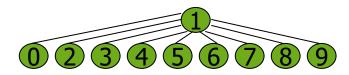




Input: p q = 6 1

$$id[\mathbf{p}]=0 = id[\mathbf{q}]=1$$

$$S_{0-1-2-3-4-5-6-7-8-9} = \{0,1,2,3,4,5,6,7,8,9\}$$



```
#include <stdio.h>
#define N 10000
main() {
 int i, t, p, q, id[N];
 for (i=0; i<N; i++)
   id[i] = i:
 printf("Input pair p q: ");
 while (scanf("%d %d", &p, &q) ==2) {
    if (id[p] == id[q])
      printf("%d %d already connected\n", p,q);
    else {
      for (t = id[p], i = 0; i < N; i++)
        if (id[i] == t)
          id[i] = id[q];
        printf("pair %d %d not yet connected\n", p, q);
      printf("Input pair p q: ");
```

Quick union

Represent sets S_i of connected pairs with an **array id**:

initially all the objects point to themselves

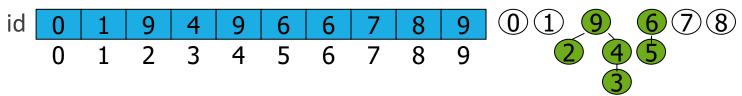
 each object points either to an object to which it is connected or to itself (no loops).

Notation (id[i])* stands for id[id[id[... id[i]]]]

If objects i are j connected

$$(id[i])^* = (id[j])^*$$

Example



Quick union

Algorithm:

- repeat for all the pairs (p, q):
 - read pair (p, q)
 - if(id[p])* = (id[q])*
 - do nothing (the pair is already connected) and move on to the next pair,
 - else
 - id[(id[p])*] = (id[q])* (connect the pair).

Quick union

- **find:** scan a **"chain"** of objects, upper bound linear cost in the number of objects, in general well below upper bound **O(n)**
- union: simple, as it is enough that an object points to another object, unit cost O(1)
- overall number of operations related to

pairs * chain length

- **5**
- 6

- 8

9

Initially



Input:
$$p q = 3 4$$



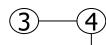
$$id[p]=3 \neq id[q]=4$$







(2)



6

8

9

Input:
$$p q = 4 9$$



$$id[p]=4 \neq id[q]=9$$

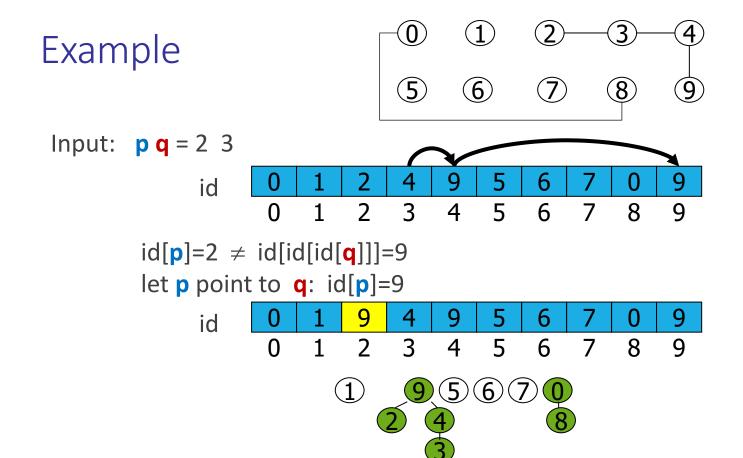


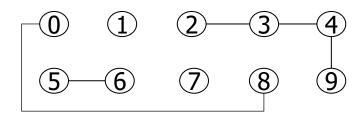




Input:
$$pq = 80$$

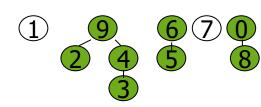
$$id[\mathbf{p}]=8 \neq id[\mathbf{q}]=0$$

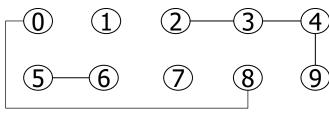




Input: pq = 56

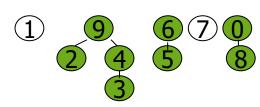
$$id[\mathbf{p}]=5 \neq id[\mathbf{q}]=6$$

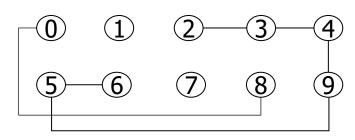




$$id[id[p]]=9 = id[q]=9$$

unchanged

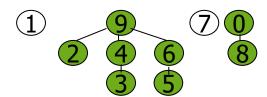


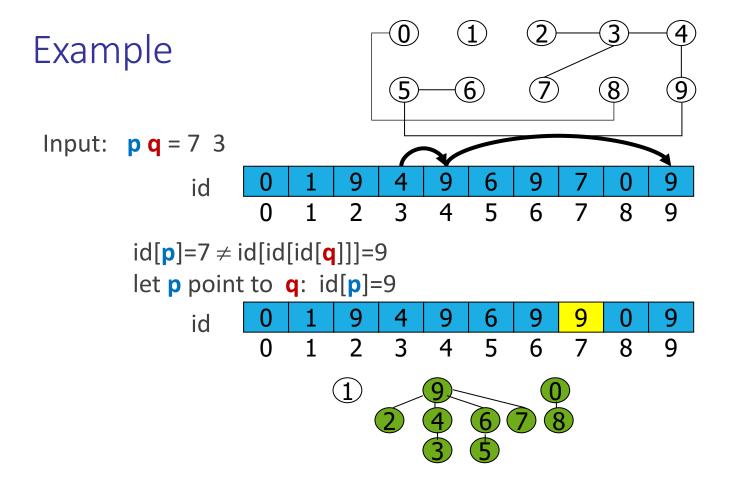


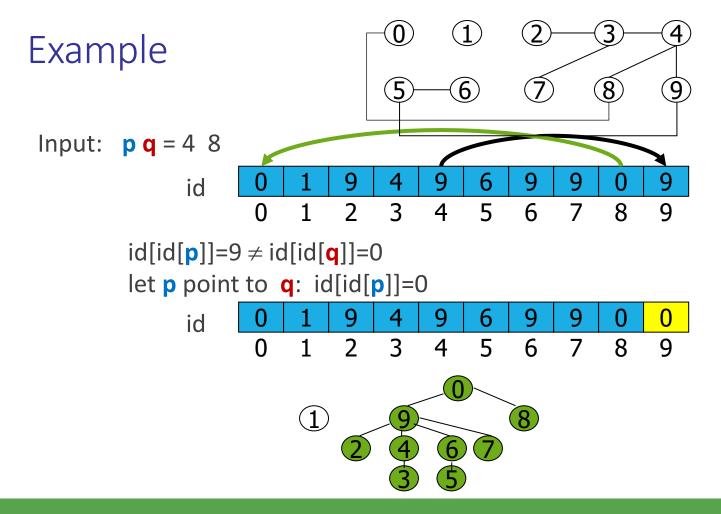
Input: pq = 59

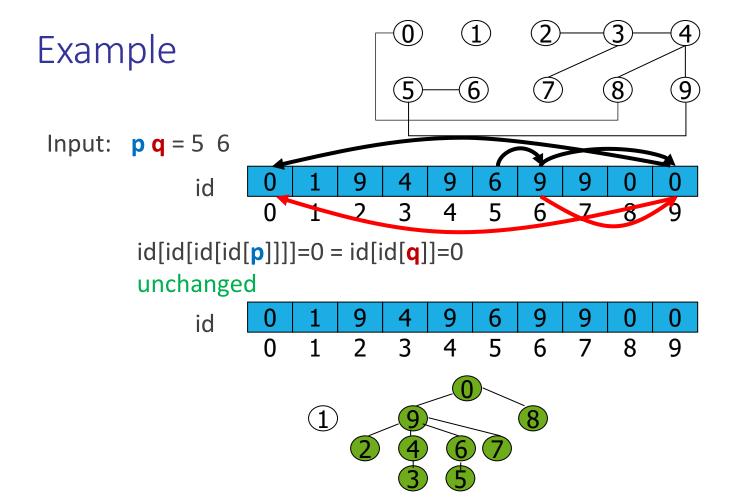


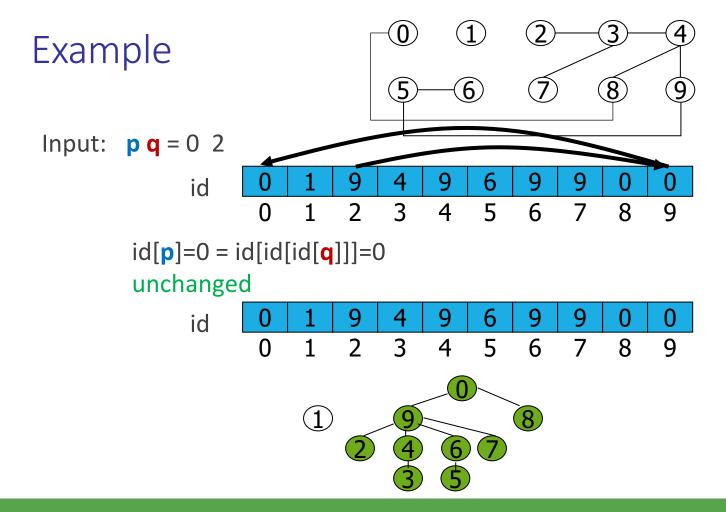
$$id[id[p]]=6 \neq id[q]=9$$

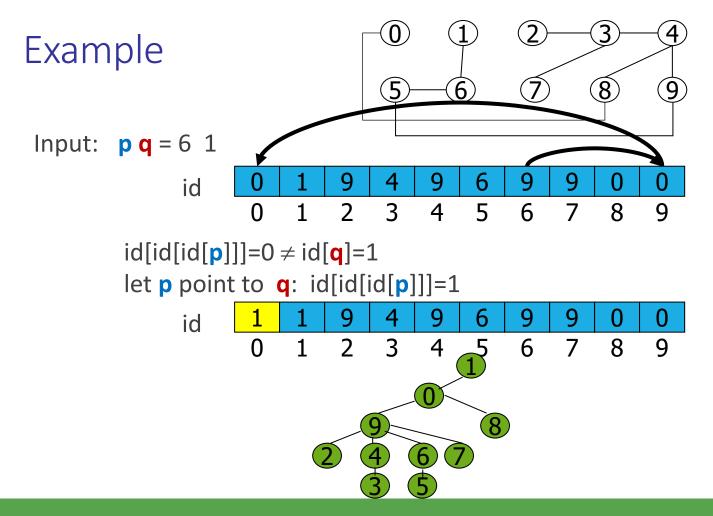










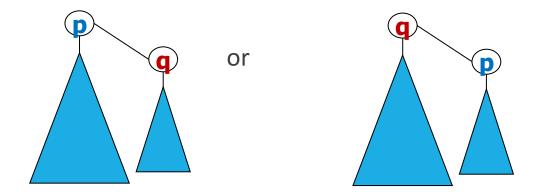


```
#include <stdio.h>
#define N 10000
main() {
 int i, j, p, q, id[N];
  for(i=0; i<N; i++)</pre>
  id[i] = i;
  printf("Input pair p q: ");
  while (scanf("%d %d", &p, &q) ==2) {
    for (i = p; i!= id[i]; i = id[i]);
    for (j = q; j!= id[j]; j = id[j]);
    if (i == j)
      printf("pair %d %d already connected\n", p,q);
    else {
      id[i] = j;
      printf("pair %d %d not yet connected\n", p, q);
    printf("Input pair p q: ");
```

Quick union Optimization

Weighted quick union:

- To shorten the chain's length, keep track of the number of elements in each tree (array SZ) and connect the smaller tree to the larger one.
- According to which one is the larger, there might be 2 solutions:



NB: it doesn't matter whether if p appears at the right or at the left of q.

(0)

5

1

6

- 2
- 3

8

4

9

Initially

id



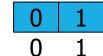
0123456789

Input:
$$p q = 3 4$$



$$id[p]=3 \neq id[q]=4$$

let
$$\mathbf{p}$$
 point to \mathbf{q} : id[\mathbf{p}]=4





1

2

3-4



6

 $\widehat{\mathbf{7}}$

3

Input: p q = 4 9

$$id[\mathbf{p}]=4 \neq id[\mathbf{q}]=9$$

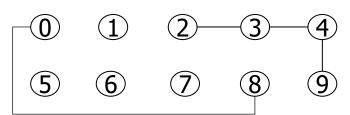
let the smaller tree q pointo to the larger tree p: id[q]=4

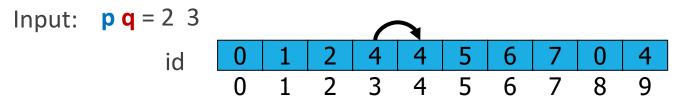


Input:
$$\mathbf{p} \mathbf{q} = 8 \ 0$$

$$id[\mathbf{p}]=8 \neq id[\mathbf{q}]=0$$

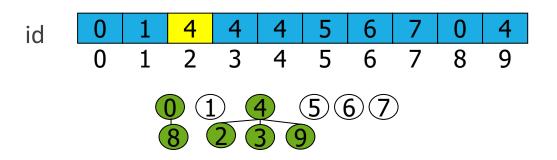


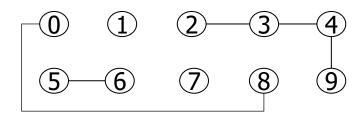




$$id[p]=2 \neq id[id[q]]=4$$

let the smaller tree q pointo to the larger tree p: id[p]=4

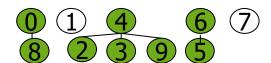


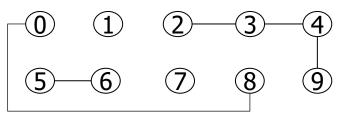


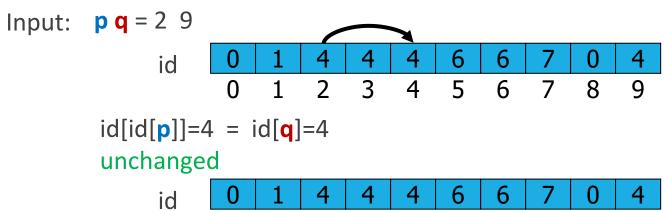
Input: pq = 56

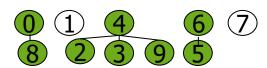


$$id[\mathbf{p}]=5 \neq id[\mathbf{q}]=6$$

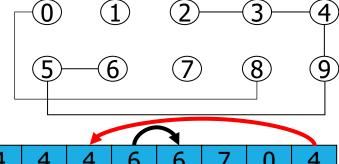








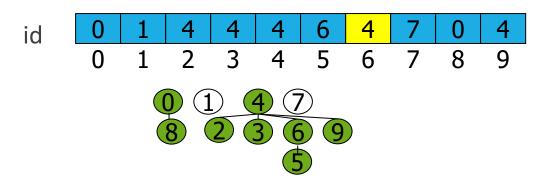




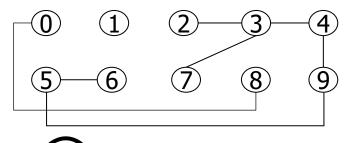
Input: p q = 5 9

$$id[id[p]]=6 \neq id[id[q]]=4$$

let the smaller tree q pointo to the larger tree p: id[id[p]]=4





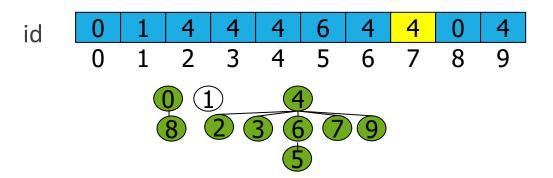


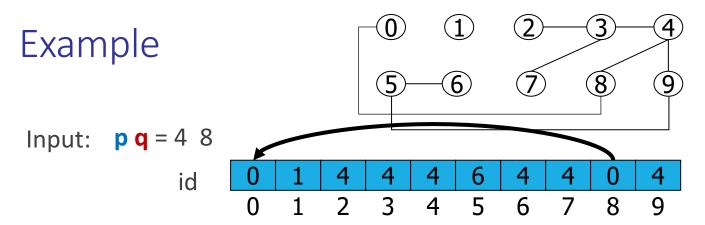
Input:
$$p q = 7 3$$



$$id[p]=7 \neq id[id[q]]=4$$

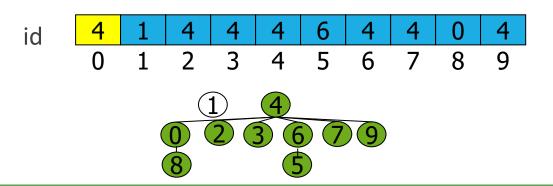
let the smaller tree **q** pointo to the larger tree **p**: id[**p**]=4



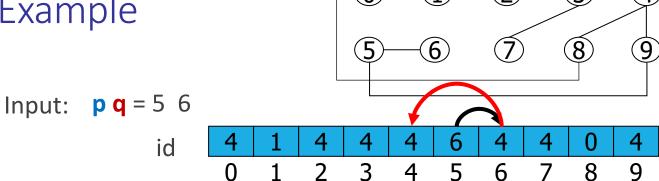


$$id[p]=4 \neq id[id[q]]=0$$

let the smaller tree q pointo to the larger tree p: id[id[q]]=4

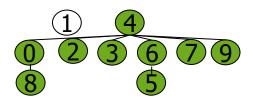


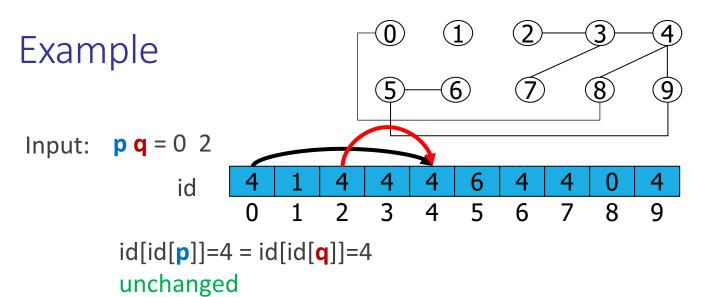


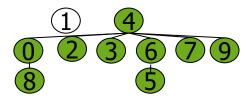


$$id[id[id[p]]]=4 = id[id[q]]=4$$

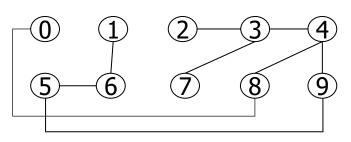
unchanged



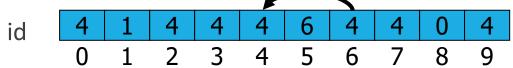






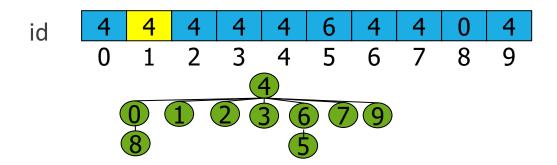


Input: pq = 61



$$id[id[p]]=4 \neq id[q]=1$$

let the smaller tree q pointo to the larger tree p: id[q]=4



```
int i, j, p, q, id[N], sz[N];
for(i=0; i<N; i++) { id[i] = i; sz[i] =1; }
printf("Input pair p q: ");
while (scanf("%d %d", &p, &q) ==2) {
  for (i = p; i!= id[i]; i = id[i]);
  for (j = q; j!= id[j]; j = id[j]);
  if (i == j)
    printf("pair %d %d already connected\n", p,q);
  else {
    printf("pair %d %d not yet connected\n", p, q);
   if (sz[i] <= sz[j]) {
      id[i] = j; sz[j] += sz[i]; }
     else { id[j] = i; sz[i] += sz[j];}
  printf("Input pair p q: ");
```

Quick union Optimization

- find: scanning a "chain" of objects, cost at most logarithmic in the number of objects O(logn)
- union: simple, because it is enough that an object points to another object, unit cost O(1)
- globally the number of operations is bounded by

numb. of pairs * "chain" length

but the chain's length grows logarithmically!

Why logatithmic?

Worst-case: given **n** elements, each union connects **2** trees of the same size



0 2 4 6 8

$$p q = 0 1$$

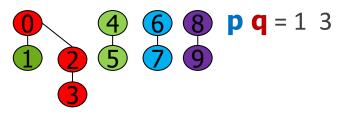
$$p q = 2 3$$

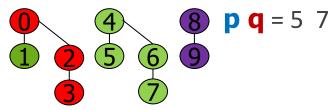
$$p q = 4 5$$

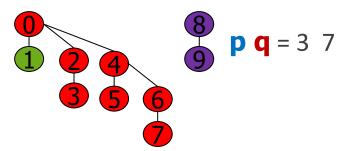
$$p q = 6 7$$

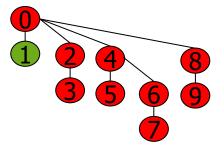
$$p q = 8 9$$

Why logatithmic?









Why logatithmic?

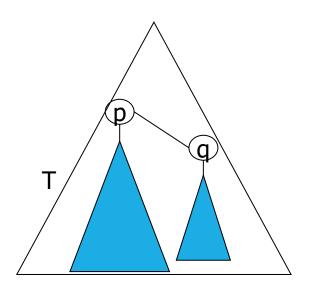
Each **tree** containing **2**^h **nodes** has height **h**.

With a union operation, in the worst case, we merge 2 trees with the same number of nodes 2^h. The result is a tree with 2^{h+1} nodes, thus its height is h+1.

Height grows linearly with the number of union operations.

How many union operations are required?

If $T_1 \ge T_2$, each time we merge a smaller tree into a larger one, we create a tree whose size **T** is at least **twice** the size of T_2 .



If, at each step, the number of elements in the tree doubles at least and if there are **N** elements, after **i** steps there will be at least **2**ⁱ elements in the tree.

As the inequality $2^i \le N$ holds, the number of union operations required is $i \le log_2N$.