

**Algorithm:**MERGE (A,  $p$ ,  $q$ ,  $r$ )

1.  $n1 \leftarrow q - p + 1$
- $n2 \leftarrow r - q$
- Create arrays  $L[1 \dots n1 + 1]$  and  $R[1 \dots n2 + 1]$
- for  $i \leftarrow 1$  to  $n1$
- do  $L[i] \leftarrow A[p + i - 1]$
- For  $j \leftarrow 1$  to  $n2$
- do  $R[j] \leftarrow A[q + j]$
8.  $L[n1 + 1] \leftarrow \infty$
9.  $R[n2 + 1] \leftarrow \infty$
10.  $i \leftarrow 1$
- ← 11.  $j \leftarrow 1$
- for  $k \leftarrow p$  to  $r$
- do if  $L[i] \leq R[j]$
- then  $A[k] \leftarrow L[i]$

15.	$i$	$i + 1$
16.	Else $A[k]$	$R[j]$
17.	$j$	$j + 1$

We implement it so that it takes  $\Theta(n)$  time, where  $n = r - p + 1$ , which is the number of elements being merged.

## ALGORITHM:

**PARTITION** ( $A, p, r$ )

1.  $x \leftarrow A[p]$
2.  $i \leftarrow p-1$

3.  $j \leftarrow r+1$ 
  - while TRUE do
  - Repeat  $j \leftarrow j-1$
  - until  $A[j] \leq x$
  - Repeat  $i \leftarrow i+1$
  - until  $A[i] \geq x$
  - if  $i < j$
  - then exchange  $A[i] \leftrightarrow A[j]$
  - else return  $j$

Partition selects the first key,  $A[p]$  as a pivot key about which the array will be partitioned: Keys  $\leq A[p]$  will be moved towards the left.

Keys  $\geq A[p]$  will be moved towards the right.

**QuickSort** ( $A, p, r$ )

- If  $p < r$  then
- $q \leftarrow \text{Partition}(A, p, r)$
- Recursive call to Quick Sort ( $A, p, q$ )
- Recursive call to Quick Sort ( $A, q+1, r$ )

## ALGORITHM:

**Greedy-fractional-knapsack** ( $w, v, W$ )

```
FOR  $i = 1$  to  $n$ 
  do  $x[i] = 0$ 
weight = 0
while weight <  $W$ 
  do  $i =$  best remaining item
  IF weight +  $w[i] \leq W$ 
    then  $x[i] = 1$ 
    weight = weight +  $w[i]$ 
  else
     $x[i] = (W - \text{weight}) / w[i]$ 
    weight =  $W$ 
return  $x$ 
```

**Analysis:** If the items are already sorted into decreasing order of  $v_i / w_i$ , then the while-loop takes a time in  $O(n)$ . Therefore, the total time including the sort is in  $O(n \log n)$ .

## KRUSKAL ALGORITHM:

Start with an empty set  $A$ , and select at every stage the shortest edge that has not been chosen or rejected, regardless of where this edge is situated in the graph.

```
KRUSKAL( $G$ ) :  
1  $A = \emptyset$   
2 foreach  $v \in G.V$ :  
3   MAKE-SET( $v$ )  
4 foreach  $(u, v)$  ordered by weight( $u, v$ ), increasing:  
5   if FIND-SET( $u$ )  $\neq$  FIND-SET( $v$ ):  
6      $A = A \cup \{(u, v)\}$   
7     UNION( $u, v$ )  
8 return  $A$ 
```

- Make\_SET( $v$ ): Create a new set whose only member is pointed to by  $v$ . Note that for this operation  $v$  must already be in a set.

## PRIMS ALGORITHM:

- Initialize a tree with a single vertex, chosen arbitrarily from the graph.
- Grow the tree by one edge: Of the edges that connect the tree to vertices not yet in the tree, find the minimum-weight edge, and transfer it to the tree.
- Repeat step 2 (until all vertices are in the tree).

```
MST-PRIM( $G, w, r$ )  
1 for each  $u \in V[G]$   
2   do  $key[u] \leftarrow \infty$   
3    $\pi[u] \leftarrow \text{NIL}$   
4  $key[r] \leftarrow 0$   
5  $Q \leftarrow V[G]$   
6 while  $Q \neq \emptyset$   
7   do  $u \leftarrow \text{EXTRACT-MIN}(Q)$   
8   for each  $v \in \text{Adj}[u]$   
9     do if  $v \in Q$  and  $w(u, v) < key[v]$   
10      then  $\pi[v] \leftarrow u$   
11       $key[v] \leftarrow w(u, v)$ 
```

**Time Complexity:** A simple implementation using an [adjacency matrix](#) graph representation and searching an array of weights to find the minimum weight edge to add requires  $\mathcal{O}(V^2)$  running time. Using a simple [binary heap](#) data structure and an [adjacency list](#) representation, Prim's algorithm can be shown to run in time  $\mathcal{O}(E \log V)$  where  $E$  is the number of edges and  $V$  is the number of vertices.

## ALGORITHM:

Algorithm All pair shortest path:

```
let dist be a  $|V| \times |V|$  array of minimum distances initialized to  $\infty$  (infinity)  
2 for each vertex  $v$   
3   dist[v][v]  $\leftarrow 0$   
4 for each edge  $(u, v)$   
5   dist[u][v]  $\leftarrow w(u, v)$  // the weight of the edge  $(u, v)$   
6 for  $k$  from 1 to  $|V|$   
7   for  $i$  from 1 to  $|V|$   
8     for  $j$  from 1 to  $|V|$   
9       if dist[i][j]  $>$  dist[i][k] + dist[k][j]  
10        dist[i][j]  $\leftarrow$  dist[i][k] + dist[k][j]  
11     end if
```

This is an  $\mathcal{O}(n^3)$  algorithm, where  $n$  is the number of vertices in the digraph.

## ALGORITHM:

### Iterative LCS:

```
int lcs_length(char * A, char * B)
{
    allocate storage for array L;
    for (i = m; i >= 0; i--)
        for (j = n; j >= 0; j--)
        {
            if (A[i] == '\0' || B[j] == '\0') L[i,j] = 0;
            else if (A[i] == B[j]) L[i,j] = 1 + L[i+1, j+1];
            else L[i,j] = max(L[i+1, j], L[i, j+1]);
        }
    return L[0,0];
}
```

### The subsequence itself

What if you want the subsequence itself, and not just its length? This is important for some but not all of the applications we mentioned. Once we have filled in the array L described above, we can find the sequence by working forwards through the array.

```
sequence S = empty;
i = 0;
j = 0;
while (i < m && j < n)
```

```
{
    if (A[i]==B[j])
    {
        add A[i] to end of S;
        i++; j++;
    }
    else if (L[i+1,j] >= L[i,j+1]) i++;
    else j++;
}
```

So we can find longest common subsequences in time  $O(mn)$ .

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            else L[i,j] = max(L[i+1, j], L[i, j+1]);
        }
    return L[0,0];
}
```

Advantages of this method include the fact that iteration is usually faster than recursion, we don't need to initialize the matrix to all -1's, and we save three if statements per iteration since we don't need to test whether  $L[i, j]$ ,  $L[i+1, j]$ , and  $L[i, j+1]$  have already been computed (we know in advance that the answers will be no, yes, and yes). One disadvantage over memorizing is that this fills in the entire array even when it might be possible to solve the problem by looking at only a fraction of the array's cells.

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**NAIVE ALGORITHM:**

Generate all possible configurations of queens on board and print a configuration that satisfies the given constraints.

```

while there are untried configurations
{
    generate the next configuration
    if queens don't attack in this configuration then
    {
        print this configuration;
    }
}

```

**BACKTRACKINGALGORITHM:**

Algorithm N\_QUEEN (k,n)

```

for i= 1 to n do
    if PLACE (k,i)then
        x[k]=i
        if k==n then
            print X[1...n]
        else
            N_QUEEN(K+1,n)
        end
    end
end

```

**Function PLACE (K,i)**

//k is the number of queen being processed

//i is the number of columns

for j=1 to k-1 do

if x[j]==i then

return false

end

end

return true

**COMPLEXITY ANALYSIS:**

The recurrence of n-Queen problem is defined as  $T(n)=n*T(n-1)+n^2$ . Solution to recurrence would be  $O(n!)$

**NAIVEALGORITHM:**

Generate all possible configurations of colors and print a configuration that satisfies the given constraints.

```

while there are untried configurations
{
    generate the next configuration
    if no adjacent vertices are colored with same color
    {
        print this configuration;
    }
}

```

There will be  $v^m$  configurations of colors.

**BACKTRACKINGALGORITHM:**

The idea is to assign colors one by one to different vertices, starting from the vertex 0. Before assigning a color, we check for safety by considering already assigned colors to the adjacent vertices. If we find a color assignment which is safe, we mark the color assignment as part of solution. If we do not find a color due to clashes then we backtrack and return false.

**ALGORITHM:**

GRAPH\_COLORING (G, COLOR, i)

```

If CHECK_VERTEX(i) == i then
    if i == N then
        print COLOR[1.....n]
    else
        j = 1
        while (j <= M) do
            COLOR (i+1) = j
            j = j+1
        end
    end
end
Function CHECK_VERTEX (i)
for j = 1 to i-1 do
    if Adjacent (i,j) then
        if COLOR(i) == COLOR(j) then
            return 0
        end
    end
end
return 1

```

**ALGORITHM:**

NAIVE\_STRING\_MATCHER (T,P)

```

n = length[T]
m = length[P]
for s = 0 to n-m do
    if P[1....m] = T[(s+1)....(s+m)] then
        print "pattern occurs with shift s"
    end
end

```

**ANALYSIS:** In best case searchable text does not contain any of prefix of pattern .only one comparison requires moving pattern one position right. Worst case occurs when pattern is at last position and there are spurious hits all the ways.

**ALGORITHM:**

```
//T is the text of length n
// P is the pattern of length m
Rabin karp (string T[1...n], string P[1...m])
    hash T = hash (T[1...n])
    hash P = hash (P[1...m])
    for i=1 to n-m+1
```

```
        if hash T= hash P then
            if P[1...m]=T[ s+1 .....s+m] then
                else
            print match found at shift S
            match not found
```

**ANALYSIS:** In most cases it runs in linear time ie:  $O(n)$  and worst case is  $O(mn)$  it can happen only when prime no. used for hashing is very small.