

DYNAMIC PROGRAMMING & OTHER ALGORITHMS

Dynamic Programming is mainly an optimization over plain recursion.

The idea is to simply store the results of subproblems, so that we do not have to re-compute them when needed later.

EX: FIBONACCI

SEQUENCE 1, 1, 2, 3, 5, 8, 13, 21,

...

FIBONACCI IN CODE:

```
void fib ()  
  
{  
  
    fibresult[0] = 1;  
  
    fibresult[1] = 1,  
  
    for (int i = 2; i<n; i++)  
  
        fibresult[i] = fibresult[i-1] + fibresult[i-2];  
  
}
```

DYNAMIC PROGRAMMING VS GREEDY ALGORITHM

GREEDY ALGORITHM

In a greedy Algorithm,
we make whatever

choice seems best at the
moment in the hope that
it will lead to global
optimal solution.

DYNAMIC PROGRAMMING

In Dynamic Programming we make decision at each step considering current problem and solution to

previously solved sub problem to calculate optimal solution.

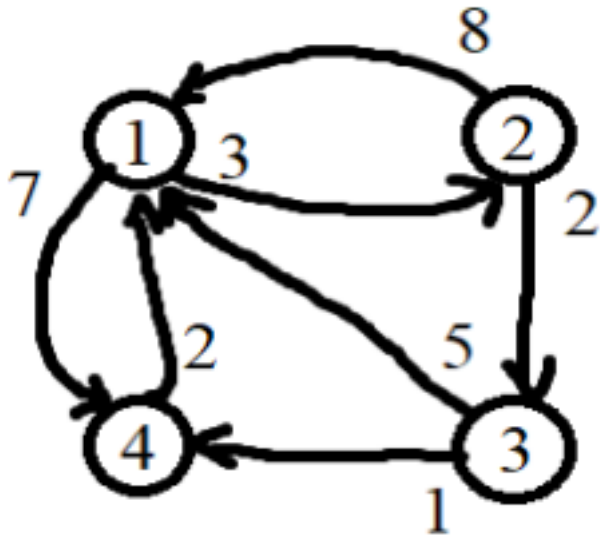
FLOYD-WARSHALL ALGORITHM

The Floyd-Warshall Algorithm is for solving all pairs shortest path problems. The problem is to find the shortest distances between every pair of vertices in a given edge-weighted directed Graph.

FLOYD-WARSHALL ALGORITHM

Given Graph:

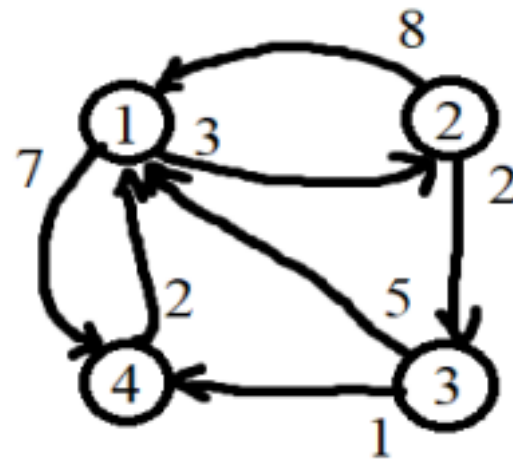
edges



vertex

vertices

HOW TO FIND THE SHORTEST PATH?



$$A^0 = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{bmatrix} 0 & 3 & \infty & 7 \\ 8 & 0 & 2 & \infty \\ 5 & \infty & 0 & 1 \\ 4 & \infty & \infty & 0 \end{bmatrix} \end{matrix}$$



Step
1:
Creat
e a
matrix
 A_0 of
dimen
sion
 $n \times n$
where
 n is
the

number of vertices. The row and the column are indexed as i and j respectively. i and j are the vertices of the graph.

Each cell $A[i][j]$ is filled with the distance from the i th vertex to the j th vertex. If there is no path from i th vertex to j th vertex, the cell is left as infinity.

$$A^1 = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{bmatrix} 0 & 3 & \infty & 7 \\ 8 & 0 & 2 & 15 \\ 5 & 8 & 0 & 1 \\ 2 & 8 & \infty & 0 \end{bmatrix} \end{matrix}$$

$$A^1 = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{bmatrix} 0 & 3 & \infty & 7 \\ 8 & 0 & & \\ 5 & & 0 & \\ 2 & & & 0 \end{bmatrix} \end{matrix}$$



Step 2: Now, create a matrix A_1 using matrix A_0 . The elements in the first column and the first row are left as they are. The remaining cells are filled in the following way.

Let k be the intermediate vertex in the shortest path from source to destination. In this step, k is the first vertex. $A[i][j]$ is

filled with $(A[i][k] + A[k][j])$ if $(A[i][j] > A[i][k] + A[k][j])$.

That is, if the direct distance from the source to the destination is greater than the path through the vertex k , then the cell is filled with $A[i][k] + A[k][j]$.

$$A^0 = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{bmatrix} 0 & 3 & \infty & 7 \\ 8 & 0 & 2 & \infty \\ 5 & \infty & 0 & 1 \\ 4 & \infty & \infty & 0 \end{bmatrix} \end{matrix}$$

$$A^1 = \begin{array}{c|cccc} & 1 & 2 & 3 & 4 \\ \hline 1 & 0 & 3 & \infty & 7 \\ 2 & 8 & 0 & 2 & 15 \\ 3 & 5 & 8 & 0 & 1 \\ 4 & 2 & 8 & \infty & 0 \end{array}$$

$$A^1 = \begin{array}{c|cccc} & 1 & 2 & 3 & 4 \\ \hline 1 & 0 & 3 & \infty & 7 \\ 2 & 8 & 0 & & \\ 3 & 5 & & 0 & \\ 4 & 2 & & & 0 \end{array}$$



SOURCE CODE:

```
for(i=1; i<=n; i++)
{
    for (j=1; j<=n; j++)
    {
        A[l,j] =min(A[i,j] , A[i,k] + A[k,j]); }
    }
```


LET US
UNDERSTAND LCS
WITH AN EXAMPLE.

THE FOLLOWING STEPS
ARE FOLLOWED FOR
FINDING THE LONGEST
COMMON SUBSEQUENCE.

1. Create a table of dimension $n+1*m+1$ where n and m are the lengths of X and Y respectively.

The first row and the first column are filled with zeros.

3. If the character corresponding to the current row and current column are matching, then fill the current cell by adding one to the diagonal element. Point an arrow to the diagonal cell.

5. Step 2 is repeated until the table is filled.

Thus, the longest
common
subsequence is CA.

CONSTRAINTS

**APPLICATIONS OF
BACKTRACKING**

SUM OF SUBSET PROBLEM

GRAPH COLORING

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Subset Sum Problem Solution using
Backtracking Algorithm

The main idea is to add the number to the
stack and track the sum of stack values.

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GRAPH COLORING

A graph coloring is an
assignment of labels traditionally called
"colors" to each vertex.

Example:

In this approach, we color a single vertex
and then move to its adjacent (connected)
vertex to color it with different color. After

coloring, we again move to another adjacent vertex that is uncolored and repeat the process until all vertices of the given graph are colored.

GRAPH COLORING

CONSTRAINTS

APPLICATIONS OF BACKTRACKING

SUM OF SUBSET PROBLEM

	C1	C2	C1	C2	C1	C2
--	----	----	----	----	----	----

CONSTRAINTS

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**SUM OF SUBSET
PROBLEM**

**GRAPH
COLORING**

Hamilton Path is a path that goes through every Vertex of a graph exactly once.

A Hamilton Circuit is a Hamilton Path that begins and ends at the same vertex.

Example:

How does hashing work:

- It uses functions or algorithms to map object data to a representative integer value.

-A
hash
can
then
be
used
to
narrow down
searches when
locating these
items on that
object data map.

A problem-solving technique that works by maintaining a window, a contiguous part of the data. Its disadvantage is that if we want to calculate something about the next window, we don't need to do it from scratch, we just take the value of the actual window and update it in some way, which is faster.

Rabin-Karp algorithm is an algorithm used for searching/matching patterns in the text using a hash function. (

Hashing the process of converting a key to a fixed-size value)

Unlike Naive string-matching algorithm, it does not travel through every character in the

initial
phase
e
rather
it
filters
the

characters that do not match and
then performs the
comparison.

A

sequence of characters is taken and
checked for the possibility
of the presence of the required string. If the possibility is found
then, character matching is performed.

Example:

Given a string s of size n and a pattern p of size m characters of s with p

We have “string s ” of n characters, and “string p ” of m characters.

Two equal ‘string’ must have the same number of characters, m in this case because we are searching for p , So, the idea is to compare all substring of s of m characters with p .

We have “string s” of n characters, and “string p” of m characters.

‘abcbafdeabaccabebf’.

Two equal ‘string’ must have the same number of characters, m in this case because we are searching for p,

So, the idea is to compare all substring of s of m characters with p.

Example. Values of ‘s’ and ‘p’, we are finding the value ‘cab’ in