

Fundamental Algorithmic Techniques V

February 3, 2026



Outline

Dynamic Programming

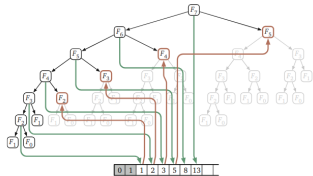
Optimal Substructure

Overlapping Subproblems

Fibonacci and Memoization

Memoized $\mathcal{O}(n)$, space $\mathcal{O}(n)$

```
1: function ITERFIBO1( $n$ )
2:    $F[0] \leftarrow 0$ 
3:    $F[1] \leftarrow 1$ 
4:   for  $i = 2$  to  $n$  do
5:      $F[i] \leftarrow F[i-1] + F[i-2]$ 
6:   end for
7:   return  $F[n]$ 
8: end function
```



Bottom-up $\mathcal{O}(n)$, space $\mathcal{O}(1)$

```
1: function ITERFIBO2( $n$ )
2:    $prev \leftarrow 1$ 
3:    $curr \leftarrow 0$ 
4:   for  $i = 1$  to  $n$  do
5:      $next \leftarrow curr + prev$ 
6:      $prev \leftarrow curr$ 
7:      $curr \leftarrow next$ 
8:   end for
9:   return  $curr$ 
10: end function
```

Matrix iteration $\mathcal{O}(\log n)$ (rep. squaring)

$$\begin{bmatrix} F_{n+1} \\ F_n \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}^n \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

Fibonacci identities:

$$F_{2n} = \dots = F_n(2F_{n-1} + F_n)$$

$$F_{2n-1} = \dots = F_{n-1} + F_n^2$$

Edit Distances

compare s_1 , s_2 with operations (cost 1):

- 1 insert
- 2 remove
- 3 replace

Naive Algo: example of overlappings!

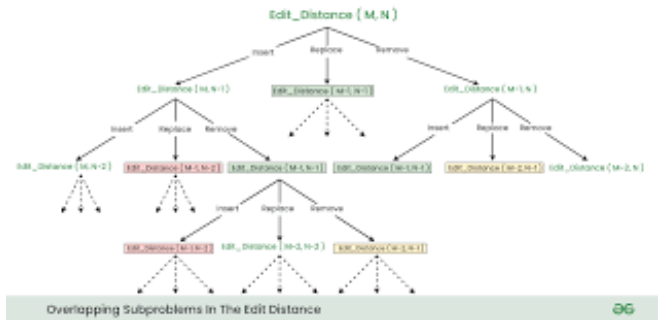
- 1 Last character of s_1 , s_2 are same:

$$\text{ED}(s_1, s_2, m, n) = \text{ED}(s_1, s_2, m-1, n-1)$$

- 2 $\text{ED}(s_1, s_2, m, n) = 1 + \min(\text{ED}(s_1, s_2, m, n-1),$
 $\text{ED}(s_1, s_2, m-1, n),$
 $\text{ED}(s_1, s_2, m-1, n-1))$

time complexity: $\mathcal{O}((3)^{n_1+n_2})$ and $\mathcal{O}(n_1 \cdot n_2)$

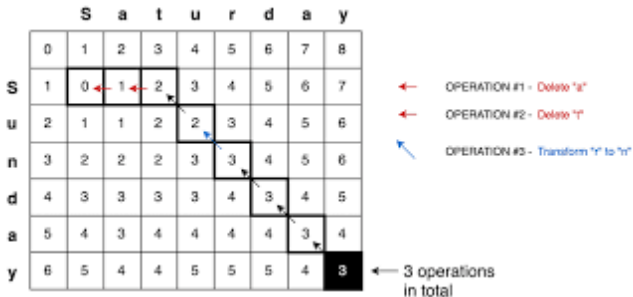
Edit Distances : Memoisation



Memoization Strategies:

- Top-Down ED **Memoisation** - $\mathcal{O}(m \cdot n)$ time and $\mathcal{O}(m \cdot n)$ space
- Bottom-Up ED **Tabulation** - $\mathcal{O}(m \cdot n)$ time and $\mathcal{O}(m \cdot n)$ space

Edit Distances : Tabulation



Costs with: insert, remove, replace
Result bottom right!

Other name: **Levenshtein Distance**

Optimal Substructure

Optimal substructure if an optimal solution constructed from optimal solutions of its subproblems.

Examples:

- shortest path on road or graph
- Fibonacci: $F(n) = F(n) + F(n - 1)$
- rod sold at prices for subsets

Counterexamples:

- shortest path on a Graph without passing twice same node

If optimal Substructure, you can write:

$$OPT(n) = \min(OPT(n - 1), OPT(n - 2), \dots) + f(n)$$

Dynamical Programming \Leftrightarrow Optimal Substructures

Overlapping Subproblems

overlapping subproblems if the same subproblem is solved multiple time Classic examples:

- 1 Fibonacci numbers
- 2 shortest paths
- 3 knapsack problem
- 4 edit distance.

Overlapping Subproblems \Leftrightarrow memoization/tabulation/others!
No Overlapping \Leftrightarrow No Dynamic Programming!

0/1 Knapsack Problem

- A knapsack with integer capacity $W > 0$,
- n items, where item i has:
 - weight $w_i \in \mathbb{Z}^+$,
 - value $v_i \in \mathbb{R}^+$.

Each item may be taken at most once.

Goal: Maximize total value, weight $\leq W$!



- Overlapping subproblems & Optimal Substructure

Idea: $ks[i][w]$ = maximum value achievable using the first i items with capacity w ,

- Dynamic programming applies!

complex problem: solvable by simple bottom up tabulation

Iteration:

$$ks[i][w] = \begin{cases} ks[i-1][w], & \text{if } w_i > w, \\ \max \left(ks[i-1][w], ks[i-1][w - w_i] + v_i \right), & \text{if } w_i \leq w. \end{cases}$$

0/1 Knapsack by Tabulation

Capacity: $W = 8$

Items:

i	w_i	v_i
1	2	3
2	3	4
3	4	5
4	5	6

Table $ks[i][w]$:

$i \backslash w$	0	1	2	3	4	5	6	7	8
0	0	0	0	0	0	0	0	0	0
1	0	0	3	3	3	3	3	3	3
2	0	0	3	4	4	7	7	7	7
3	0	0	3	4	5	7	8	9	9
4	0	0	3	4	5	7	8	9	10

Optimal value: $ks[4][8] = 10$

Selected items: **2 and 4**

(weight: $3 + 5 = 8$, value: $4 + 6 = 10$)

- Time complexity: $O(nW)$
- Space complexity: $O(nW)$ (optimizable to $O(W)$)