

# Dynamic Programming

Computing Lab

<http://www.isical.ac.in/~dfs/lab>

# General structure

## Key idea

1. Break given **problem** into **subproblems**, and **subproblems** into even **smaller subproblems**.
2. Solve a **subproblem** using the answers to **smaller subproblems**.
3. Solve the given **problem** using the answers to the **subproblems**.

## Challenge

Identifying the **problem**  $\longrightarrow$  **subproblems**  $\longrightarrow$  **smaller subproblems** structure.

**Related ideas:** (more about DP vs. these ideas later)

- Divide and conquer
- Recursion

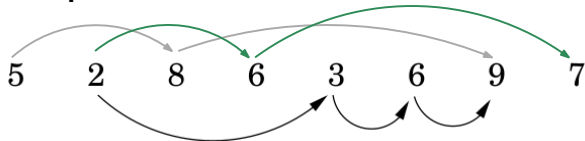
- 1 Longest increasing subsequence
- 2 Edit / Levenstein distance
- 3 Matrix chain multiplication
- 4 Maximum independent set in trees

# Definition

**Input:** a sequence of numbers  $a_1, \dots, a_n$ .

**Increasing subsequence:**  $a_{i_1}, a_{i_2}, \dots, a_{i_k}$  where  $i_1 < i_2 < \dots < i_k$  and  $a_{i_1} < a_{i_2} < \dots < a_{i_k}$ .

**Example:**



Increasing subsequences: 5, 8, 9 (grey)    2, 6, 7 (green)

Longest increasing subsequence: 2, 3, 6, 9 (black)    (also 2, 3, 6, 7)

# Approach

**Case I:**  $n = 1 \rightarrow$  trivial

**Case II:**  $n = 2 \rightarrow$  need to check whether  $a_2 > a_1$

**Case III:**  $n = 3$

**Case III A:**  $a_2 < a_1 \rightarrow$  candidates:  $\langle a_1 \rangle$   $\langle a_2 \rangle$

**Case III B:**  $a_2 = a_1 \rightarrow$  uninteresting; ignored

**Case III A:**  $a_2 > a_1 \rightarrow$  candidates:  $\langle a_1 \rangle$   $\langle a_2 \rangle$   $\langle a_1, a_2 \rangle$

**Question:** Can we forget *any* of the candidates so far?

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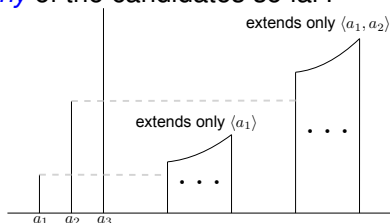
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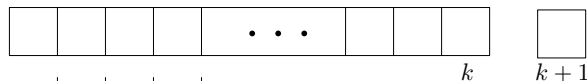
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**Question:** Can we forget *any* of the candidates so far?

**Answer: NO!** Consider



Problem  $\rightarrow$  subproblem  $\rightarrow$  smaller subproblem structure

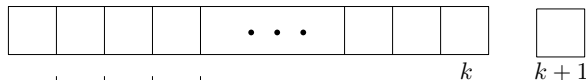


Store the LIS **up to and including**  $i = 0, 1, 2, \dots$  (how?)

**Question:** What is the LIS that ends at  $k + 1$ ?

**Sub-question:** Which increasing sequences end at  $k + 1$ ?

# Problem $\rightarrow$ subproblem $\rightarrow$ smaller subproblem structure



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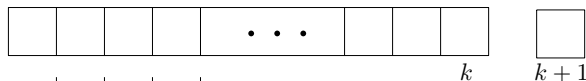
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**Sub-question:** Which increasing sequences end at  $k + 1$ ?

**Algorithm:**

```
for (k = 1; k < ac-1; k++)
  for (j = 0; j < k; j++)
    if (A[k] > A[j] &&
        length[k] < length[j] + 1) { /* try <= instead of < */
      length[k] = length[j] + 1;
      predecessor[k] = j;
    }
```

# Problem $\rightarrow$ subproblem $\rightarrow$ smaller subproblem structure



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        }
```

How to make this more efficient?

# Some implementation details

```
k--;  
max_length = length[k];  
end_of_lis = k--;  
while (k >= 0) {  
    if (length[k] > max_length) { /* try >= instead of > */  
        max_length = length[k];  
        end_of_lis = k;  
    }  
    k--;  
}
```

## Some more implementation details

```
/* Print the actual LIS: sorry this is confusing */
l = length[end_of_lis];
printf("Longest increasing subsequence (%d elements): ", l);
for (i = l-1, j = end_of_lis;
     j != -1 && i >= 0;
     i--, j = predecessor[j])
    /* Being lazy, sorry: reusing the length array to store actual
lis */
    length[i] = j;
assert(i == -1 && j == -1);
for (i = 0; i < l; i++)
    printf("A[%d] = %d, ", length[i], A[length[i]]);
```

- 1 Longest increasing subsequence
- 2 Edit / Levenstein distance**
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# Definition

**Input:** two strings,  $s = s_1s_2s_3 \dots s_m$  and  $t = t_1t_2t_3 \dots t_n$ .

(Indexing starts from 1 to simplify discussion; code will use indexing from 0 as usual.)

**Edit distance:** minimum no. of insertions, deletions, substitutions  
requimygreen to convert  $s$  to  $t$

≡ minimum cost of *aligning*  $s$  and  $t$  using insertions, deletions, substitutions.

## Examples:

---

a	s	t	r	o	n	o	m	e	r	_	_	_	_	
								S	S	I	I	I	I	Edit distance = 6
a	s	t	r	o	n	o	m	i	c	a	l	l	y	

---

---

a	s	t	r	o	n	o	m	e	r					
			D				S	S	S	D				Edit distance = 5
a	s	t	_	o	n	i	s	h	_					

---

# Problem $\rightarrow$ subproblem $\rightarrow$ smaller subproblem structure

## Motivation:

For LIS, we looked at “prefixes” of the given array  $\Rightarrow$  for edit distance, we consider prefixes of  $s$  and  $t$ .

## Approach:

Let  $E(i, j)$  be the edit distance between  $s_1s_2s_3 \dots s_i$  and  $t_1t_2t_3 \dots t_j$  (we want  $E(m, n)$ ).

**Question:** How to express  $E(i, j)$  in terms of  $E()$  for smaller strings?

**Key idea:** Consider the last position in the alignment of  $s_{1\dots i}$  and  $t_{1\dots j}$ .

Case I	Case II	Case III
$s[i]$	—	$s[i]$
—	$t[j]$	$t[j]$
D	I	S / NOP

$$E(i, j) = \min\{E(i-1, j)+1, E(i, j-1)+1, E(i-1, j-1)+(s[i] \stackrel{?}{=} t[j])\}$$

## Initialisation

```
edit_info[0][0].distance = 0;
edit_info[0][0].operation = NOP;
for (i = 1; i < m+1; i++) {
    edit_info[i][0].distance = i;
    edit_info[i][0].operation = DELETE;
}
for (j = 1; j < n+1; j++) {
    edit_info[0][j].distance = j;
    edit_info[0][j].operation = INSERT;
}
```

## Updates

```
for (i = 1; i < m+1; i++) {
    for (j = 1; j < n+1; j++) {
        e = &edit_info[i][j];
        e->distance = edit_info[i-1][j].distance + 1;
        e->operation = DELETE;

        if (edit_info[i][j-1].distance + 1 < e->distance) {
            e->distance = edit_info[i][j-1].distance + 1;
            e->operation = INSERT;
        }

        same = (s[i-1] == t[j-1]) ? 1 : 0;
        d = edit_info[i-1][j-1].distance + (1-same);
        if (d < e->distance) {
            e->distance = d;
            e->operation = 1-same; // see note before #define NOP,
SUBST, etc.
        }
    }
}
```

## Computing the edits

```
if (NULL == (ops = Malloc(m+n, uint))) // I think this is actually
unnecessarily conservative
    ERR_MSG("levenstein: out of memory\n");
for (i = m, j = n, k = 0; i>=1 || j>=1; k++) {
    ops[k] = edit_info[i][j].operation;
    switch (ops[k]) {
    case INSERT:
        j--; break;           Case II in slide 12
    case DELETE:
        i--; break;         Case I in slide 12
    case NOP:
    case SUBST:
        i--, j--; break;    Case III in slide 12
    default:
        fprintf(stderr, "Unknown operation %u\n", ops[k]);
        exit(1);
    }
}
```

# Time and space complexity related issues

- Provided implementation:  $O(mn)$  space,  $O(mn)$  time
- Easy extension: compute edit distance only in  $O(m)$  space,  $O(mn)$  time
- Will not cover: edit distance **and** alignment in  $O(m)$  space,  $O(mn)$  time

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# Problem statement

**Problem:** Multiply  $A_{n_1 \times n_2}^{(1)} \times A_{n_2 \times n_3}^{(2)} \times \cdots \times A_{n_m \times n_{m+1}}^{(m)}$

**Example:**  $A_{10 \times 20} \times B_{20 \times 100} \times C_{100 \times 30}$

**Options:**

1.  $(A_{10 \times 20} \times B_{20 \times 100}) \times C_{100 \times 30}$
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No. of arithmetic operations

$$\approx 10 \times 20 \times 100 + 10 \times 100 \times 30 \\ = 50,000$$

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# Top-down approach

$$A_{n_1 \times n_2}^{(1)} \times A_{n_2 \times n_3}^{(2)} \times \dots \times A_{n_m \times n_{m+1}}^{(m)}$$

## Top down approach:

1. Break-up for the final / top-level multiplication

$$\left( A^{(1)} \times \dots \times A^{(k)} \right) \times \left( A^{(k+1)} \times \dots \times A^{(m)} \right)$$

$$k = 1, \dots, m - 1$$

2. Recursively solve left sub-chain and right sub-chain

Disadvantage: the same sub-problem is solved many times.

# Bottom-up approach

$$A_{n_1 \times n_2}^{(1)} \times A_{n_2 \times n_3}^{(2)} \times \dots \times A_{n_m \times n_{m+1}}^{(m)}$$

1. Only one way to compute  $A^{(i)} \times A^{(i+1)}$
2. Two ways to compute  $A^{(i)} \times A^{(i+1)} \times A^{(i+2)}$
3. Three ways to compute  $A^{(i)} \times A^{(i+1)} \times A^{(i+2)} \times A^{(i+3)}$ :
  - (a)  $A^{(i)} \times (A^{(i+1)} \times A^{(i+2)} \times A^{(i+3)})$
  - (b)  $(A^{(i)} \times A^{(i+1)}) \times (A^{(i+2)} \times A^{(i+3)})$
  - (c)  $(A^{(i)} \times A^{(i+1)} \times A^{(i+2)}) \times A^{(i+3)}$

# Bottom-up approach

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(b)  $(A^{(i)} \times A^{(i+1)}) \times (A^{(i+2)} \times A^{(i+3)})$

(c)  $(A^{(i)} \times A^{(i+1)} \times A^{(i+2)}) \times A^{(i+3)}$

*We already  
know the best  
way to do this.*



# Algorithm

$$Cost[i, j] = \begin{cases} 0 & \text{if } j = 1 \\ \min_{1 \leq k \leq j-1} \{ Cost[i, k] + Cost[i+k, j-k] + n_i n_{i+k} n_{i+j} \} & \text{if } j > 1 \end{cases}$$

position

length of chain

J	1	2	3	4	...	m
1	0	$n_1.n_2.n_3$				
2	0	$n_2.n_3.n_4$				
3	0	$n_3.n_4.n_5$				
4	0	$n_4.n_5.n_6$				

This is the value we want.

Fill in the table in increasing order of  $j$ .

Store cost and position in table.

m-1	0	← $n_{m-1}.n_m.n_{m+1}$
m	0	

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# Definition

**Input:** a tree  $T(V, E)$

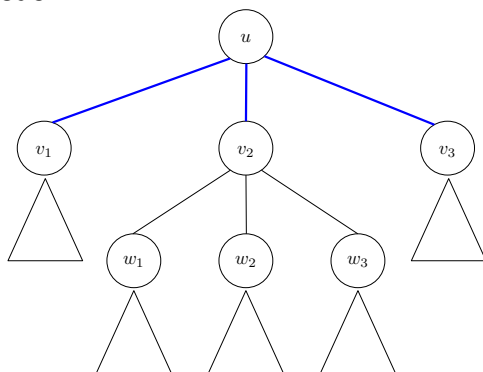
**Independent set:**  $S \subset V$  s.t.  $\forall u, v \in S, (u, v) \notin E$

**Required output:** an independent set that has the largest cardinality among all independent sets of  $T$

Problem  $\rightarrow$  subproblem  $\rightarrow$  smaller subproblem structure

**Preprocessing:** convert the tree to a *rooted tree* (how?)

**Bottom-up approach:**



$$I(u) = \max\left\{1 + \sum_{w \in \text{grandchildren}(u)} I(w), \sum_{v \in \text{children}(u)} I(v)\right\}$$

# Exercises

1. Implement the matrix chain multiplication algorithm.

**Input format:** Note that a chain containing  $N$  matrices can be represented as  $N + 1$  positive integers. Take these numbers as command-line arguments.

**Output format:** Print the smallest number of multiplication operations that are needed to compute the product, as well as the bracketing (how?).

2. *Longest common subsequence:* Given two strings,  $s = s_1s_2s_3 \dots s_m$  and  $t = t_1t_2t_3 \dots t_n$ , find a subsequence  $u = u_1u_2 \dots u_l$  of maximum length that occurs in both  $s$  and  $t$ .

# References

- Chapter 6 of *Algorithms* by Dasgupta, Papadimitriou, Vazirani.  
<https://people.eecs.berkeley.edu/~vazirani/algorithms/chap6.pdf>
- Chapter 15 of *Introduction to Algorithms* (2nd ed.) by Cormen, Leiserson, Rivest, Stein.