

NPTEL MOOC, JAN-FEB 2015
Week 7, Module 5

DESIGN AND ANALYSIS OF ALGORITHMS

Edit Distance

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Document similarity

- * “The students were able to appreciate the concept optimal substructure property and its use in designing algorithms”
- * “The lecture taught the students to appreciate how the concept of optimal substructures can be used in designing algorithms”
- * “The lecture taught the students ~~were able~~ to appreciate how the concept of optimal substructures ~~property~~ can ~~d~~ itbse used in designing algorithms”
- * 28 characters inserted, 18 ~~deleted~~, 2 substituted

Edit distance

- * Minimum number of editing operations needed to transform one document to the other
 - * Insert a character
 - * Delete a character
 - * Substitute a character by another one
- * In our example,
28 characters inserted, 18 ~~deleted~~, 2 ~~substituted~~
- * Edit distance is at most 48

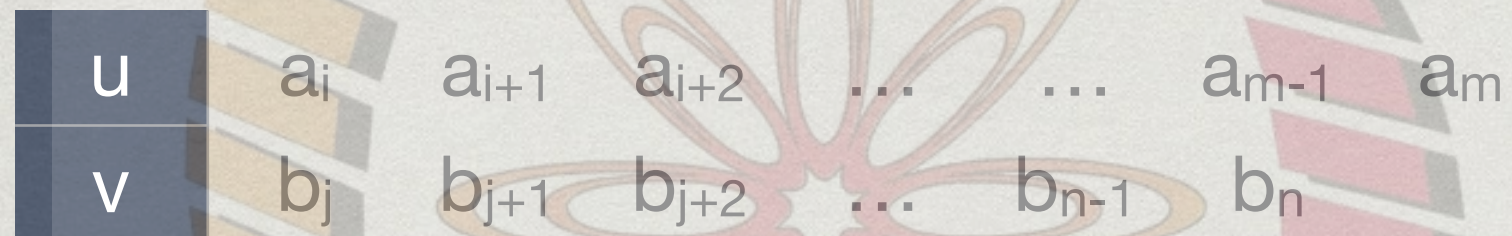
Edit distance

- * Also called Levenshtein distance
 - * First proposed by Vladimir Levenshtein
- * Applications
 - * Suggest spelling corrections in word processor, search engine queries
 - * Another way of comparing genetic similarity across species

Edit distance and LCS

- * Longest common subsequence of u and v
 - * What remains after minimum number of deletes to make them equal
- * Deleting a letter in u equivalent to inserting it in v
 - * “secret”, “bisect” — LCS is “sect”
 - * delete “r”, “e” in “secret”, “b”, “i” in “bisect”
 - * delete “r”, “e” then insert “b”, “i” in “secret”
- * LCS is equivalent to edit distance without substitution

Inductive structure for edit distance



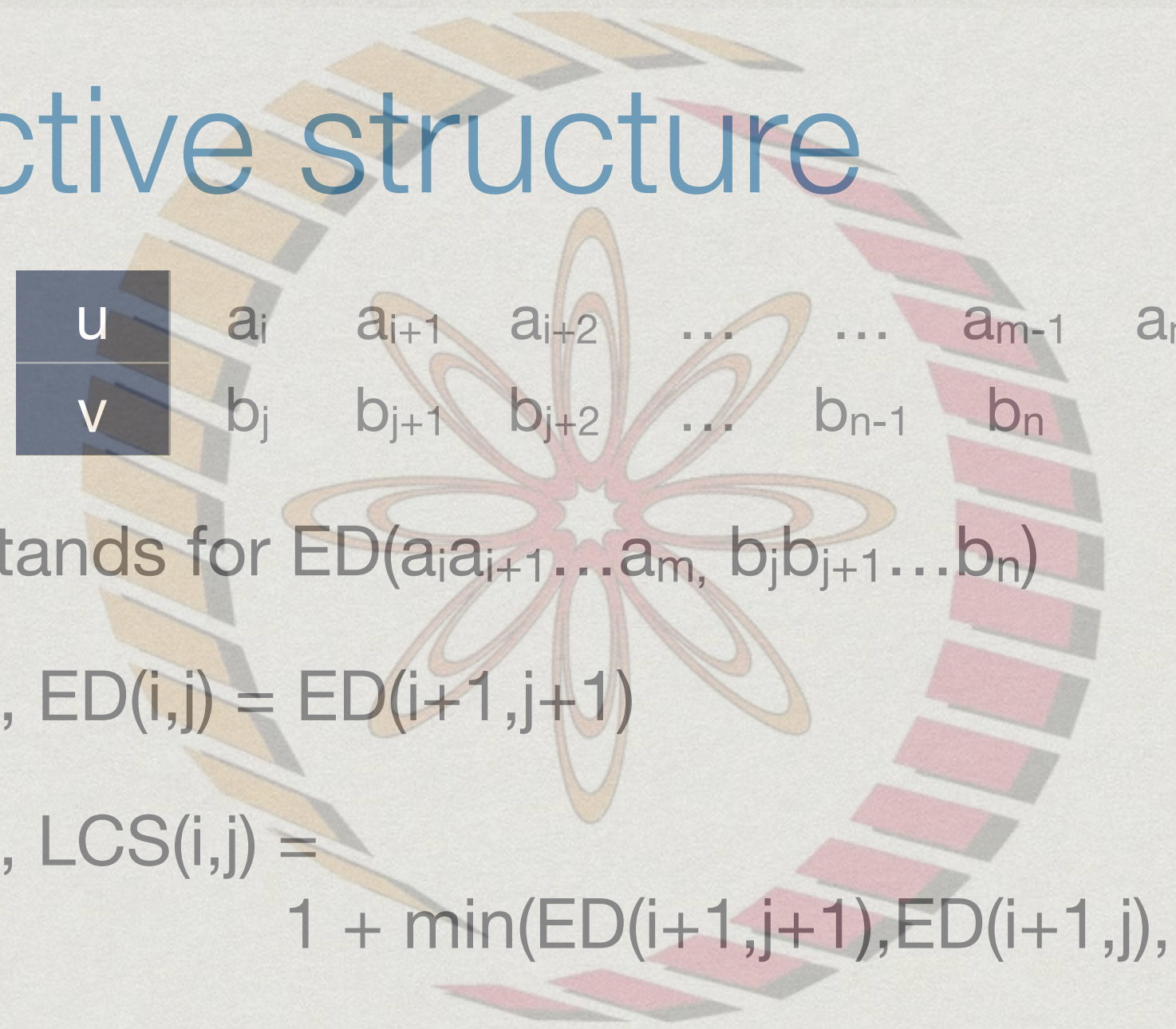
- * Recall LCS
 - * If $a_i = b_j$, $\text{LCS}(i,j) = 1 + \text{LCS}(i+1,j+1)$
 - * If $a_i \neq b_j$, $\text{LCS}(i,j) = \max(\text{LCS}(i+1,j), \text{LCS}(i,j+1))$
 - * Boundary condition when one of the words is empty

Edit distance...

u	a_i	a_{i+1}	a_{i+2}	a_{m-1}	a_m
v	b_j	b_{j+1}	b_{j+2}	...	b_{n-1}	b_n	

- * Aim is to transform u into v
 - * If $a_i = b_j$, $ED(i,j) = ED(i+1,j+1)$ — nothing to be done at (a_i, b_j)
 - * If $a_i \neq b_j$, can do one of three things
 - * Substitute a_i by b_j : $1 + ED(i+1,j+1)$
 - * Delete a_i : $1 + ED(i+1,j)$
 - * Insert b_j before a_i : $1 + ED(i,j+1)$
 - * Take the **minimum** of these

Inductive structure



u	a_i	a_{i+1}	a_{i+2}	\dots	\dots	a_{m-1}	a_m
v	b_j	b_{j+1}	b_{j+2}	\dots	b_{n-1}	b_n	

- * $ED(i,j)$ stands for $ED(a_i a_{i+1} \dots a_m, b_j b_{j+1} \dots b_n)$
- * If $a_i = b_j$, $ED(i,j) = ED(i+1,j+1)$
- * If $a_i \neq b_j$, $LCS(i,j) =$
 $1 + \min(ED(i+1,j+1), ED(i+1,j), ED(i,j+1))$
- * As with LCS/LCW, extend positions to $m+1, n+1$
 - * $ED(m+1,j) = n-j+1$ for all j # Insert $b_j b_{j+1} \dots b_n$ in u
 - * $ED(i,n+1) = m-i+1$ for all i , # Insert $a_i a_{i+1} \dots a_m$ in v

Subproblem dependency

- * Like LCS, $ED(i,j)$ depends on $ED(i+1,j+1)$, $ED(i+1,j)$ and $ED(i,j+1)$
- * Dependencies for $ED(m,n)$ are known
- * Start at $ED(m,n)$ and fill by row, column or diagonal

		0	1	2	3	4	5	6
		s	e	c	r	e	t	.
0	b							
1	i							
2	s							
3	e							
4	c							
5	t							
6	.							

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		0	1	2	3	4	5	6
		s	e	c	r	e	t	.
0	b							6
1	i							5
2	s							4
3	e							3
4	c							2
5	t							1
6	.	6	5	4	3	2	1	0

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		0	1	2	3	4	5	6
		s	e	c	r	e	t	.
0	b						5	6
1	i						4	5
2	s						3	4
3	e						2	3
4	c						1	2
5	t						0	1
6	.	6	5	4	3	2	1	0

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		0	1	2	3	4	5	6
		s	e	c	r	e	t	.
0	b					5	5	6
1	i					4	4	5
2	s					3	3	4
3	e					2	2	3
4	c					1	1	2
5	t					1	0	1
6	.	6	5	4	3	2	1	0

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		0	1	2	3	4	5	6	
		s	e	c	r	e	t	•	
0	b					5	5	5	6
1	i					4	4	4	5
2	s					3	3	3	4
3	e					2	2	2	3
4	c					2	1	1	2
5	t					2	1	0	1
6	•	6	5	4	3	2	1	0	

Subproblem dependency

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		0	1	2	3	4	5	6
		s	e	c	r	e	t	.
0	b			5	5	5	5	6
1	i			4	4	4	4	5
2	s			3	3	3	3	4
3	e			3	2	2	2	3
4	c			2	2	1	1	2
5	t			3	2	1	0	1
6	.	6	5	4	3	2	1	0

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		0	1	2	3	4	5	6
		s	e	c	r	e	t	.
0	b		5	5	5	5	5	6
1	i		4	4	4	4	4	5
2	s		3	3	3	3	3	4
3	e		2	3	2	2	2	3
4	c		3	2	2	1	1	2
5	t		4	3	2	1	0	1
6	.	6	5	4	3	2	1	0

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		0	1	2	3	4	5	6
		s	e	c	r	e	t	.
0	b	4	5	5	5	5	5	6
1	i	3	4	4	4	4	4	5
2	s	2	3	3	3	3	3	4
3	e	3	2	3	2	2	2	3
4	c	4	3	2	2	1	1	2
5	t	5	4	3	2	1	0	1
6	.	6	5	4	3	2	1	0

Recovering the solution

- * Trace back the path
- * Transforming “secret” to “bisect”
- * Del “b” : $(0,0) \rightarrow (1,0)$
- * Del “i” : $(1,0) \rightarrow (2,0)$
- * Ins “r” : $(5,3) \rightarrow (5,4)$
- * Ins “e”: $(5,4) \rightarrow (5,5)$

		0	1	2	3	4	5	6
		s	e	c	r	e	t	•
0	b	4	5	5	5	5	5	6
1	i	3	4	4	4	4	4	5
2	s	2	3	3	3	3	3	4
3	e	3	2	3	2	2	2	3
4	c	4	3	2	2	1	1	2
5	t	5	4	3	2	1	0	1
6	•	6	5	4	3	2	1	0

ED(u,v), DP

```
function ED(u,v) # u[0..m], v[0..n]
for r = 0,1,...,m+1 { ED[r][n+1] = m-r+1 }
for c = 0,1,...,m+1 { ED[m+1][c] = n-c+1 }
for c = n,n-1,...,0
  for r = m,m-1,...,0
    if (u[r] == v[c])
      ED[r][c] = ED[r+1][c+1]
    else
      ED[r][c] = 1 + min(ED[r+1][c+1],
                        ED[r+1][c],
                        ED[r][c+1])
return(ED[0][0])
```


Complexity

- * Again $O(mn)$ using dynamic programming (or memoization)
- * Need to fill an $O(mn)$ size table
- * Each table entry takes constant time to compute

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Space complexity

- * For LCW, LCS, ED
 - * Need to fill an $O(mn)$ size table
 - * Do we need to store the entire table?
- * Filling column by column, only need next column and current column
 - * Or next row and current row
- * Reduce space to $O(n)$, assuming $m \geq n$