12.7. The Four-Russians speedup

2016.04.26 조현진

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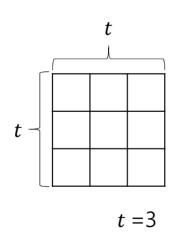
Definitions

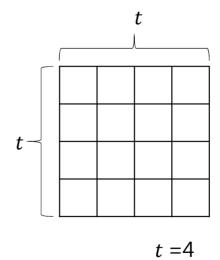
- S₁ : String 1
- S₂ : String 2
- *n* : both length of S₁ and S₂
- i: character position of S_1 ($0 \le i \le n$)
- j: character position of S_2 ($0 \le j \le n$)
- $S_1(i) : i'$ th character in S_1
- $S_2(j)$: j'th character in S_2
- σ : aphabet size $(Ex. \ \sigma\{A,C,G,T\}=4)$
- D(i, j): Edit Distance at (i, j)

t -block

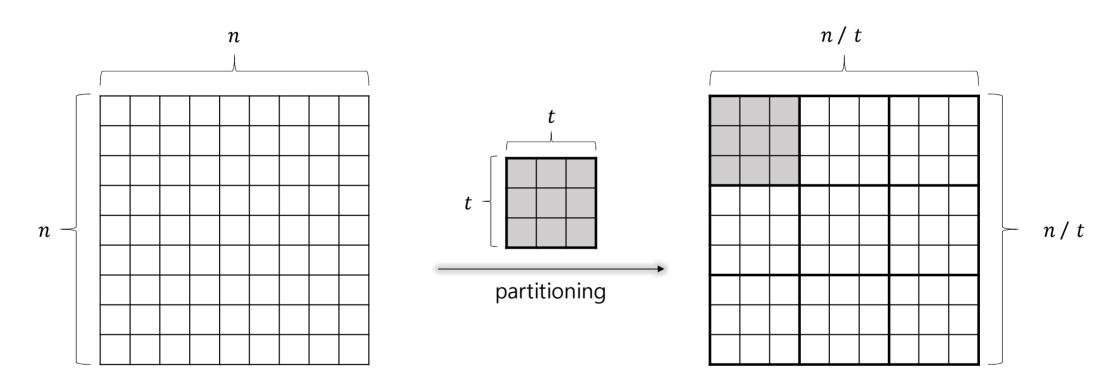
Definition

A t-block is a t by t square in the dynamic programming table.

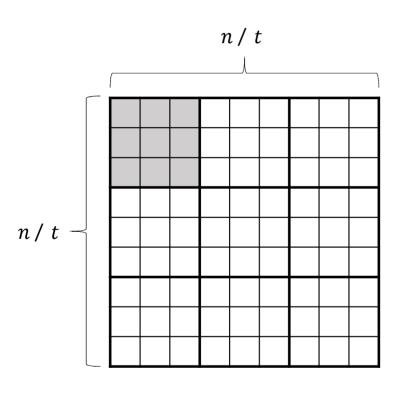




The Four-Russians speedup concept



The Four-Russians speedup concept



<Goal>

- O(t) time per t-block
- Time Complexity $O(n^2) => O(\frac{n^2}{\log n})$

Block function

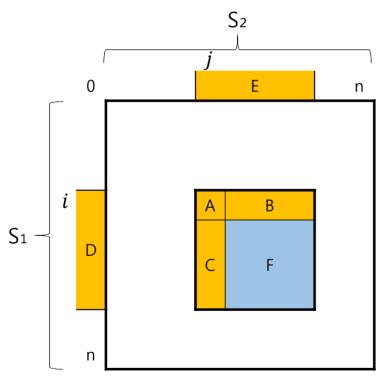
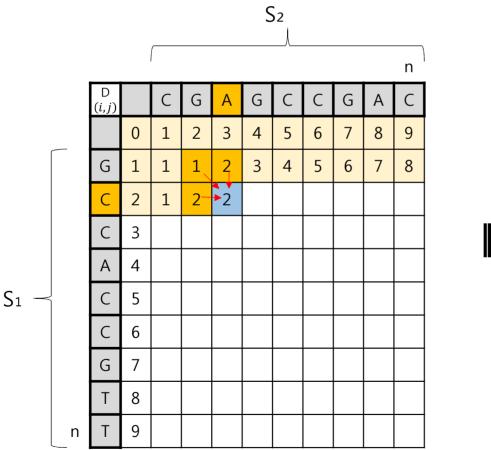


Figure 12.21

Lemma 12.7.1.

the distance values in a t-block starting in position (i,j) are a function of the values in its first row and column and the substrings $S_1[i,i+1,...,i+t-1]$ and $S_2[j,j+1,...,j+t-1]$.

Block function



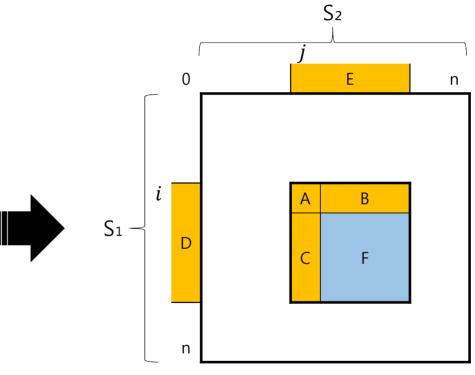


Figure 12.21

Block function

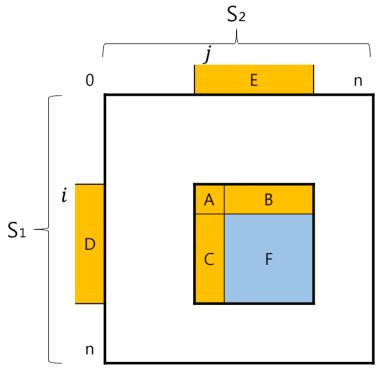


Figure 12.21

Definition

Given lemma 12.7.1, and using the notation shown in Figure 12.21, we define the block function as the function from the five inputs (A, B, C, D, E) to the output F.

{F} = blockFunction(A, B, C, D, E)

Restricted Block Function (RBF)

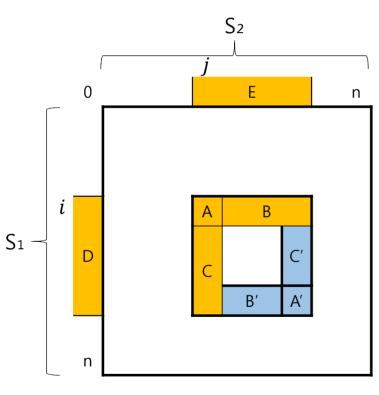
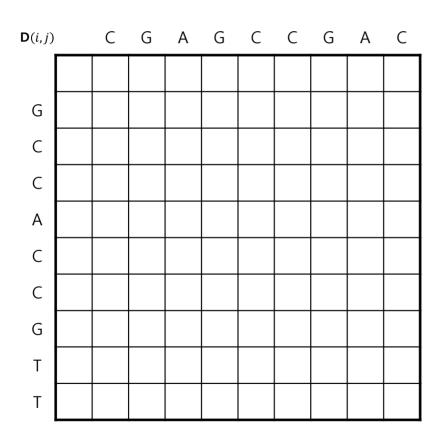


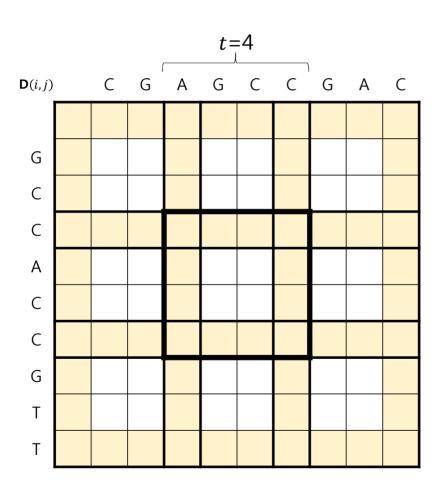
Figure 12.21

the values in the last row and column of a t-block are also a function of the inputs (A, B, C, D, E).

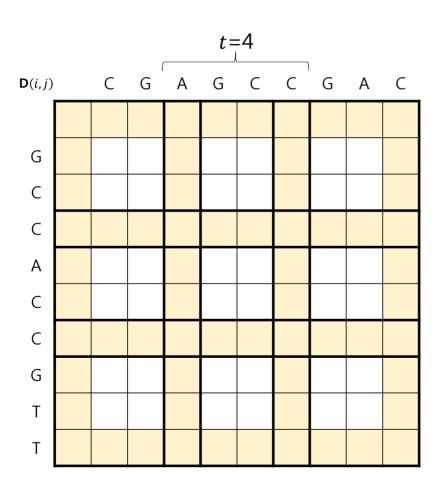
$$\{A', B', C'\} = RBF(A, B, C, D, E)$$



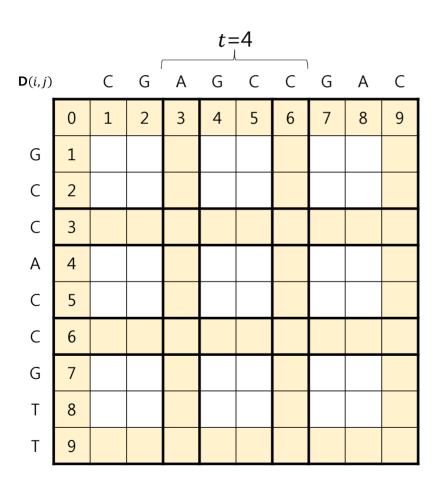
① Cover the (n + 1) by (n + 1) dynamic programming table with t-blocks that overlap by 1 row & 1 column



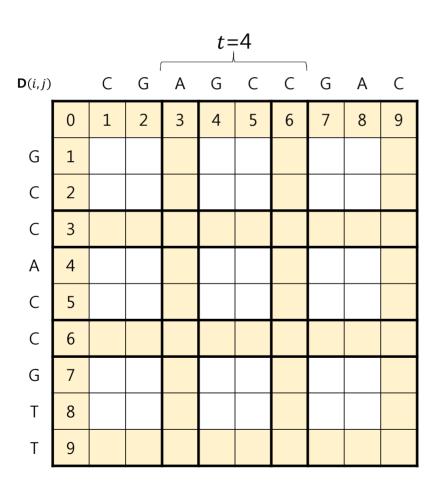
① Cover the (n + 1) by (n + 1) dynamic programming table with t-blocks that overlap by 1 row & 1 column

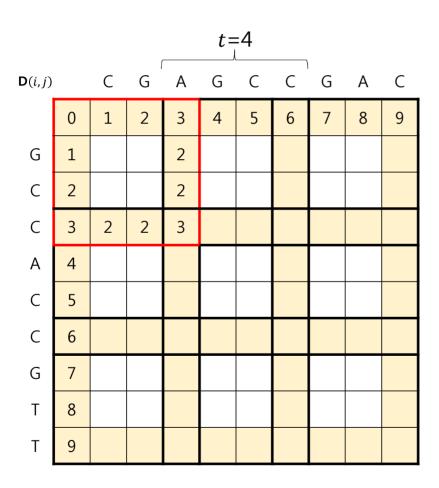


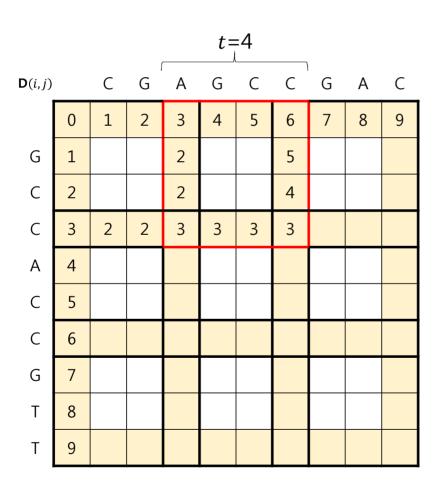
② Initialize the values in the first row and column of the full table according to the base conditions of the recurrence.

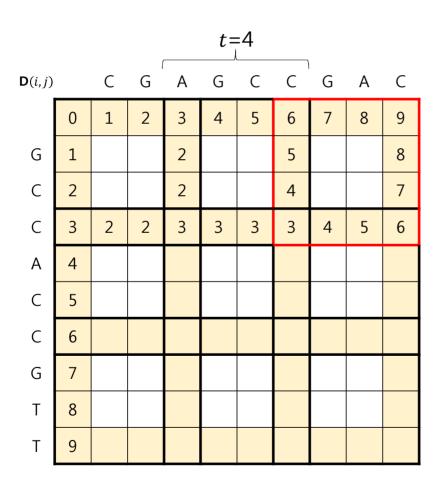


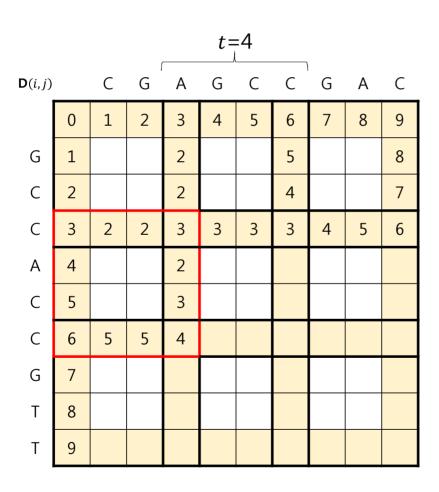
② Initialize the values in the first row and column of the full table according to the base conditions of the recurrence.



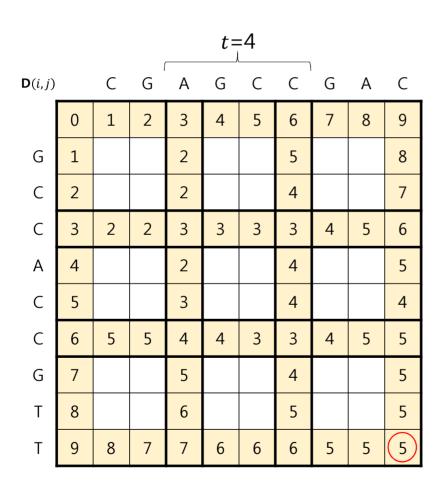








	t=4									
$\mathbf{D}(i,j)$		С	G	Α	G	С	C	G	Α	С
	0	1	2	3	4	5	6	7	8	9
G	1			2			5			8
С	2			2			4			7
С	3	2	2	3	3	3	3	4	5	6
Α	4			2			4			5
С	5			3			4			4
С	6	5	5	4	4	3	3	4	5	5
G	7			5			4			5
Т	8			6			5			5
Т	9	8	7	7	6	6	6	5	5	5



4 the value in cell (n, n) is the edit distance of S_1 and S_2

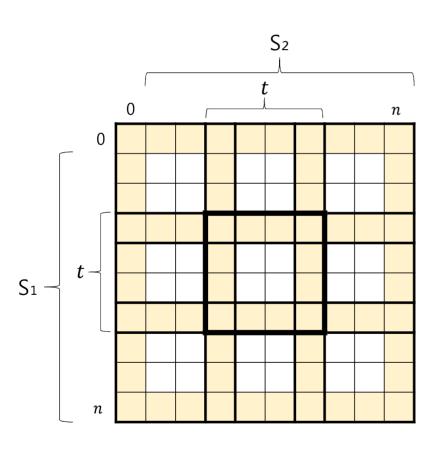
any instance of the restricted block function can be computed $O(t^2)$ time, but that gains us nothing-!!

The Four-Russians idea for the RBF

• precomputation and storing subproblems.

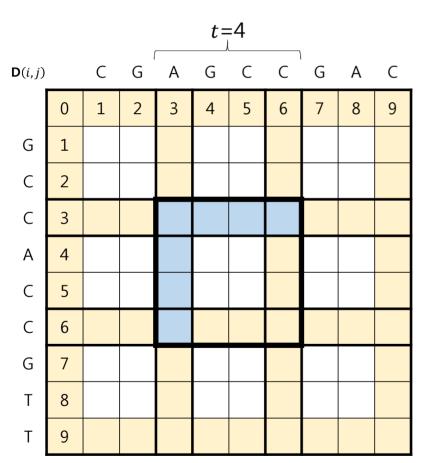
Is it any faster than the original $O(n^2)$ method?

The Four-Russians idea for the RBF

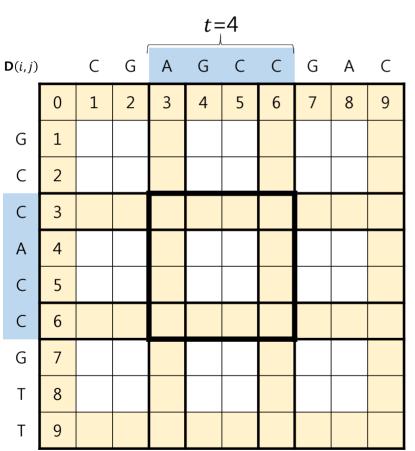


- assume that precomputation is done
- the size of input and the output of the RBF are both O(t)
- thus, output can be retrieved in O(t)
- there are $\theta(\frac{n^2}{t^2})$ blocks
- total time = $0(\frac{n^2}{t})$
- set $t = \log n$, then $O(\frac{n^2}{\log n})$

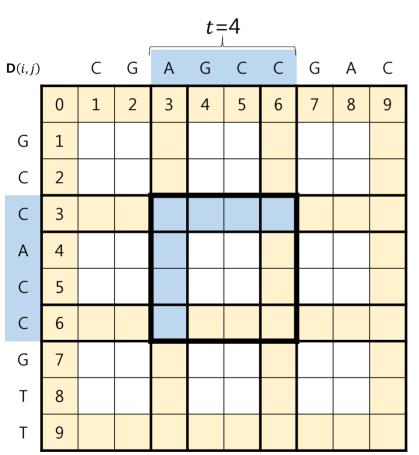
What about precomputation time?



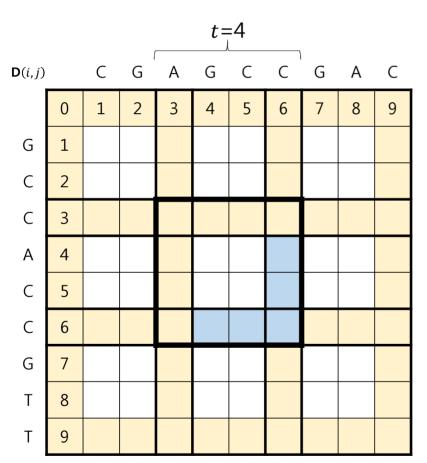
• Every cell has $0 \sim n$ values, so there are $(n+1)^t$ possible values for any t-length row or column



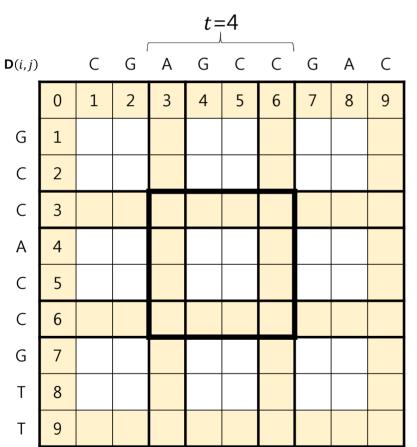
- Every cell has $0 \sim n$ values, so there are $(n + 1)^t$ possible values for any t-length row or column
- If alphabet has size is σ , then there are σ^t possible substrings of length t



- Every cell has $0 \sim n$ values, so there are $(n+1)^t$ possible values for any t-length row or column
- If alphabet has size is σ , then there are σ^t possible substrings of length t
- Therefore, there are $(n + 1)^{2t}\sigma^{2t}$ distinct input combinations for RBF



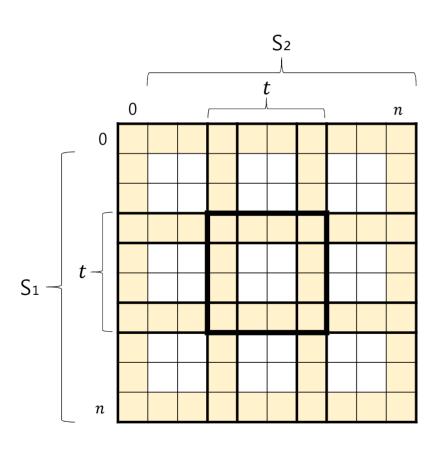
• For each input, it takes $\theta(t^2)$ time to evaluate the last row and column of the resulting t-block (by running standard dynamic programming)



- For each input, it takes $\theta(t^2)$ time to evaluate the last row and column of the resulting t-block (by running standard dynamic programming)
- Thus, total precomputation time is $\theta((n+1)^{2t}\sigma^{2t} t^2)$
- t must be at least 1, so $\Omega(n^2)$ is used

we need another trick-!!

The Four-Russians idea for the RBF



- assume that precomputation is done
- the size of input and the output of the RBF are both O(t)
- thus, output can be retrieved in O(t)
- there are $\theta(\frac{n^2}{t^2})$ blocks
- total time = $O(\frac{n^2}{t})$
- set $t = \log n$, then $O(\frac{n^2}{\log n})$

Lemma 12.7.2.

In any row, column, or diagonal of the dynamic programming table for edit distance, two adjacent cells can have a value that differs by at most one.

Example)	$\mathbf{D}(i,j)$		C	G	Α	G	С	С	G	Α	С
		0	1	2	3	4	5	6	7	8	9
	G	1	1	1	2	3	4	5	6	7	8
	С	2	1	2	2	3	3	4	5	6	7
	С	3	2	2	3	3	3	3	4	5	6
	Α	4	3	3	2	3	4	4	4	4	5
	С	5	4	4	3	3	3	4	5	5	4
	С	6	5	5	4	4	3	3	4	5	5
	G	7	6	5	5	4	4	4	3	4	5
	Т	8	7	6	6	5	5	5	4	4	5
	Т	9	8	7	7	6	6	6	5	5	5

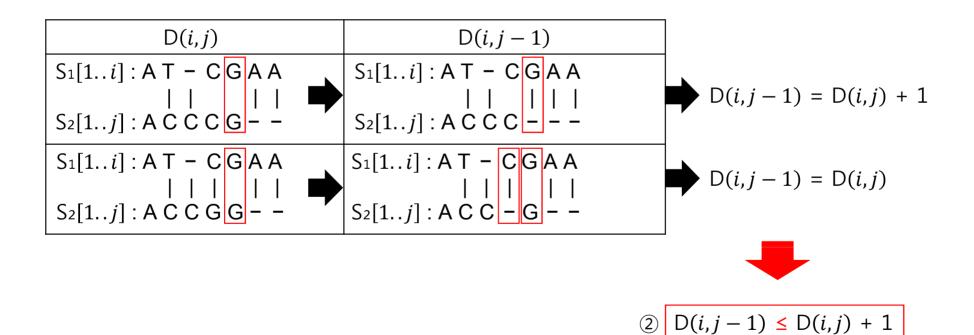
Proof

• Certainly, $D(i,j) \leq D(i,j-1) + 1$... ①

$$D(i,j) = mi \begin{cases} D(i-1,j) + 1 \\ D(i,j-1) + 1 \\ D(i-1,j-1) + t(i,j) \end{cases}$$

Proof

Case 1. $S_2(j)$ is matched with some character of $S_1[1..i]$



Proof



Case 2. $S_2(j)$ is not matched with $S_1[1..i]$

D(i,j)	D(i, j-1)
$S_1[1i] : AT - CGAA$ $ $ $S_2[1j] : ACCCT$	$S_1[1i] : AT - CGAA$ $S_2[1j] : ACCCC$
$S_1[1i] : A T T C G A$	$S_1[1i] : A T T C G A - $

$$\mathsf{D}(i,j-1) = \mathsf{D}(i,j)$$

$$\mathsf{D}(i,j-1) = \mathsf{D}(i,j) - 1$$



③
$$D(i,j-1) ≥ D(i,j) - 1$$

①
$$D(i,j) \le D(i,j-1) + 1$$
 $D(i,j) \le D(i,j-1) + 1$



$$D(i, j - 1) - 1 \le D(i, j) \le D(i, j - 1) + 1$$

Proof

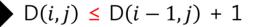


is same as



$$D(i,j) \leq D(i-1,j) + 1$$
 $D(i,j) \leq D(i-1,j) + 1$

③
$$D(i-1,j)$$
 ≥ $D(i,j)$ - 1





$$D(i-1,j) - 1 \le D(i,j) \le D(i-1,j) + 1$$



• Certainly, $D(i,j) \leq D(i-1,j-1) + 1$... ①

$$D(i,j) = min \begin{cases} D(i-1,j) + 1 \\ D(i,j-1) + 1 \\ D(i-1,j-1) + t(i,j) \end{cases}$$

Proof



Case 1. optimal alignment of $S_1[1..i]$ and $S_2[1..j]$ aligns i against j

$S_1[1i] : AT - CGA$ $S_1[1i] : AT - CG$	
$ $ $S_2[1j] : A C C C G A$ $ $ $S_2[1j] : A C C C G$	D(i-1,j-1) = D(i,j)
$S_1[1i] : AT - CGA$ $ $	D(i-1,j-1) = D(i,j) - 1



Proof



Case 2. optimal alignment doesn't align *i* against *j*



Proof



②
$$D(i-1,j-1) \leq D(i,j)$$

$$(3) D(i-1,j-1) \leq D(i,j)$$



$$D(i-1,j-1) \le D(i,j) \le D(i-1,j-1) + 1$$

Lemma 12.7.2. has been proven-!!

Definition

The *offset vector* is a t-length vector of values from $\{-1, 0, 1\}$, where the first entry must be zero.

offset vector: 0 -1 0 1

5	4	4	5	



0	-1	0	1

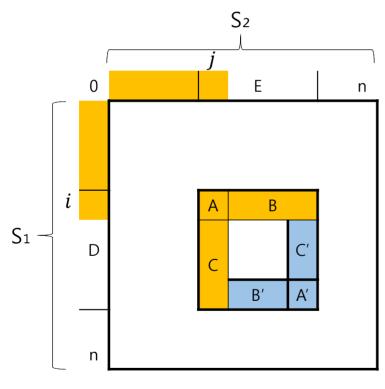
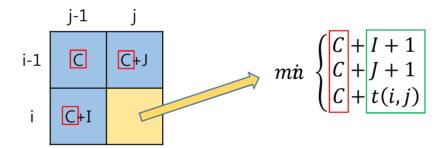


Figure 12.21

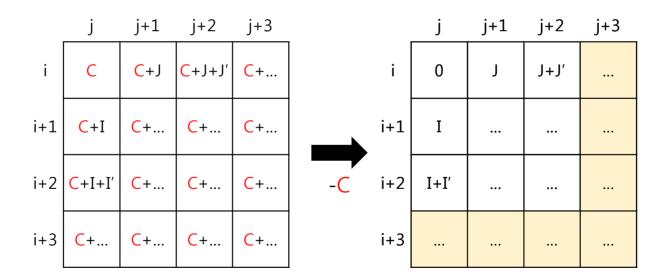
Theorem 12.7.1.

Consider a t-block with upper left corner in position (i,j). The two offset vectors for the last row and last column of the block can be determined from the two offset vectors for the first row and column of the block and from substrings $S_1[1..i]$ and $S_2[1..j]$. That is, no Edit Distance value is needed in the input in order to determine the offset vectors in the last row and column of the block.

Proof



Proof



Theorem 12.7.1. has been proven-!!

offset function

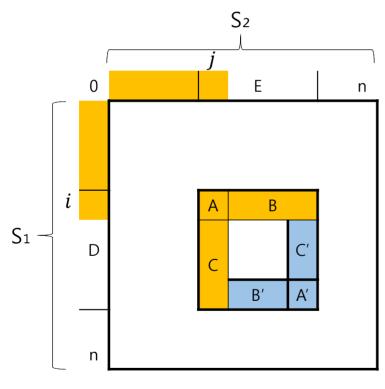
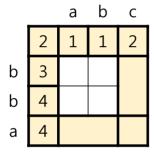
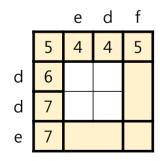


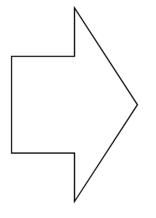
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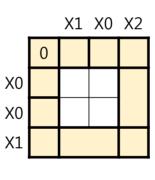
Definition

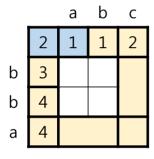
The function that determines the two offset vectors for the last row and last column from the two offset vectors for the first row and column of a block together with substrings $S_1[1..i]$ and $S_2[1..j]$ is called the *offset function*.

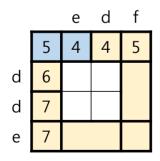


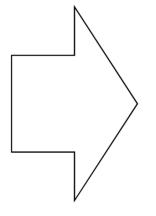


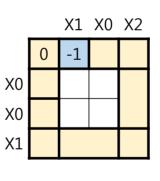


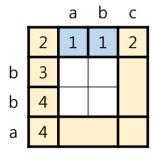


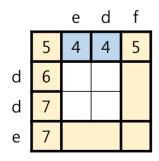


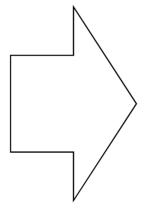




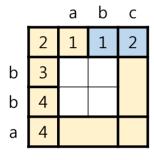


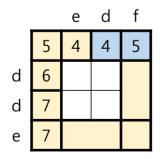


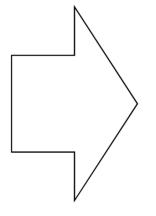


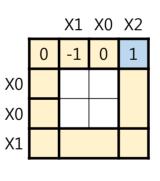


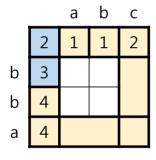
		X1	X0	X2
	0	-1	0	
X0				
X0				
X1				

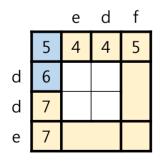


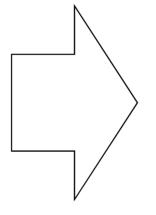


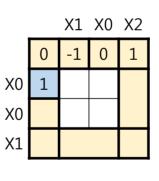


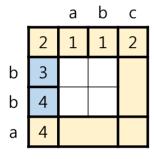


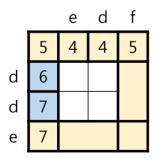


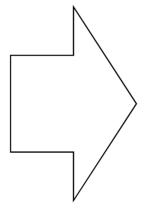




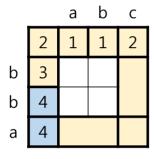


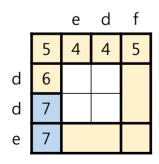


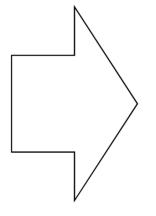




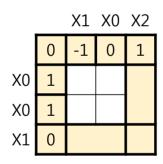
_		X1	X0	X2
	0	-1	0	1
(0	1			
(0	1			
(1				







		X1	X0	X2
	0	-1	0	1
X0	1			
X0	1			
X1	0			



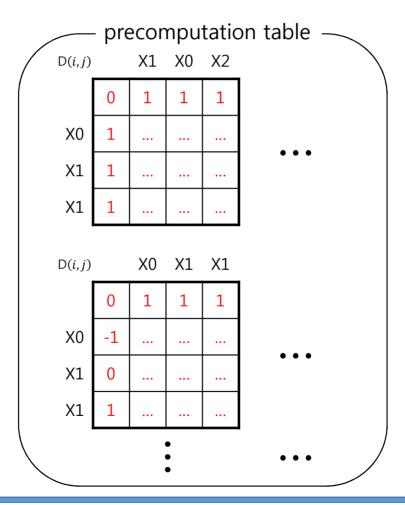
$$\theta((n+1)^{2t}\sigma^{2t} \ t^2)$$

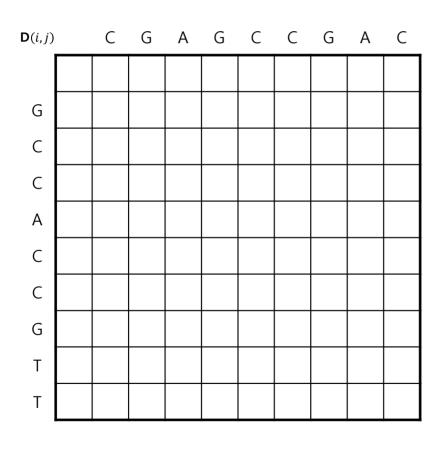
$$\theta((3)^{2t}\sigma^{2t} \ t^2)$$

$$t = \frac{\log_{3\sigma} n}{2}$$

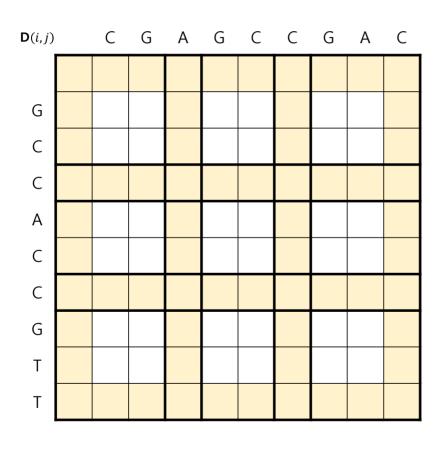
$$\theta\left((3\sigma)^{2\left(\frac{\log_{3\sigma} n}{2}\right)\left(\frac{\log_{3\sigma} n}{2}\right)^2\right) \longrightarrow O(n(\log n)^2)$$

Precomputation table

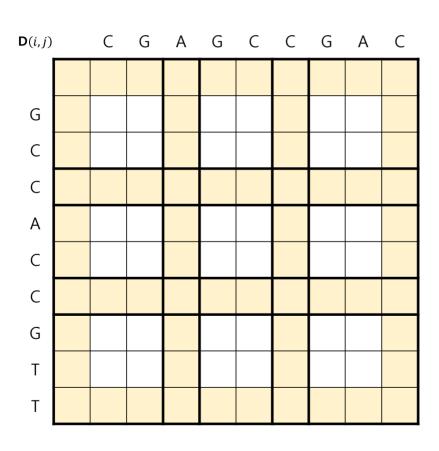




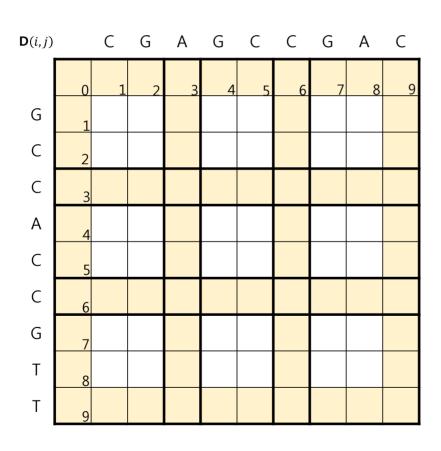
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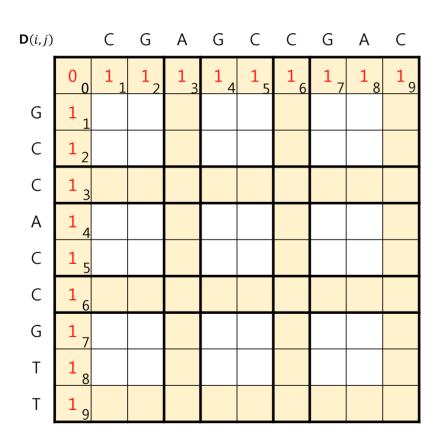
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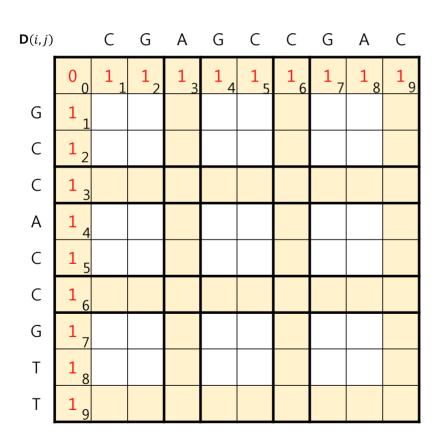
② Initialize the values in the first row and column of the full table according to the base conditions of the recurrence. Compute the offset values in the first row and column.

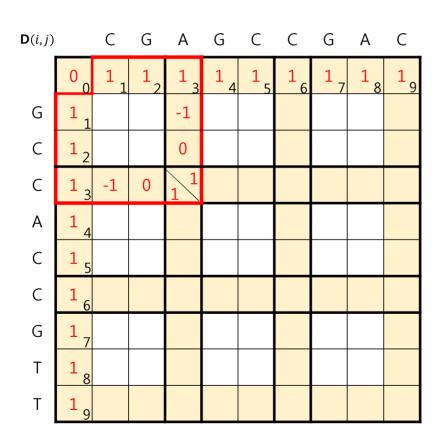


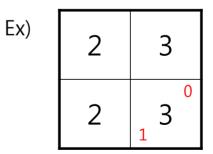
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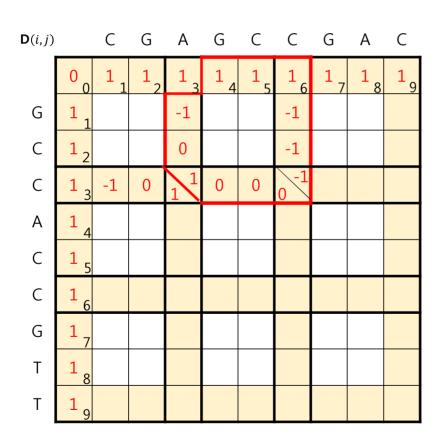


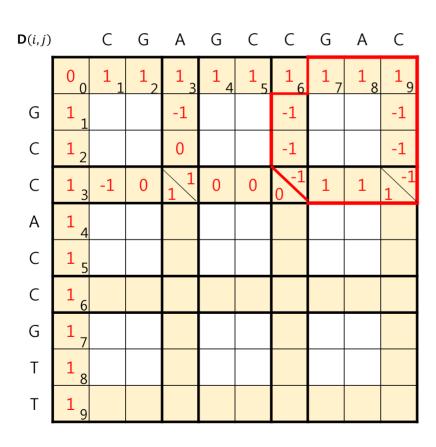
2 Initialize the values in the first row and column of the full table according to the base conditions of the recurrence. Compute the offset values in the first row and column.

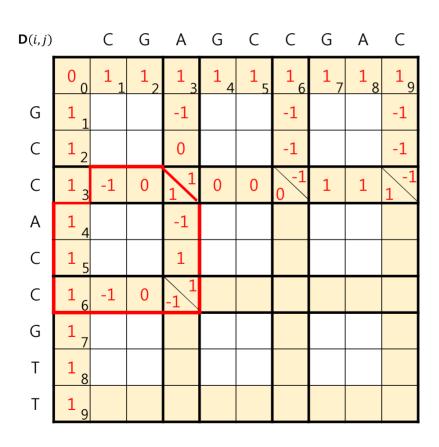


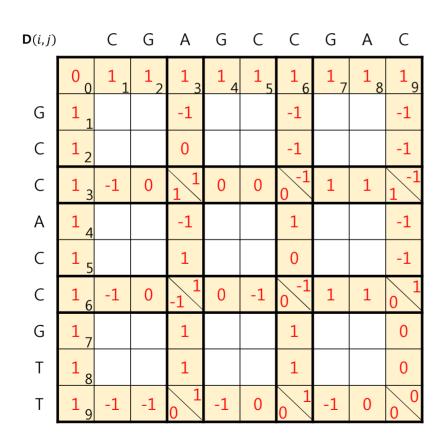


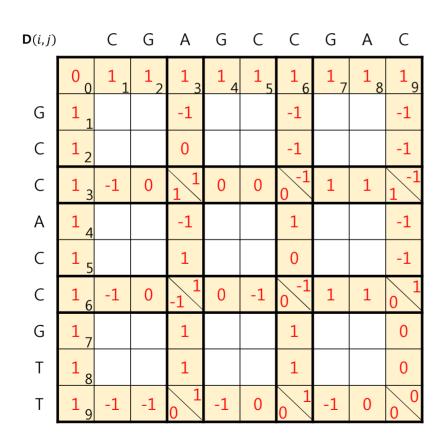




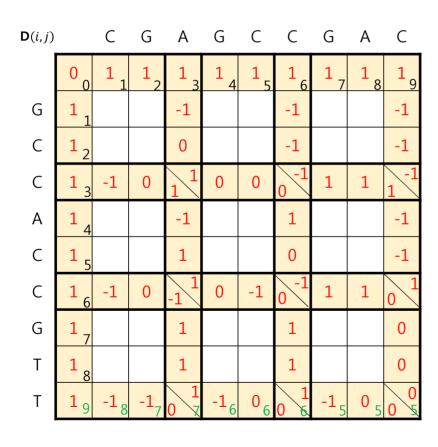






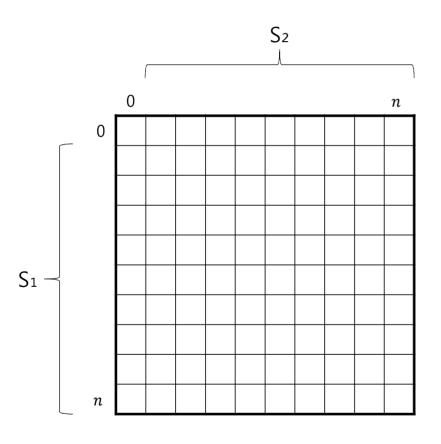


4 The value in cell (n, n) is the edit distanace of S_1 and S_2 . Let Q be the total of the offset values computed for cells in row n. D(n, n) = D(n, 0) + Q = n + Q



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Summary



<Goal>

- O(t) time per t-block

- Time Complexity
$$O(n^2) => O(\frac{n^2}{\log n})$$