

CSE 421 Algorithms

Sequence Alignment

Sequence Alignment

What

Why

A Dynamic Programming Algorithm

Sequence Similarity: What

G G A C C A

T A C T A A G

T C C A A G

Sequence Similarity: What

G G A C C A

T A C T A A G

| | | | |

T C C - A A G

Sequence Similarity: Why

Bio

Most widely used comp. tools in biology

New sequence always compared to data bases

Similar sequences often have similar origin or function

Recognizable similarity after $10^8 - 10^9$ yr

DNA sequencing & assembly

Other

spell check/correct, diff, svn/git/..., plagiarism, ...

Terminology

String: ordered list of letters TATAAG

Prefix: consecutive letters from front
empty, T, TA, TAT, ...

Suffix: ... from end
empty, G, AG, AAG, ...

Substring: ... from ends or middle
empty, TAT, AA, ...

Subsequence: ordered, nonconsecutive
TT, AAA, TAG, ...

Sequence Alignment

a c b c d b
 / \
c a d b d

a c - - b c d b
 | | |
- c a d b - d -

Defn: An *alignment* of strings S , T is a pair of strings S' , T' (with dashes) s.t.

(1) $|S'| = |T'|$, and $(|S| = \text{“length of } S\text{”})$

(2) removing all dashes leaves S , T

Alignment Scoring

Mismatch = -1
Match = 2

a c b c d b
c a d b d

a c - - b c d b
- c a d b - d -
-1 2 -1 -1 2 -1 2 -1

Value = 3*2 + 5*(-1) = +1

The *score* of aligning (characters or dashes) x & y is $\sigma(x,y)$.

Value of an alignment $\sum_{i=1}^{|S'|} \sigma(S'[i], T'[i])$

An *optimal alignment*: one of max value
(Assume $\sigma(-,-) < 0$)

Alignment by Dynamic Programming?

Common Subproblems?

Plausible: probably re-considering alignments of various small substrings unless we're careful.

Optimal Substructure?

Plausible: left and right "halves" of an optimal alignment probably should be optimally aligned (though they obviously interact a bit at the interface).

(Both made rigorous below.)

Optimal Substructure (In More Detail)

Optimal alignment *ends* in 1 of 3 ways:

last chars of S & T aligned with each other

last char of S aligned with dash in T

last char of T aligned with dash in S

(never align dash with dash; $\sigma(-, -) < 0$)

In each case, the *rest* of S & T should be *optimally* aligned to each other

Optimal Alignment in $O(n^2)$ via “Dynamic Programming”

Input: $S, T, |S| = n, |T| = m$

Output: **value** of optimal alignment

Easier to solve a “harder” problem:

$V(i,j)$ = value of optimal alignment of
 $S[1], \dots, S[i]$ with $T[1], \dots, T[j]$
for **all** $0 \leq i \leq n, 0 \leq j \leq m$.

Base Cases

$V(i,0)$: first i chars of S all match dashes

$$V(i,0) = \sum_{k=1}^i \sigma(S[k], -)$$

$V(0,j)$: first j chars of T all match dashes

$$V(0,j) = \sum_{k=1}^j \sigma(-, T[k])$$

General Case

Opt align of $S[1], \dots, S[i]$ vs $T[1], \dots, T[j]$:

$$\left[\begin{array}{c} \sim\sim\sim\sim S[i] \\ \sim\sim\sim\sim T[j] \end{array} \right], \left[\begin{array}{c} \sim\sim\sim\sim S[i] \\ \sim\sim\sim\sim - \end{array} \right], \text{ or } \left[\begin{array}{c} \sim\sim\sim\sim - \\ \sim\sim\sim\sim T[j] \end{array} \right]$$

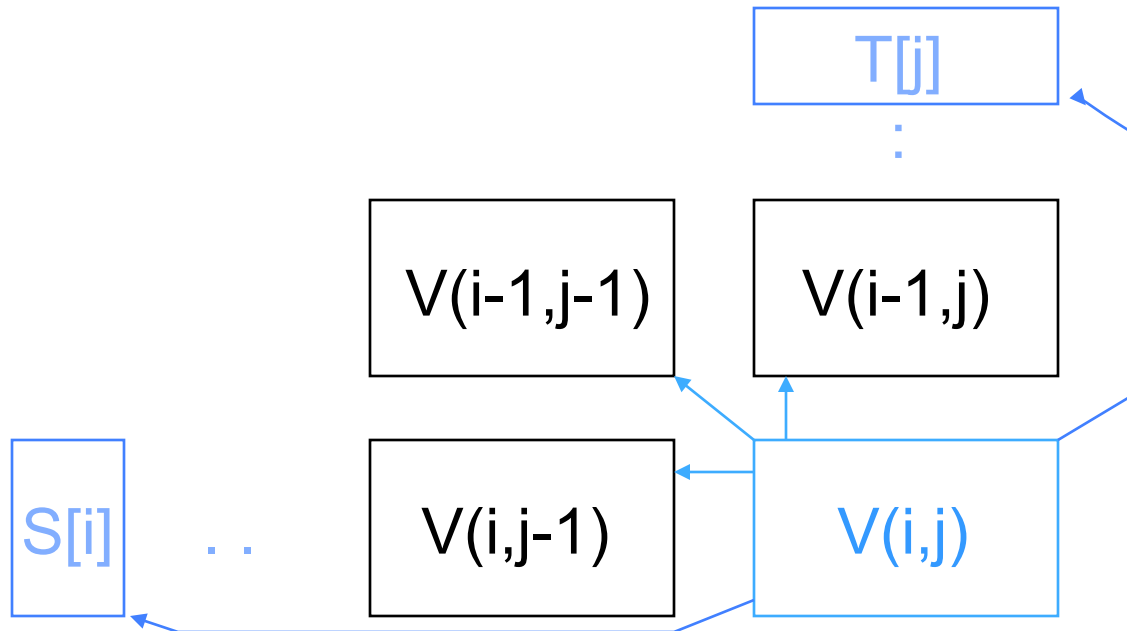
Opt align of
 $S_1 \dots S_{i-1}$ &
 $T_1 \dots T_{j-1}$

$$V(i,j) = \max \left\{ \begin{array}{l} V(i-1,j-1) + \sigma(S[i],T[j]) \\ V(i-1,j) + \sigma(S[i], -) \\ V(i,j-1) + \sigma(-, T[j]) \end{array} \right\},$$

for all $1 \leq i \leq n, 1 \leq j \leq m$.

Calculating One Entry

$$V(i,j) = \max \left\{ \begin{array}{l} V(i-1,j-1) + \sigma(S[i],T[j]) \\ V(i-1,j) + \sigma(S[i], -) \\ V(i,j-1) + \sigma(-, T[j]) \end{array} \right\}$$



Mismatch = -1
Match = 2

Example

i \ j	0	1	2	3	4	5
0	0	-1	-2	-3	-4	-5
1	-1					
2	-2					
3	-3					
4	-4					
5	-5					
6	-6					

← T

↑ S

c
-

 Score(c,-) = -1

Mismatch = -1
Match = 2

Example

	j	0	1	2	3	4	5	
i			c	a	d	b	d	←T
0		0	-1	-2	-3	-4	-5	
1	a	-1						
2	c	-2						
3	b	-3						
4	c	-4						
5	d	-5						
6	b	-6						

↑S

-
a

 Score(-,a) = -1

Mismatch = -1
 Match = 2

Example

	j	0	1	2	3	4	5
i			c	a	d	b	d
0		0	-1	-2	-3	-4	-5
1	a	-1					
2	c	-2					
3	b	-3					
4	c	-4					
5	d	-5					
6	b	-6					

← T

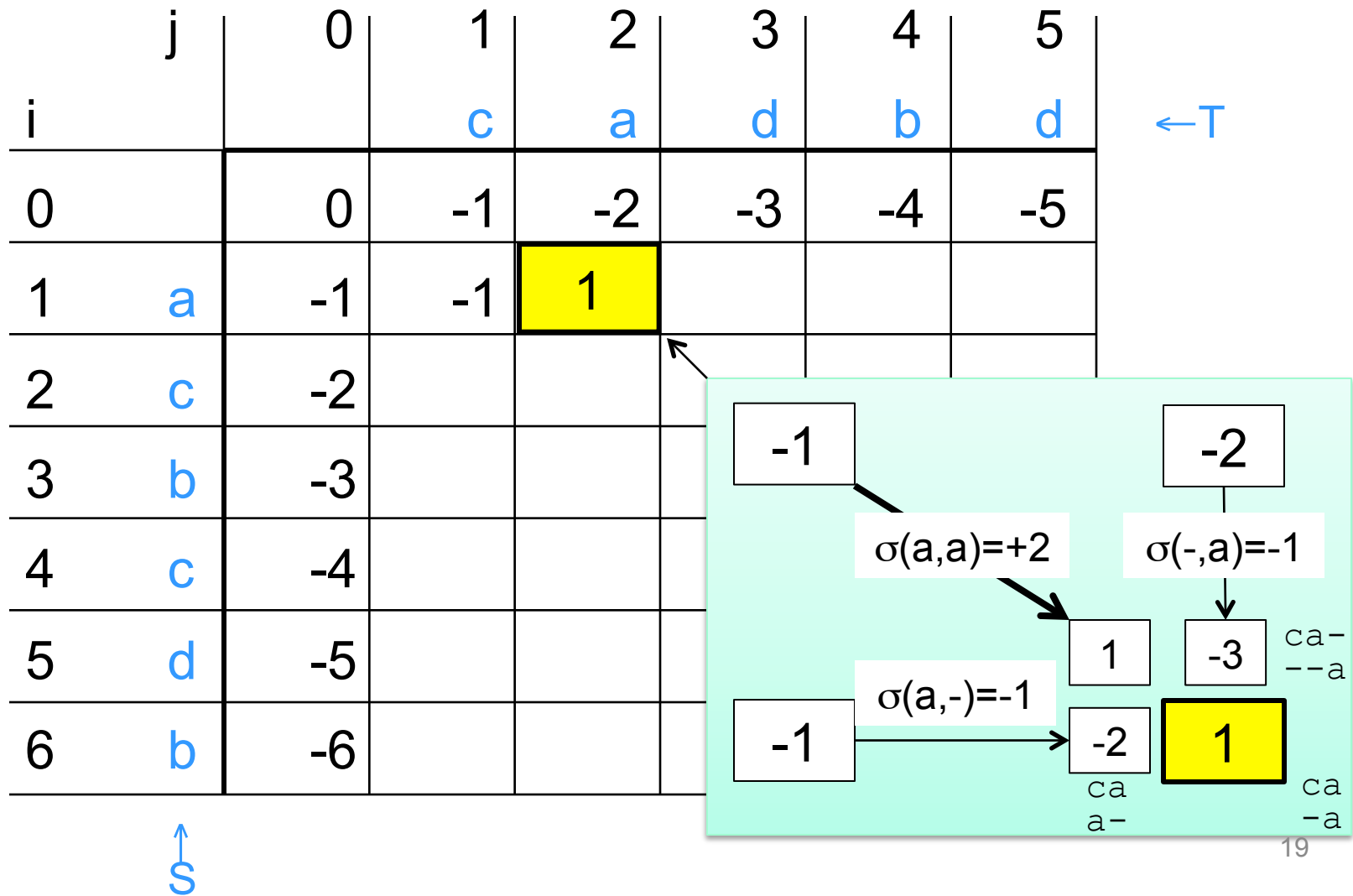
-	-
a	c
-1	

Score(-,c) = -1

↑ S

Mismatch = -1
Match = 2

Example



Example

Mismatch = -1

Match = 2

i \ j	0	1	2	3	4	5
0	0	-1	-2	-3	-4	-5
1	a	-1	-1	1		
2	c	-2	1			
3	b	-3				
4	c	-4				
5	d	-5				
6	b	-6				

←T

Time =
O(mn)

↑S

Mismatch = -1
Match = 2

Example

	j	0	1	2	3	4	5	
i			c	a	d	b	d	←T
0		0	-1	-2	-3	-4	-5	
1	a	-1	-1	1	0	-1	-2	
2	c	-2	1	0	0	-1	-2	
3	b	-3	0	0	-1	2	1	
4	c	-4	-1	-1	-1	1	1	
5	d	-5	-2	-2	1	0	3	
6	b	-6	-3	-3	0	3	2	

↑
S

Finding Alignments: Trace Back

Arrows = (ties for) max in $V(i,j)$; 3 LR-to-UL paths = 3 optimal alignments

	j	0	1	2	3	4	5	
i			c	a	d	b	d	←T
0		0	-1	-2	-3	-4	-5	
1	a	-1	-1	1	0	-1	-2	
2	c	-2	1	0	0	-1	-2	
3	b	-3	0	0	-1	2	1	
4	c	-4	-1	-1	-1	1	1	
5	d	-5	-2	-2	1	0	3	
6	b	-6	-3	-3	0	3	2	

↑S

Complexity Notes

Time = $O(mn)$, (value and alignment)

Space = $O(mn)$

Easy to get **value** in Time = $O(mn)$ and
Space = $O(\min(m,n))$

Possible to get value *and alignment* in
Time = $O(mn)$ and Space = $O(\min(m,n))$
(KT section 6.7)

Significance of Alignments

Is “42” a good score?

Compared to what?

Usual approach: compared to a specific “null model”, such as “random sequences”

Interesting stats problem; much is known

Variations

Local Alignment

Preceding gives *global* alignment, i.e. full length of both strings;

Might well miss strong similarity of part of strings amidst dissimilar flanks

Gap Penalties

10 adjacent spaces cost 10 x one space?

Many others

Similarly fast DP algs often possible

Summary: Alignment

Functionally similar proteins/DNA often have recognizably similar sequences even after eons of divergent evolution

Ability to find/compare/experiment with “same” sequence in other organisms is a huge win

Surprisingly simple scoring works well in practice: score positions separately & add, usually w/ fancier gap model like affine

Simple dynamic programming algorithms can find *optimal* alignments under these assumptions in poly time (product of sequence lengths)

This, and heuristic approximations to it like BLAST, are workhorse tools in molecular biology, and elsewhere.

Summary: Dynamic Programming

Keys to D.P. are to

- a) identify the subproblems (usually repeated/overlapping)
- b) solve them in a careful order so all small ones solved before they are needed by the bigger ones, and
- c) build table with solutions to the smaller ones so bigger ones just need to do table lookups (*no* recursion, despite recursive formulation implicit in (a))
- d) Implicitly, optimal solution to whole problem devolves to optimal solutions to subproblems

A really important algorithm design paradigm