

# Algorithms in Bioinformatics

More dynamic  
programming



# Recap

- HW - up after class  
due on Tuesday the 2<sup>nd</sup>
- No class next Thursday

# Dynamic Programming: aka "Smart Recursion"

High level idea:

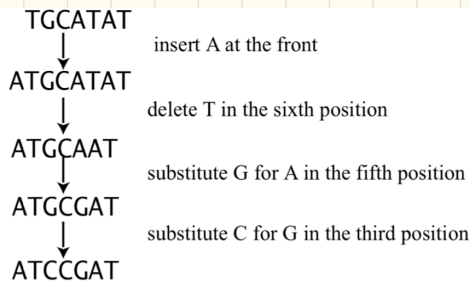
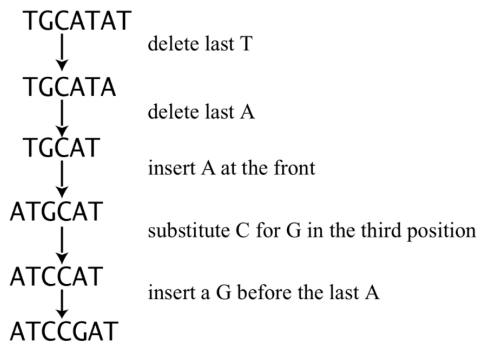
- Formulate recursion
- Notice recursive alg.  
is slow
- Store recursive call  
answers for lookup

↳

Wait, could just do  
this iteratively

# Last time: Edit distance

Example:



Alignment matrix:

A	T	-	G	T	T	A	T	-
A	T	C	G	T	-	A	-	C

(at most  $m+n$  columns)

Another way:

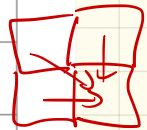
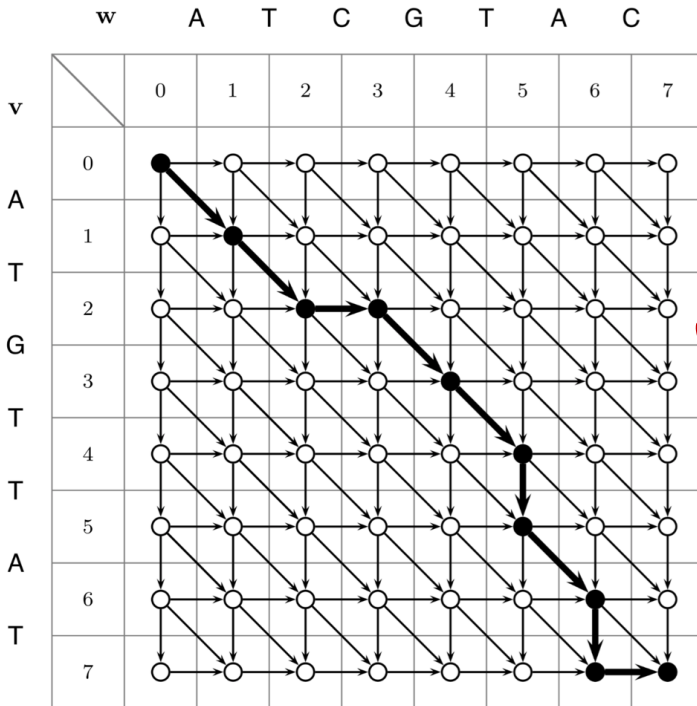
Write # of repetitions:

v =	0	1	2	2	3	4	5	6	7	7
	A	T	-	G	T	T	A	T	-	
w =	A	T	C	G	T	-	A	-	C	
	0	1	2	3	4	5	5	6	6	7

This leads to formulation from last time.

Book's view:

$v$	=	0	1	2	2	3	4	5	6	7	7	
			A	T	-	G	T	T	A	T	-	
$w$	=		A	T	C	G	T	-	A	-	C	
			0	1	2	3	4	5	5	6	6	7



↘	↘	→	↘	↘	↓	↘	↓	→
A	T	-	G	T	T	A	T	-
A	T	C	G	T	-	A	-	C

# Now : Longest Common Subsequence

**Longest Common Subsequence Problem:**

Find the longest subsequence common to two strings.

ATGAA  
GTAAA

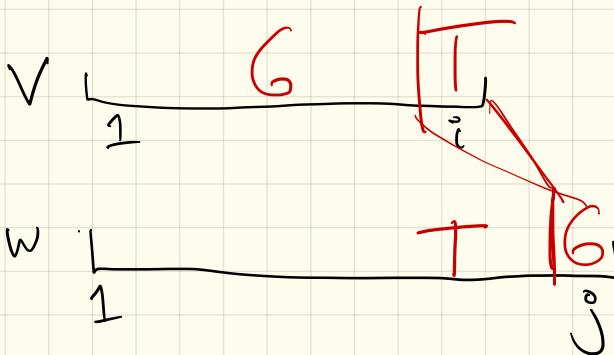
**Input:** Two strings,  $v$  and  $w$ .

**Output:** The longest common subsequence of  $v$  and  $w$ .

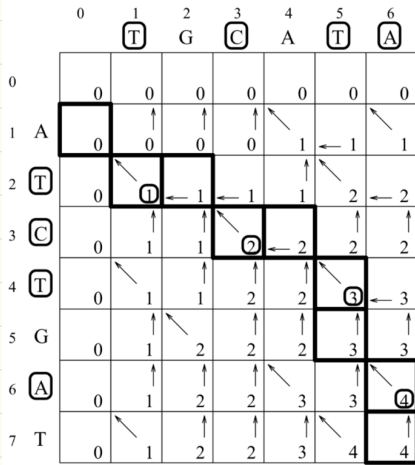
Let  $s_{i,j}$  = length of LCS  
between  $v[1..i]$  and  
 $w[1..j]$ .

If I look at last letter(s),  
2 options:

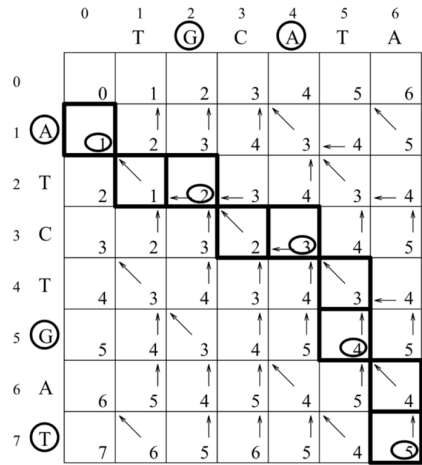
- last letter is common
- last letter isn't



Aside: LCS  $\neq$  Edit distance  
are different!



Computing similarity  $s(V,W)=4$   
V and W have a subsequence TCTA in common



Computing distance  $d(V,W)=5$   
V can be transformed into W by deleting A,G,T and inserting G,A

Alignment:     A T - C - T G A T  
                 - T G C A T - A -

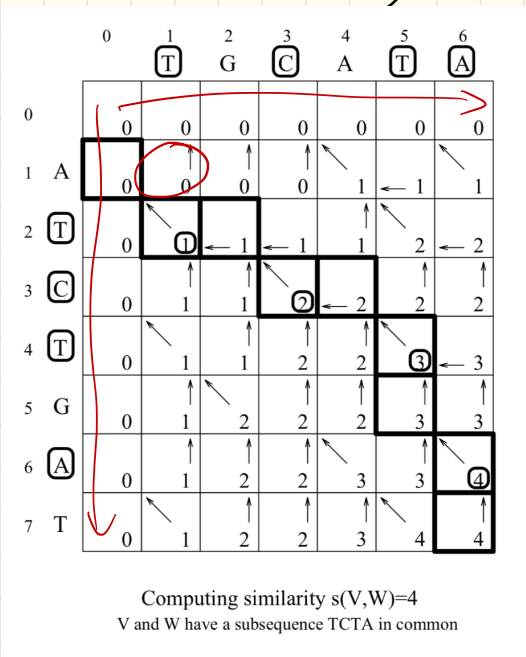
(But connected still...)  
see HW..

So: Try them all!  
 $V[1..i], w[1..j]$

$$s_{i,j} = \max \begin{cases} s_{i-1,j} \\ s_{i,j-1} \\ s_{i-1,j-1} + 1, \text{ if } v_i = w_j \end{cases}$$

↑ these are in LCS

Then think about not recomputing -  
 so store in big table  
 again



Alignment:

A	T	-	C	-	T	G	A	T
-	T	G	C	A	T	-	A	-



# Pseudo code :

LCS(v, w)

1 **for**  $i \leftarrow 0$  **to**  $n$

2      $s_{i,0} \leftarrow 0$

3 **for**  $j \leftarrow 1$  **to**  $m$

4      $s_{0,j} \leftarrow 0$

5 **for**  $i \leftarrow 1$  **to**  $n$

6     **for**  $j \leftarrow 1$  **to**  $m$

7          $s_{i,j} \leftarrow \max \begin{cases} s_{i-1,j} \\ s_{i,j-1} \\ s_{i-1,j-1} + 1, & \text{if } v_i = w_j \end{cases}$

8          $b_{i,j} \leftarrow \begin{cases} \text{"}\uparrow\text{"} & \text{if } s_{i,j} = s_{i-1,j} \\ \text{"}\leftarrow\text{"} & \text{if } s_{i,j} = s_{i,j-1} \\ \text{"}\swarrow\text{"}, & \text{if } s_{i,j} = s_{i-1,j-1} + 1 \end{cases}$

9 **return**  $(s_{n,m}, \mathbf{b})$

PRINTLCS( $\mathbf{b}, \mathbf{v}, i, j$ )

1 **if**  $i = 0$  **or**  $j = 0$

2     **return**

3 **if**  $b_{i,j} = \text{"}\swarrow\text{"}$

4     PRINTLCS( $\mathbf{b}, \mathbf{v}, i - 1, j - 1$ )

5     **print**  $v_i$

6 **else**

7     **if**  $b_{i,j} = \text{"}\uparrow\text{"}$

8         PRINTLCS( $\mathbf{b}, \mathbf{v}, i - 1, j$ )

9     **else**

10         PRINTLCS( $\mathbf{b}, \mathbf{v}, i, j - 1$ )

Now: Back to some biology!

LCS is a way to score similarity:

- +1 for a match
- +0 for a mismatch (indel)

Edit distance is too!

- insertion, deletion + substitution all cost +1.

Biology changes are more complex..

Generalize:

Make a scoring matrix.  $\delta$ :

Ex:  $\delta$ :

	G	C	T	A	-
G	1				
C		1	-3		
T			1		
A				1	
-					1

Annotations:

- Red arrow pointing to the diagonal: usually negative
- Red circle around -3: C-T mutation charge
- Red circle around +1: T insertion/deletion charge

# New goal :

## Global Alignment Problem:

Find the best alignment between two strings under a given scoring matrix.

**Input:** Strings  $v$ ,  $w$  and a scoring matrix  $\delta$ .

**Output:** An alignment of  $v$  and  $w$  whose score (as defined by the matrix  $\delta$ ) is maximal among all possible alignments of  $v$  and  $w$ .

Same type of recursion:

When looking at  $v[1..i]$   
and  $w[1..j]$

- could match ~~them~~  $v[i] + w[j]$

Score  $\rightarrow T[i, j-1] + \delta(v[i], w[j])$   
for  $v[1..i-1]$   
for  $w[1..j-1]$

- could match  $v[i]$  to -

$T[i-1, j] + \delta(v[i], -)$

- could match  $w[j]$  to -

$T[i, j-1] + \delta(-, w[j])$

End recurrence:

$$s_{i,j} = \max \begin{cases} s_{i-1,j} + \delta(v_i, -) \\ s_{i,j-1} + \delta(-, w_j) \\ s_{i-1,j-1} + \delta(v_i, w_j) \end{cases}$$

remove  $v[i]$  + recurse

remove  $w[j]$  + rec.

match last two

Example:

- charge mismatches by  $-\mu$
- indels by  $-\sigma$
- and add  $+1$  for matches

Get score =

$$\begin{aligned} & \# \text{ matches} - \mu (\# \text{ mismatches}) \\ & = \sigma (\# \text{ indels}) \end{aligned}$$

Then:

$$s_{i,j} = \max \begin{cases} s_{i-1,j} - \sigma \\ s_{i,j-1} - \sigma \\ s_{i-1,j-1} - \mu, \text{ if } v_i \neq w_j \\ s_{i-1,j-1} + 1, \text{ if } v_i = w_j \end{cases}$$

(Note: LCS is example of this!)

Runtime:  $\sigma = 0$  same alg!  $O(mn)$

# How to get $\delta$ ?

- For DNA, usually just ask for  $\mu$  &  $\sigma$  as part of input.
- For amino acids - harder!
  - Point Accepted Mutations (PAM)
  - Block Substitution (BLOSUM)  
(Two most common ones)

Reason: probability that Ser mutates to Phe is  $\approx 3$  times higher than Trp to Phe.

But - gets complex!

(Read 6.7 for details - it's constructed iteratively.)

# Local alignment

Instead of entire string, might want to find substrings with good alignment.

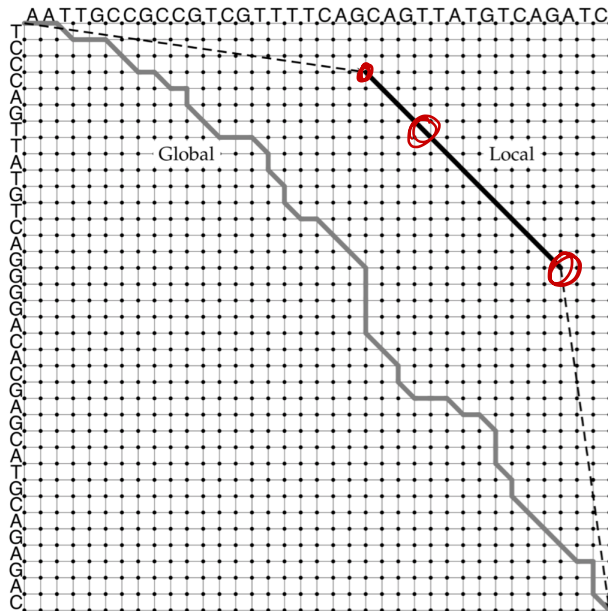
(Book's example: homeobox genes)

High level issue:

431

```
--T--CC-C-AGT--TATGT-CAGGGGACACG--A-GCATGCAGA-GAC
|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
AATTGCCGCC-GTCGT-T-TTCAG----CA-GTTATG--T-CAGAT--C

      tccAGTTATGTcAGgggacacgagcatgcagagac
      | | | | | | | | | | | | | | | | | | | |
aattgccgccgtcggttttcagCAGTTATGTcAGatc
```



Essentially, since we charge:

- $-μ$  for mismatches
- $-σ$  for indels,

Global alignment will likely miss this good substring entirely.

So:

$$S_{ij} =$$

#### Local Alignment Problem:

Find the best local alignment between two strings.



**Input:** Strings  $v$  and  $w$  and a scoring matrix  $\delta$ .

**Output:** Substrings of  $v$  and  $w$  whose global alignment, as defined by  $\delta$ , is maximal among all global alignments of all substrings of  $v$  and  $w$ .

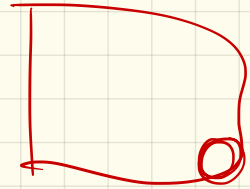
However, this is still pretty close -

- instead of best  $(0,0)$  to  $(m,n)$  path
- Want best  $(i,j)$  to  $(i',j')$  path, for any  $i',j',i,j$

Really — just need to let  
 you start match at any  
 $(i, j)$  if matching  $v[1..i]$   
 +  $w[1..j]$  is bad

So:

$$s_{i,j} = \max \begin{cases} 0 \\ s_{i-1,j} + \delta(v_i, -) \\ s_{i,j-1} + \delta(-, w_j) \\ s_{i-1,j-1} + \delta(v_i, w_j) \end{cases}$$



Then:

- Build a matrix
- fill it in
- Look at entire
- $m \times n$  matrix for highest score

→ for  $i \leftarrow 1$  to  $n$

→ for  $j \leftarrow 1$  to  $m$

Runtime:  
 $O(mn) + O(mn)$  (4 lookups, 3 additions)  
 →  $O(mn)$  (to take max)  
 also  $O(mn)$  space



Next time:

- Alignment w/ Gap penalties
- Multiple alignment

Read 6.3  
for next time



