

“If you align expectations with reality, you will never be disappointed.”
-- Terrell Owens

Dynamic Programming

- Breaks the problem into overlapping subproblems
- Uses memorization to keep solutions to subproblems we've already seen.
- works for either DNA or protein sequences, although the substitution matrices used differ
- finds an optimal alignment

Global alignment algorithm: *Needleman-Wunsch.*

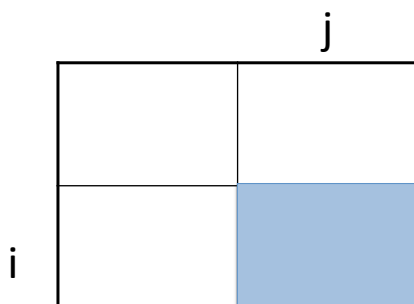
- Align sequence x and y.
- F is the DP matrix; s is the substitution matrix; d is the linear gap penalty.

$$F(0,0) = 0$$

$$F(i, j) = \max \begin{cases} F(i-1, j-1) + s(x_i, y_j) \\ F(i-1, j) + d \\ F(i, j-1) + d \end{cases}$$

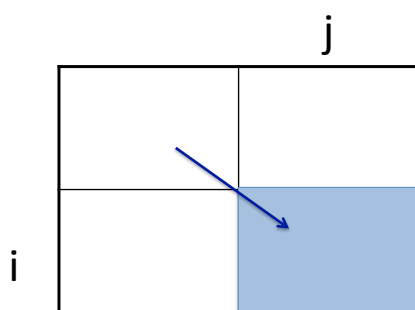
Dynamic Programming Idea

- three possible options; in each we'll choose a different pairing for end of alignment, and add this to the best alignment of previous characters, consider position i in seq 1 and position j in seq2:



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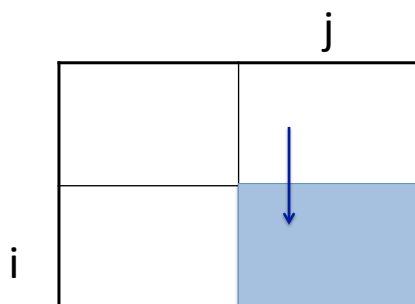


If i aligns to j , then the previous position in each sequence must be $i-1$ and $j-1$.

My score must change by $s(i,j)$.

Dynamic Programming Idea

- three possible options; in each we'll choose a different pairing for end of alignment, and add this to the best alignment of previous characters, consider position i in seq 1 and position j in seq2:

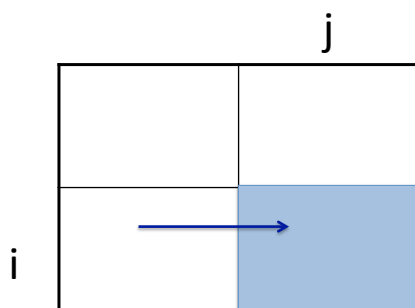


If i aligns to a gap, then the previous position in each sequence must be $i-1$ and j .

My score must change by the gap penalty.

Dynamic Programming Idea

- three possible options; in each we'll choose a different pairing for end of alignment, and add this to the best alignment of previous characters, consider position i in seq 1 and position j in seq2:

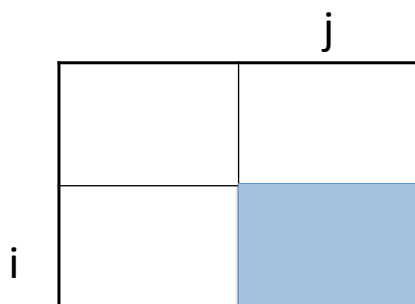


If j aligns to a gap, then the previous position in each sequence must be i and $j-1$.

My score must change by the gap penalty.

Dynamic Programming Idea

- three possible options; in each we'll choose a different pairing for end of alignment, and add this to the best alignment of previous characters, consider position i in seq 1 and position j in seq2:



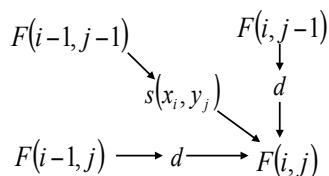
Regardless which choice is best, the value in $F(i,j)$ now contains the best global alignment score of $\text{seq1}(1\dots i)$ to $\text{seq2}(1\dots j)$.

Needleman-Wunsch

	A	C	G	T
A	2	-7	-5	-7
C	-7	2	-7	-5
G	-5	-7	2	-7
T	-7	-5	-7	2

Find the optimal alignment of AAG and AGC.
Use a gap penalty of $d=-5$.

		A	A	G
	0	-5	-10	-15
A	-5	2	-3	-8
G	-10	-3	-3	-1
C	-15	-8	-8	-6



Keeping track of which choice is made
(by the arrows) enables rapid
traceback.

Needleman-Wunsch

Find the optimal alignment of AAG and AGC.
Use a gap penalty of $d=-5$.

- The best score for the entire global alignment is ALWAYS in the lower right corner.
- Following the arrows backwards and recording each choice can recover the optimal alignment.

		A	A	G
	0	-5		
A		2	-3	
G				-1
C				-6

AAG- AAG-
-AGC A-GC

Needleman-Wunsch

Find the optimal alignment of AAG and AGC.
Use a gap penalty of $d=-5$.

- Each arrow introduces one character at the end of each aligned sequence.
- A horizontal move puts a gap in the left sequence.
- A vertical move puts a gap in the top sequence.
- A diagonal move uses one character from each sequence.

		A	A	G
	0	→ -5		
A		↘ 2	→ -3	
G				↘ -1
C				↓ -6

AAG-
-AGC

AAG-
A-GC

Computational Complexity

- initialization: $O(m)$, $O(n)$ where sequence lengths are m , n
- filling in rest of matrix: $O(mn)$
- traceback: $O(m + n)$
- hence, if sequences have nearly same length, the computational complexity is

$$O(n^2)$$

Notice!

$$O(n^2) \ll \binom{2n}{n}$$

For example: $n = 11$

Needleman-Wunch = 121

Total Alignments = 705,432

So Needleman-Wunch is at least
5830 fold faster than iterating
through all possible alignments!!

Needleman-Wunch Summary

- Requires:
 - A scoring scheme for matches, mismatches and gaps
 - A recurrence relationship (the iteration steps):

$$F(0,0) = 0$$

$$F(i,j) = \max \begin{cases} F(i-1,j-1) + s(x_i, y_j) \\ F(i-1,j) + d \\ F(i,j-1) + d \end{cases}$$

- There are 2 parts to the algorithm: fill (find best score in lower right) and traceback (recover alignment)

Smith-Waterman: Local alignment

- To be local, should be able to “start over” if the score gets too bad. How?
- Two differences with respect to global alignment:
 - No score is negative.
 - Traceback begins at the highest score in the matrix and continues until you reach 0.


Local Alignment DP Algorithm

- original formulation: Smith & Waterman, *Journal of Molecular Biology*, 1981
- interpretation of array values is somewhat different
 - $F(i, j)$ = score of the best alignment of a prefix of $x[1 \dots i]$ and a prefix of $y[1 \dots j]$

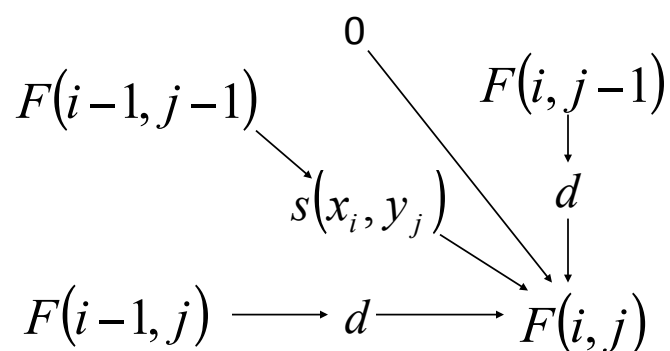
Local alignment DP

- Align sequence x and y.
- F is the DP matrix; s is the substitution matrix; d is the linear gap penalty.

$$F(0,0) = 0$$

$$F(i, j) = \max \begin{cases} F(i-1, j-1) + s(x_i, y_j) \\ F(i-1, j) + d \\ F(i, j-1) + d \\ 0 \end{cases}$$


Local DP in equation form

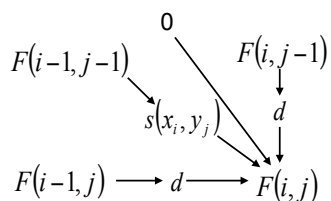


Smith-Waterman

	A	C	G	T
A	2	-7	-5	-7
C	-7	2	-7	-5
G	-5	-7	2	-7
T	-7	-5	-7	2

Find the optimal local alignment of AAG and AGC.
Use a gap penalty of $d=-5$.

		A	A	G
A				
G				
C				

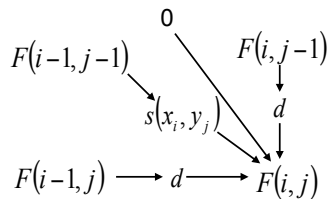


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Find the optimal local alignment of AAG and AGC.
Use a gap penalty of $d=-5$.

		A	A	G
	0	0	0	0
A	0			
G	0			
C	0			

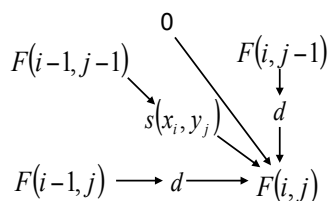


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Find the optimal local alignment of AAG and AGC.
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		A	A	G
	0	0	0	0
A	0	2	-5	
G	0	-5	0	
C	0			

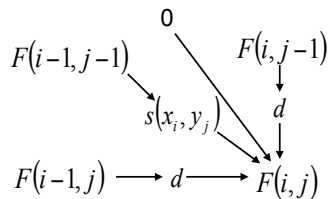


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Find the optimal local alignment of AAG and AGC.
Use a gap penalty of $d=-5$.

		A	A	G
	0	0	0	0
A	0	2		
G	0	?		
C	0	?		

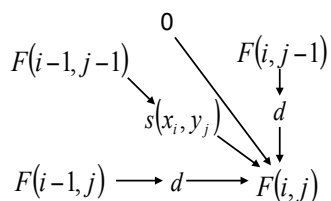


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A	2	-7	-5	-7
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Find the optimal local alignment of AAG and AGC.
Use a gap penalty of $d=-5$.

		A	A	G
	0	0	0	0
A	0	2	?	?
G	0	0	?	?
C	0	0	?	?

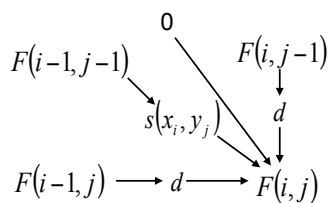


Smith-Waterman

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A	2	-7	-5	-7
C	-7	2	-7	-5
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T	-7	-5	-7	2

Find the optimal local alignment of AAG and AGC.
Use a gap penalty of $d=-5$.

		A	A	G
	0	0	0	0
A	0	2	2	0
G	0	0	0	4
C	0	0	0	0

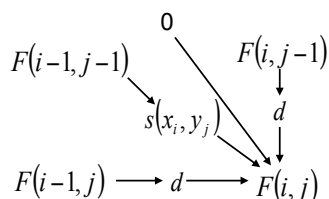


Smith-Waterman

	A	C	G	T
A	2	-7	-5	-7
C	-7	2	-7	-5
G	-5	-7	2	-7
T	-7	-5	-7	2

Find the optimal local alignment of AAG and AGC.
Use a gap penalty of $d=-5$.

		A	A	G
	0	0	0	0
A	0	2	2	0
G	0	0	0	4
C	0	0	0	0



AG
AG

Smith-Waterman

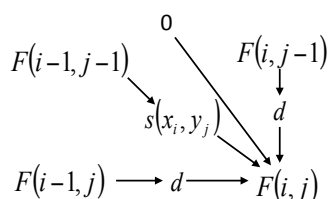
- So the best score is the MAX value of the table.
 - Not necessarily in lower right corner!
- And you traceback until you reach ZERO, i.e. until the score says “you started over here.”
 - Not necessarily in upper left corner!

Local alignment

	A	C	G	T
A	2	-7	-5	-7
C	-7	2	-7	-5
G	-5	-7	2	-7
T	-7	-5	-7	2

Find the optimal local alignment of AAG and GAAGGC.
Use a gap penalty of $d=-5$.

		A	A	G
	0	0	0	0
G	0	0	0	2
A	0	2	2	0
A	0	2	4	0
G	0	0	0	6
G	0	0	0	2
C	0	0	0	0

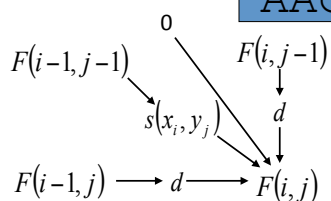


Local alignment

	A	C	G	T
A	2	-7	-5	-7
C	-7	2	-7	-5
G	-5	-7	2	-7
T	-7	-5	-7	2

Find the optimal local alignment of AAG and GAAGGC.
Use a gap penalty of $d=-5$.

		A	A	G
	0	0	0	0
G	0	0	0	2
A	0	2	2	0
A	0	2	4	0
G	0	0	0	6
G	0	0	0	2
C	0	0	0	0



Local Alignment

- Take note that for Smith-Waterman to work, at least one score in our scoring scheme **MUST** be positive!!

More On Gap Penalty Functions

- a gap of length k is more probable than k gaps of length 1
 - a gap may be due to a single mutational event that inserted/deleted a stretch of characters
 - separated gaps are probably due to distinct mutational events
- a linear gap penalty function treats these cases the same
- it is more common to use gap penalty functions involving two terms
 - a penalty g associated with opening a gap
 - a smaller penalty s for extending the gap

Gap Penalty Functions

- linear

$$w(k) = kd$$

- affine

$$w(k) = \begin{cases} g + sk, & k \geq 1 \\ 0, & k = 0 \end{cases}$$

- Linear gap penalty: every gap receives a score of d .

$$\begin{array}{c} \text{GAAT-C} \quad d = -4 \\ \text{CA-TAC} \\ \swarrow \quad \downarrow \quad \searrow \quad \swarrow \quad \searrow \\ -5 + 10 + \textcolor{blue}{-4} + 10 + \textcolor{blue}{-4} + 10 = \textcolor{blue}{17} \end{array}$$

- Affine gap penalty: opening a gap receives a score of d ; extending a gap receives a score of e .

$$\begin{array}{c} \text{G--AATC} \quad g = -3 \\ \text{CATA--C} \quad s = -1 \\ \swarrow \quad \downarrow \quad \searrow \quad \swarrow \quad \searrow \\ -5 + \textcolor{blue}{-4} + \textcolor{blue}{-1} + 10 + \textcolor{blue}{-4} + \textcolor{blue}{-1} + 10 = \textcolor{blue}{5} \end{array}$$

General gap penalties

AAAGAATTCA vs. AAAGAATTCA
A-A-A-T-CA AAA----TCA

In a linear gap scoring model, these alignments have the same score. But the second is likely more biologically plausible.

Linear model: cost of a run of k gaps is $k \cdot d$

Affine gap model: cost of a run of k gaps is $g + k \cdot s$

General gap penalties

AAAGAATTCA vs. AAAGAATTCA
A-A-A-T-CA AAA----TCA

Previous dynamic programming strategy no longer works because the score of the last character depends on the details of the previous alignment (e.g. we have “broken” the independence assumption!:

AAAGAAC	AAAGAATC
AAA----	AAA----

Instead we need to “know” how long a final run of gaps is in order to give a score to the last subproblem.

Affine gap model

- Can also be solved by dynamic programming, but it is a more sophisticated model (inherently it's a Markov model!!).