

String Algorithms and Data Structures

Semi-Global and Local Alignment

CS 199-225

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Please fill out ICES Evaluations

Feedback is important for the development of the class

If not enough people fill it out, doesn't actually get recorded

Learning Objectives

Review global alignment dynamic programming

Improve the algorithm to account for local matches

Discuss further optimizations or alternatives

Approximate Pattern Matching

Input: A text T , a pattern P , and a distance d

Output: All positions in T where P has at most d mismatches or edits

P : word

T : There would have been a time for such a word:

Alignment 1: word

Alignment 2: word

~~Not a match!~~

Match!

Distance 2 match!

Distance 0 match!

Edit Distance

Alignments:

x: G C G T A T G C G G C T A - A C G C
 | | | | | | | | | | | | | | | |
 y: G C - T A T G C G G C T A T A C G C

replacement (AKA substitution/mismatch)

x: G C G T A T G A G G C T A - A C G C
 | | | | | | | | | | | | | | | |
 y: G C - T A T G C G G C T A T A C G C

x: t h e l o n g e s t - - - -
 | | | | | | | | | | | | | | | |
 y: - - - - l o n g e s t d a y

Edit string (w/r/t x):

MMDMMMMMMMMMMIMMMM

Distance = 2

MMDMMMMRMMMMMMIMMMM

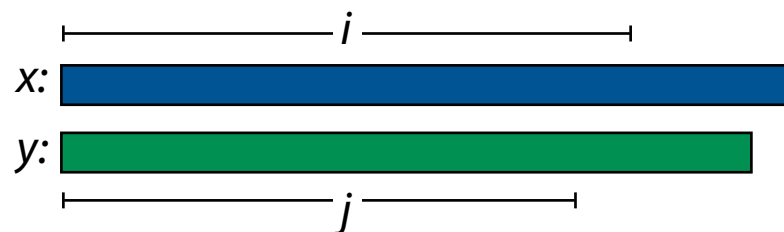
Distance = 3

DDDDMMMMMMMMIIIII

Distance = 8

Edit Distance

$D[i, j]$: edit distance between length- i prefix of x and length- j prefix of y



Optimal edit string for $D[i, j]$ is built by ***extending a shorter optimal string by 1 operation***. 3 options:

Append **D** to transcript for $D[i-1, j]$

Append **I** to transcript for $D[i, j-1]$

Append **M** or **R** to transcript for $D[i-1, j-1]$

We choose based on whichever option has the fewest edits

Edit Distance

X: GTTTAA

Y: GGTTTA

MIMMMM

D[5, 6]

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

MIMMMMD

D[6, 6]¹

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

MRMMR

D[5, 5]

G	T	T	T	A
G	G	T	T	T

MRMMRM

D[6, 6]²

G	T	T	T	A	A
G	G	T	T	T	A

MRMMRD

D[6, 5]

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

MRMMRDI

D[6, 6]³

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

Edit Distance: dynamic programming

$$D[i, j] = \min \begin{cases} D[i - 1, j] + 1 \\ D[i, j - 1] + 1 \\ D[i - 1, j - 1] + \delta(x[i - 1], y[j - 1]) \end{cases}$$

	ε	G	C	T	A	T	G	C	C	A	C	G	C
ε	0	1	2	3	4	5	6	7	8	9	10	11	12
G	1	0	1	2	3	4	5	6					
C	2	1	0	1	2	3	4	5					
G	3	2	1	1	2	3	3	4					
T	4	3	2	1	2	2	3	4					
A	5												
T	6												
G	7												
C	8												
A	9												
C	10												
G	11												
C	12												

<p>GC - - - GT</p> <p>GCTATGC</p> <p>MMIIIMR</p>	<p>GC - GT - -</p> <p>GCTATGC</p> <p>MMIRMII</p>
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What goes here in **i=4, j=7**?

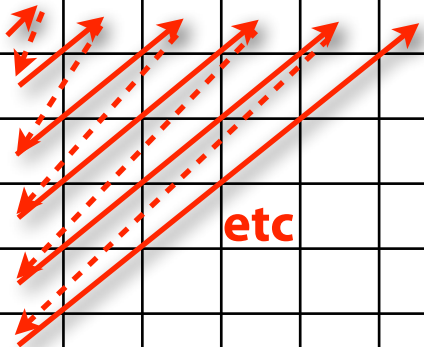
$$= \min(D[i-1, j]+1, D[i, j-1]+1, D[i-1, j-1]+delt)$$

$$= \min(4 + 1, 3 + 1, 3 + 1)$$

$$= 4$$

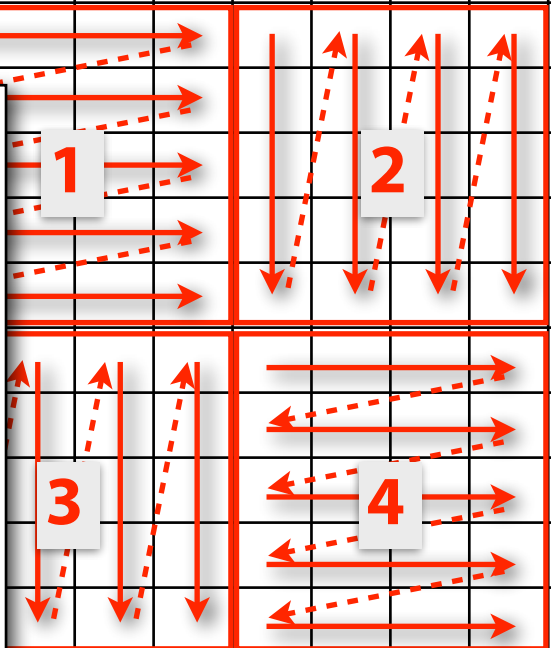
Edit Distance: dynamic programming

	ε	G	C	T	A	T	G	C	C	A	C	G	C
ε	0	1	2	3	4	5	6	7	8	9	10	11	12
G	1												
C	2												
G	3												
T	4												
A	5												
T	6												
G	7												
C	8												
A	9												
C	10												
G	11												
C	12												



Any strategy that fills in order works:

	ε	G	C	T	A	T	G	C	C	A	C	G	C
ε	0	1	2	3	4	5	6	7	8	9	10	11	12
G	1												
C	2												
G	3												
T	4												
A	5												
T	6												
G	7												
C	8												
A	9												
C	10												
G	11												
C	12												



Dynamic Programming

	e	G	C	T	A	T	G	C	C	A	C	G	C
e	0	1	2	3	4	5	6	7	8	9	10	11	12
G	1	0	1	2	3	4	5	6	7	8	9	10	11
C	2	1	0	1	2	3	4	5	6	7	8	9	10
G	3	2	1	1	2	3	3	4	5	6	7	8	9
T	4	3	2	1	2	2	3	4	5	6	7	8	9
A	5	4	3	2	1	2	3	4	5	5	6	7	8
T	6	5	4	3	2	1	2	3	4	5	6	7	8
G	7	6	5	4	3	2	1	2	3	4	5	6	7
C	8	7	6	5	4	3	2	1	2	3	4	5	6
A	9	8	7	6	5	4	3	2	2	2	3	4	5
C	10	9	8	7	6	5	4	3	2	3	2	3	4
G	11	10	9	8	7	6	5	4	3	3	3	2	3
C	12	11	10	9	8	7	6	5	4	4	3	3	2

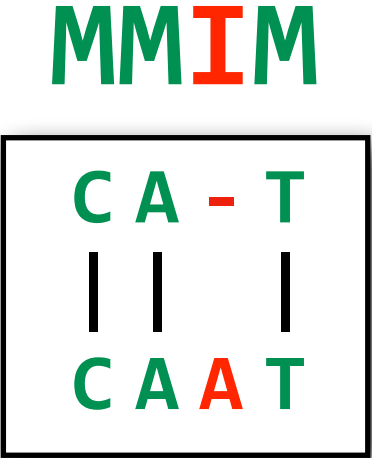
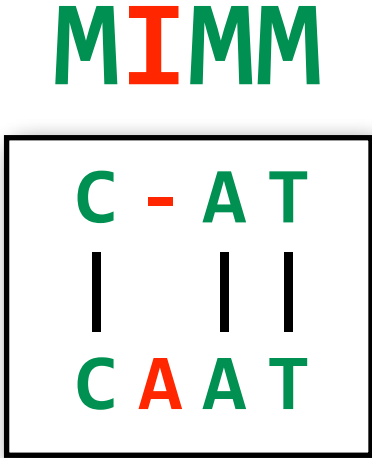
Edit Distance

Traceback corresponds to an optimal alignment / edit transcript

At each step, ask: which neighbor (\swarrow , \leftarrow or \uparrow) gave the minimum?

	ϵ	C	A	A	T
ϵ	0	1	2	3	4
C	1	0	1	2	3
A	2	1	0	1	2
T			1	1	1

Path 2



Dynamic Programming



	e	G	C	T	A	T	G	C	C	A	C	G	C
e	0	1	2	3	4	5	6	7	8	9	10	11	12
G	1	0	1	2	3	4	5	6	7	8	9	10	11
C	2	1	0	1	2	3	4	5	6	7	8	9	10
G	3	2	1	1	2	3	3	4	5	6	7	8	9
T	4	3	2	1	2	2	3	4	5	6	7	8	9
A	5	4	3	2	1	2	3	4	5	5	6	7	8
T	6	5	4	3	2	1	2	3	4	5	6	7	8
G	7	6	5	4	3	2	1	2	3	4	5	6	7
C	8	7	6	5	4	3	2	1	2	3	4	5	6
A	9	8	7	6	5	4	3	2	2	3	4	5	
C	10	9	8	7	6	5	4	3	2	3	2	3	4
G	11	10	9	8	7	6	5	4	3	3	3	2	3
C	12	11	10	9	8	7	6	5	4	4	3	3	2

Alignment:

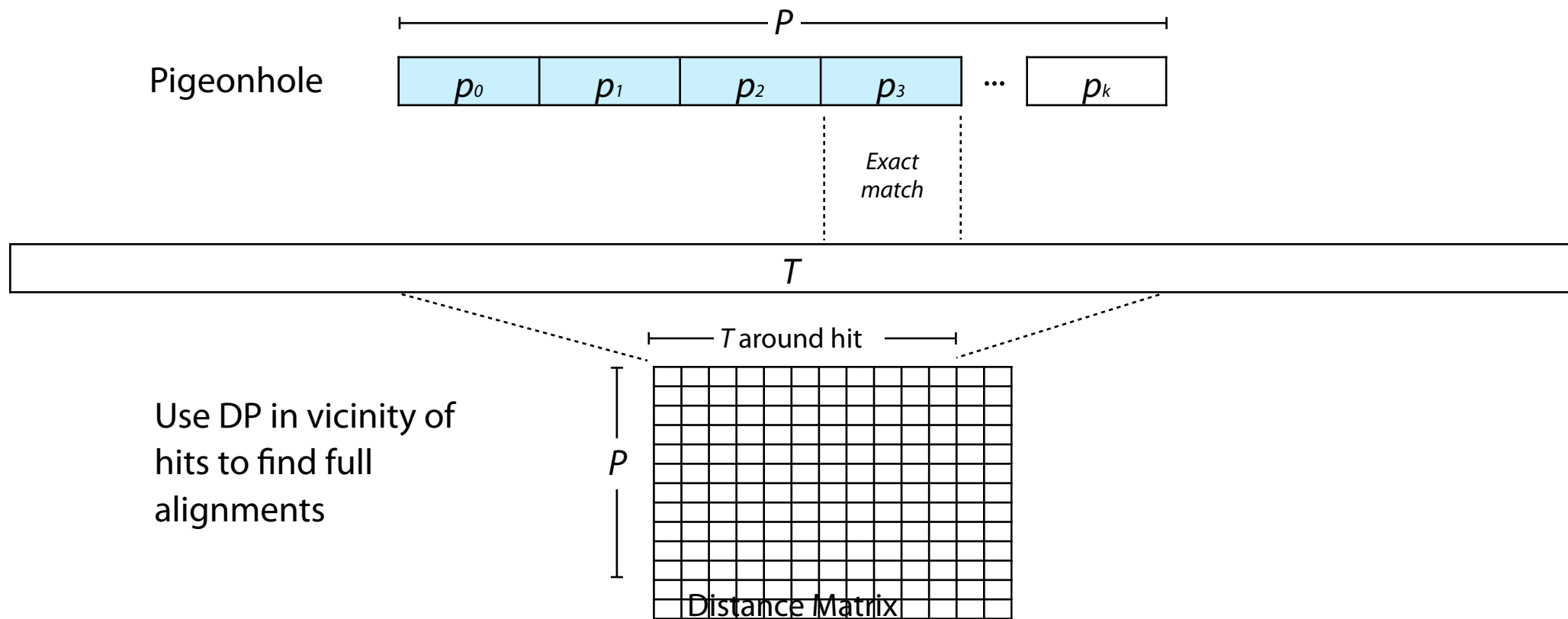
```

G C G T A T G - C A C G C
| |   | | | |   | | | |
G C - T A T G C C A C G C
  
```

MMDMMMIMMMMM

Pattern Matching w/ Dynamic Programming

“Seed and extend” works for edit distance too!



Pattern Matching w/ Dynamic Programming

Where is my best match? What is the alignment?

	ε	A	A	C	C	C	T	A	T	G	T	C	A	T	G	C	C	T	T	G	G	A
ε	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
T	1	1	1	1	1	1	0	1	0	1	0	1	1	0	1	1	1	0	0	1	1	1
A	2	1	1	2	2	2	1	0	1	1	1	1	1	1	1	2	2	1	1	1	2	1
C	3	2	2	1	2	2	2	1	1	2	2	1	2	2	2	1	2	2	2	2	2	2
G	4	3	3	2	2	3	3	2	2	1	2	2	2	3	2	2	2	3	3	2	2	3
T	5	4	4	3	3	3	3	3	2	2	1	2	3	2	3	3	3	2	3	3	3	3
C	6	5	5	4	3	3	4	4	3	3	2	1	2	3	3	3	3	3	3	4	4	4
A	7	6	5	5	4	4	4	4	4	4	3	2	1	2	3	4	4	4	4	4	5	4
G	8	7	6	6	5	5	5	5	5	4	4	3	2	2	2	3	4	5	5	4	4	5
C	9	8	7	6	6	5	6	6	6	5	5	4	3	3	3	2	3	4	5	5	5	5

Pattern Matching w/ Dynamic Programming

Where is my best match? What is the alignment?

	ε	A	A	C	C	C	T	A	T	G	T	C	A	T	G	C	C	T	T	G	G	A			
ε	0	0	0	0	0	0	0	0	0	T :	A	A	C	C	T	A	T	G	C	C	T	T	G	G	A
T	1	1	1	1	1	1	0	1	0																
A	2	1	1	2	2	2	1	0	1	P :															
C	3	2	2	1	2	2	2	1	1																
G	4	3	3	2	2	3	3	2	2																
T	5	4	4	3	3	3	3	3	2																
C	6	5	5	4	3	3	4	4	3																
A	7	6	5	5	4	4	4	4	4																
G	8	7	6	6	5	5	5	5	5																
C	9	8	7	6	6	5	6	6	6																

Pattern Matching w/ Dynamic Programming

How many matches do I have with no more than 3 edits?

	ε	A	A	C	C	C	T	A	T	G	T	C	A	T	G	C	C	T	T	G	G	A
ε	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
T	1	1	1	1	1	1	0	1	0	1	0	1	1	0	1	1	1	0	0	1	1	1
A	2	1	1	2	2	2	1	0	1	1	1	1	1	1	1	2	2	1	1	1	2	1
C	3	2	2	1	2	2	2	1	1	2	2	1	2	2	2	1	2	2	2	2	2	2
G	4	3	3	2	2	3	3	2	2	1	2	2	2	3	2	2	2	3	3	2	2	3
T	5	4	4	3	3	3	3	3	2	2	1	2	3	2	3	3	3	2	3	3	3	3
C	6	5	5	4	3	3	4	4	3	3	2	1	2	3	3	3	3	3	3	4	4	4
A	7	6	5	5	4	4	4	4	4	4	3	2	1	2	3	4	4	4	4	4	5	4
G	8	7	6	6	5	5	5	5	5	4	4	3	2	2	2	3	4	5	5	4	4	5
C	9	8	7	6	6	5	6	6	6	5	5	4	3	3	3	2	3	4	5	5	5	5

Pattern Matching w/ Dynamic Programming

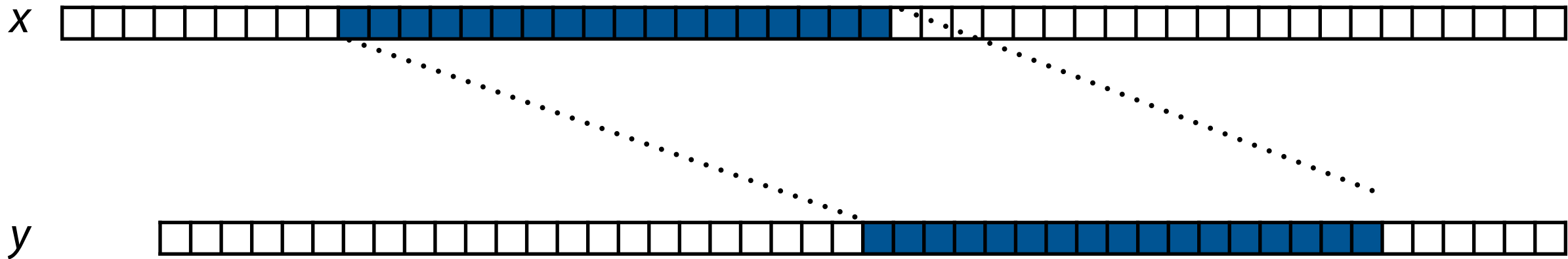


How many matches do I have with no more than 3 edits?

	ε	A	A	C	C	C	T	A	T	G	T	C	A	T	G	C	C	T	T	G	G	A
ε	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
T	1	1	1	1	1	1	0	1	0	1	0	1	1	0	1	1	1	0	0	1	1	1
A	2	1	1	2	2	2	1	0	1	1	1	1	1	1	1	2	2	1	1	1	2	1
C	3	2	2	1	2	2	2	1	1	2	2	1	2	2	2	1	2	2	2	2	2	2
G	4	3	3	2	2	3	3	2	2	1	2	2	2	3	2	2	2	3	3	2	2	3
T	5	4	4	3	3	3	3	3	2	2	1	2	3	2	3	3	3	2	3	3	3	3
C	6	5	5	4	3	3	4	4	3	3	2	1	2	3	3	3	3	3	3	4	4	4
A	7	6	5	5	4	4	4	4	4	4	3	2	1	2	3	4	4	4	4	4	5	4
G	8	7	6	6	5	5	5	5	5	4	4	3	2	2	2	3	4	5	5	4	4	5
C	9	8	7	6	6	5	6	6	6	5	5	4	3	3	3	2	3	4	5	5	5	5

Local Alignment

Local alignment is the process of comparing two strings x and y and finding the optimal alignment value of a *substring* of x to a *substring* of y .



How does this alter our dynamic programming algorithm?

Local Alignment

Local alignment is the process of comparing two strings x and y and finding the optimal alignment value of a *substring* of x to a *substring* of y .

x he_will_after_his_sour_fashion_tell_you

y struts_and_frets_his_hour_upon_the_stage

his_sour_
| | | | | | | | | |
his_hour_

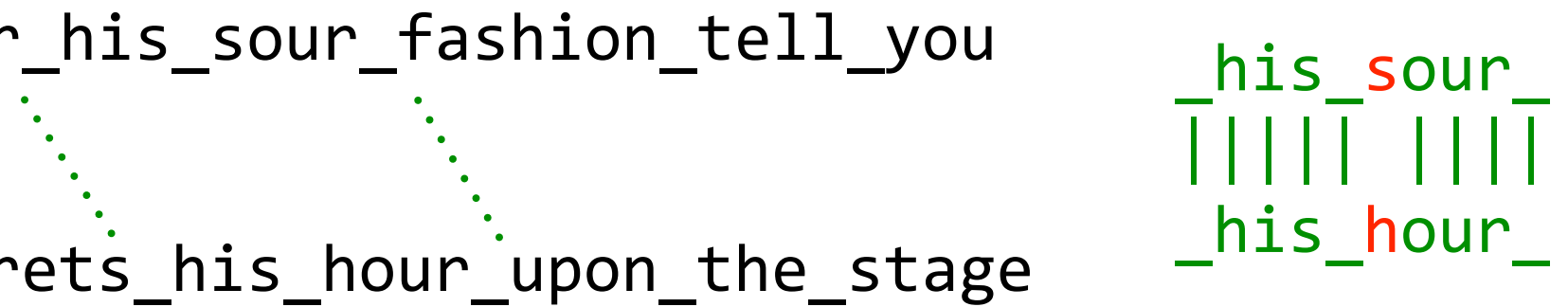
Naive: Perform a global alignment for every possible pair of substrings.

Local Alignment

Local alignment is the process of comparing two strings x and y and finding the optimal alignment value of a *substring* of x to a *substring* of y .

x he_will_after_his_sour_fashion_tell_you

y struts_and_frets_his_hour_upon_the_stage



his_sour_
| | | | | | | |
his_hour_

Better: Lets do this in $O(|X| |Y|)$ — same big O as the global! How?

From global to local alignments

To match TAC with any substring of AAC:

To match AAC with any substring of TAC:

To match any substring of AAC with any substring of TAC:

	ε	A	A	C
ε				
T				
A				
C				

From global to local alignments



Change #1: Give a scoring function for edits

	ϵ	A	A	C
ϵ	0	0	0	0
T	0	0	0	0
A	0	1	1	0
C	0	0	0	2

Change #2: Convert from distance to similarity

Change #3: Allow our alignment to start from any position in the matrix

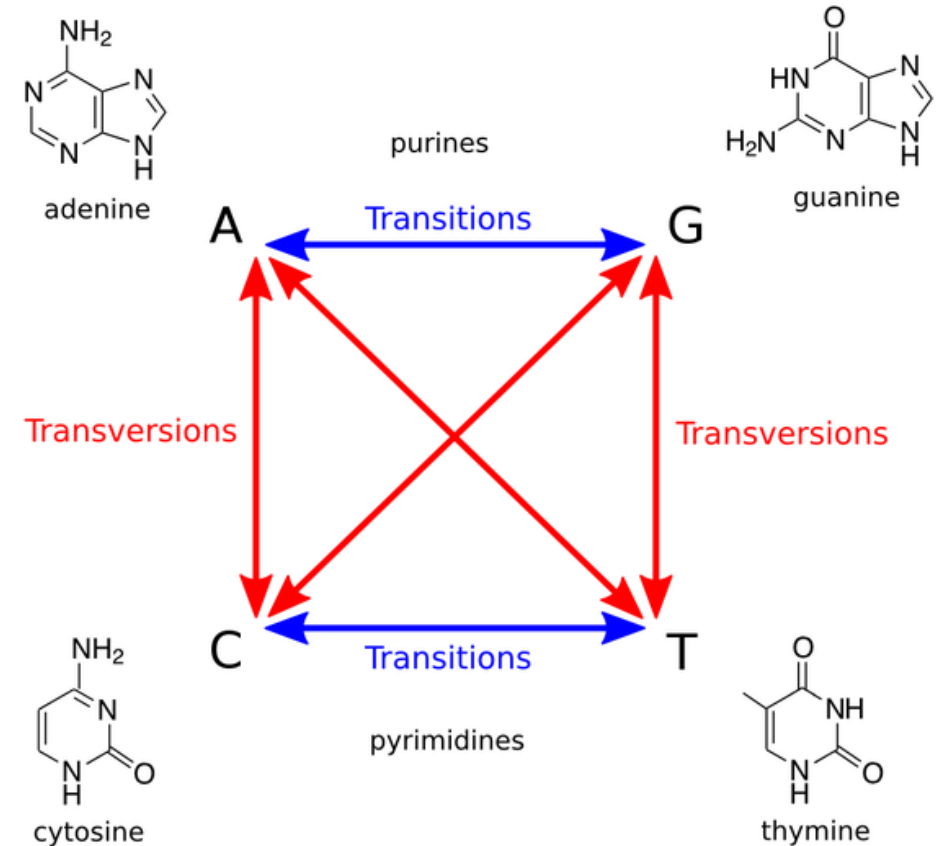
Dynamic Programming: Scoring Function

Transitions are $A \leftrightarrow G$ and $C \leftrightarrow T$ changes

Transversions are $A \leftrightarrow C$, $A \leftrightarrow T$, $C \leftrightarrow G$, $G \leftrightarrow T$

For random mutations, transitions should be half as frequent as transversions...

... **but** the transition to transversion ratio is ~ 2.1 in humans



Dynamic Programming: Scoring Function

	A	C	G	T	-
A	0	4	2	4	8
C	4	0	4	2	8
G	2	4	0	4	8
T	4	2	4	0	8
-	8	8	8	8	

2

Transitions (A ↔ G, C ↔ T)

4

Transversions

8

Gaps

Dynamic Programming: Scoring Function

Let $D[0, j] = \sum_{k=0}^{j-1} s(-, y[k])$, and let $D[i, 0] = \sum_{k=0}^{i-1} s(x[k], -)$

Otherwise, let $D[i, j] = \min \begin{cases} D[i-1, j] + s(x[i-1], -) \\ D[i, j-1] + s(-, y[j-1]) \\ D[i-1, j-1] + s(x[i-1], y[j-1]) \end{cases}$

$s(a, b)$ assigns a cost to a particular gap or substitution

$s(a, b)$:

	A	C	G	T	-
A	0	4	2	4	8
C	4	0	4	2	8
G	2	4	0	4	8
T	4	2	4	0	8
-	8	8	8	8	

2

Transitions (A↔G, C↔T)

4

Transversions (everything else)

8

Gaps

Dynamic Programming: Scoring Function

Adding a scoring function changes little in the method:

$$s(a, b)$$

	ϵ	T	A	T
ϵ	0	1	2	3
T	1			
A	2			
C	3			

	ϵ	T	A	T
ϵ	0	8	16	24
T	8			
A	16			
C	24			

	A	C	G	T	-
A	0	1	1	1	1
C	1	0	1	1	1
G	1	1	0	1	1
T	1	1	1	0	1
-	1	1	1	1	

	A	C	G	T	-
A	0	4	2	4	8
C	4	0	4	2	8
G	2	4	0	4	8
T	4	2	4	0	8
-	8	8	8	8	

Dynamic Programming: Scoring Function

Adding a scoring function changes little in the method:

$$s(a, b)$$

	ε	T	A	T
ε	0	1	2	3
T	1	0	1	2
A	2	1	0	1
C	3	2	1	1

	ε	T	A	T
ε	0	8	16	24
T	8	0	8	16
A	16	8	0	8
C	24	16	8	2

	A	C	G	T	-
A	0	1	1	1	1
C	1	0	1	1	1
G	1	1	0	1	1
T	1	1	1	0	1
-	1	1	1	1	

	A	C	G	T	-
A	0	4	2	4	8
C	4	0	4	2	8
G	2	4	0	4	8
T	4	2	4	0	8
-	8	8	8	8	

Dynamic Programming: Scoring Function

Adding a scoring function changes little in the method:

	ε	T	A	T	G	T	C	A	T	G	C
ε	0	8	16	24	32	40	48	56	64	72	80
T	8	0	8	16	24	32	40	48	56	64	72
A	16	8	0	8	16	24	32	40	48	56	64
C	24	16	8	2	10	18	24	32	40	48	56
G	32	24	16	10	2	10	18	26	34	40	48
T	40	32	24	16	10	2	10	18	26	34	42
C	48	40	32	24	18	10	2	10	18	26	34
A	56	48	40	32	26	18	10	2	10	18	26
G	64	56	48	40	32	26	18	10	6	10	18
C	72	64	56	48	40	34	26	18	12	10	10

$s(a, b)$

	A	C	G	T	-
A	0	4	2	4	8
C	4	0	4	2	8
G	2	4	0	4	8
T	4	2	4	0	8
-	8	8	8	8	

← Optimal global alignment

Dynamic Programming: Scoring Function

Adding a scoring function changes little in the method:

$$s(a, b)$$

	ε	T	A	T	G	T	C	A	T	G	C
ε	0	8	16	24	32	40	48	56	64	72	80
T	8	0	8	16	24	32	40	48	56	64	72
A	16	8	0	8	16	24	32	40	48	56	64
C	24	16	8	2	10	18	24	32	40	48	56
G	32	24	16	10	2	10	18	26	34	40	48
T	40	32	24	16	10	2	10	18	26	34	42
C	48	40	32	24	18	10	2	10	18	26	34
A	56	48	40	32	26	18	10	2	10	18	26
G	64	56	48	40	32	26	18	10	6	10	18
C	72	64	56	48	40	34	26	18	12	10	10

	A	C	G	T	-
A	0	4	2	4	8
C	4	0	4	2	8
G	2	4	0	4	8
T	4	2	4	0	8
-	8	8	8	8	

T A C G T C A - G C
 | | | | | | |
 T A T G T C A T G C
 +2 +8
 (transition) (gap)

Dynamic Programming: Similarity

Similarity changes the scoring function but also our algorithm

	ε	T	A	T
ε	0	8	16	24
T	8	0	8	16
A	16	8	0	8
C	24	16	8	2

	ε	T	A	T
ε	0	8	16	24
T	8			
A	16			
C	24			

	A	C	G	T	-
A	0	4	2	4	8
C	4	0	4	2	8
G	2	4	0	4	8
T	4	2	4	0	8
-	8	8	8	8	

	A	C	G	T	-
A	1	-3	-1	-3	-7
C	-3	1	-3	-1	-7
G	-1	-3	1	-3	-7
T	-3	-1	-3	1	-7
-	-7	-7	-7	-7	

Dynamic Programming: Similarity

Similarity changes the scoring function but also our algorithm

	ϵ	T	A	T
ϵ	0	8	16	24
T	8	0	8	16
A	16	8	0	8
C	24	16	8	2

	ϵ	T	A	T
ϵ	0	-7	-14	-21
T	-7	1	-6	-13
A	-14	-6	2	-5
C	-21	-13	-5	1

	A	C	G	T	-
A	0	4	2	4	8
C	4	0	4	2	8
G	2	4	0	4	8
T	4	2	4	0	8
-	8	8	8	8	

	A	C	G	T	-
A	1	-3	-1	-3	-7
C	-3	1	-3	-1	-7
G	-1	-3	1	-3	-7
T	-3	-1	-3	1	-7
-	-7	-7	-7	-7	

Dynamic Programming: Similarity



Similarity changes the scoring function but also our algorithm

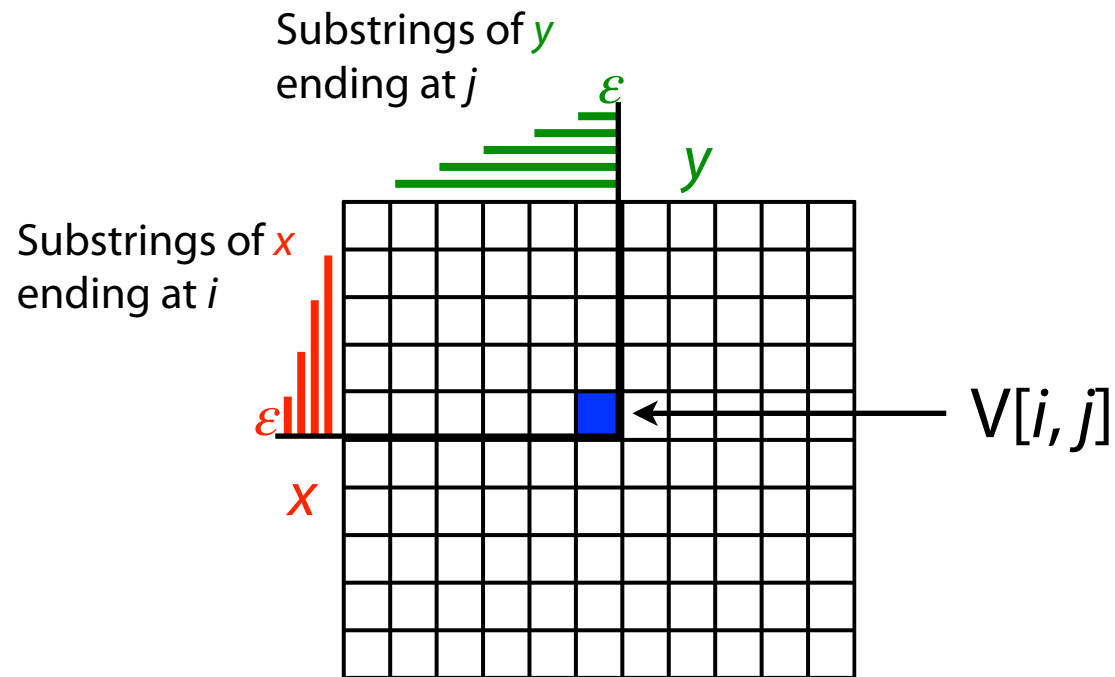
Given a scoring function s :

$$D[i, j] = \min \begin{cases} D[i - 1, j] + s(x[i - 1], -) \\ D[i, j - 1] + s(-, y[j - 1]) \\ D[i - 1, j - 1] + s(x[i - 1], y[j - 1]) \end{cases}$$

$$V[i, j] = \max \begin{cases} V[i - 1, j] + s(x[i - 1], -) \\ V[i, j - 1] + s(-, y[j - 1]) \\ V[i - 1, j - 1] + s(x[i - 1], y[j - 1]) \end{cases}$$

Dynamic Programming: Any position start

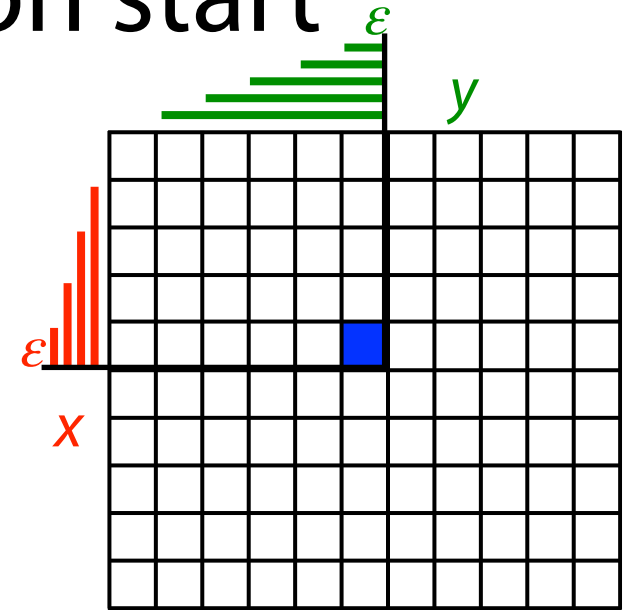
Let $V[i, j]$ be the optimal global alignment among substrings of x ending at i and substrings of y ending at j . The substrings may be empty.



How do we calculate $V[i, j]$?

Dynamic Programming: Any position start

Let $V[i, j]$ be the optimal global alignment among substrings of x ending at i and substrings of y ending at j . The substrings may be empty.



There are only so many possibilities:

Empty: let both substrings be empty, global alignment value = 0

Vertical: append **D** to transcript for $V[i-1, j]$, add penalty

Horizontal: append **I** to transcript for $V[i, j-1]$, add penalty

Diagonal: append **M** or **R** to transcript for $V[i-1, j-1]$, add match bonus or replacement penalty as appropriate

Dynamic Programming: Any position start

Let $V[0, j] = 0$, and let $V[i, 0] = 0$

$$\text{Otherwise, let } V[i, j] = \max \begin{cases} V[i - 1, j] + s(x[i - 1], -) \\ V[i, j - 1] + s(-, y[j - 1]) \\ V[i - 1, j - 1] + s(x[i - 1], y[j - 1]) \\ 0 \end{cases}$$

	ϵ	T	C	A	G
ϵ	0	0	0	0	0
C	0				
A	0				
C	0				

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

Dynamic Programming: Any position start

Let $V[0, j] = 0$, and let $V[i, 0] = 0$

Otherwise, let $V[i, j] = \max \left\{ \begin{array}{l} V[i - 1, j] + s(x[i - 1], -) \\ V[i, j - 1] + s(-, y[j - 1]) \\ V[i - 1, j - 1] + s(x[i - 1], y[j - 1]) \\ 0 \end{array} \right.$

	ϵ	T	C	A	G
ϵ	0	0	0	0	0
C	0	0	1	0	0
A	0	0	0	2	1
C	0	0	1	1	1

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

Local Alignment: Putting it all together



A local alignment scoring function must:

Assign negative weights to edits

Assign positive weights to matches

Not shown: There are constraints on what our scores can be

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

Local Alignment dynamic programming adds a fourth option to each step:

$$V[i, j] = \max \begin{cases} V[i - 1, j] + s(x[i - 1], -) \\ V[i, j - 1] + s(-, y[j - 1]) \\ V[i - 1, j - 1] + s(x[i - 1], y[j - 1]) \\ 0 \end{cases}$$

Local Alignment: Example

	e	T	A	T	A	T	G	C	G	G	C	G	T
e	0	0	0	0	0	0	0	0	0	0	0	0	0
G	0												
G	0												
T	0												
A	0												
T	0												
G	0												
C	0												
T	0												
G	0												
G	0												
C	0												
G	0												

$s(a, b)$

	A	C	G	T	-
A	2	-4	-4	-4	-6
C	-4	2	-4	-4	-6
G	-4	-4	2	-4	-6
T	-4	-4	-4	2	-6
-	-6	-6	-6	-6	

Local Alignment: Example

	e	T	A	T	A	T	G	C	G	G	C	G	T
e	0	0	0	0	0	0	0	0	0	0	0	0	0
G	0	0	0	0	0	0	2	0	2	2	0	2	0
G	0	0	0	0	0	0	2	0	2	4	0	2	0
T	0	2	0	2	0	2	0	0	0	0	0	0	4
A	0	0	4	0									
T	0												
G	0												
C	0												
T	0												
G	0												
G	0												
C	0												
G	0												

$$s(a, b)$$

	A	C	G	T	-
A	2	-4	-4	-4	-6
C	-4	2	-4	-4	-6
G	-4	-4	2	-4	-6
T	-4	-4	-4	2	-6
-	-6	-6	-6	-6	

Local Alignment: Example

	ε	T	A	T	A	T	G	C	G	G	C	G	T
ε	0	0	0	0	0	0	0	0	0	0	0	0	0
G	0	0	0	0	0	0	2	0	2	2	0	2	0
G	0	0	0	0	0	0	2	0	2	4	0	2	0
T	0	2	0	2	0	2	0	0	0	0	0	0	4
A	0	0	4	0	4	0	0	0	0	0	0	0	0
T	0	2	0	6	0	6	0	0	0	0	0	0	2
G	0	0	0	0	2	0	8	2	2	2	0	2	0
C	0	0	0	0	0	0	2	10	4	0	4	0	0
T	0	2	0	2	0	2	0	4	6	0	0	0	2
G	0	0	0	0	0	0	4	0	6	8	2	2	0
G	0	0	0	0	0	0	2	0	2	8	4	4	0
C	0	0	0	0	0	0	0	4	0	2	10	4	0
G	0	0	0	0	0	0	2	0	6	2	4	12	6

$s(a, b)$

	A	C	G	T	-
A	2	-4	-4	-4	-6
C	-4	2	-4	-4	-6
G	-4	-4	2	-4	-6
T	-4	-4	-4	2	-6
-	-6	-6	-6	-6	

Local Alignment: Example

	ε	T	A	T	A	T	G	C	G	G	C	G	T
ε	0	0	0	0	0	0	0	0	0	0	0	0	0
G	0	0	0	0	0	0	2	0	2	2	0	2	0
G	0	0	0	0	0	0	2	0	2	4	0	2	0
T	0	2	0	2	0	2	0	0	0	0	0	0	4
A	0	0	4	0	4	0	0	0	0	0	0	0	0
T	0	2	0	6	0	6	0	0	0	0	0	0	2
G	0	0	0	0	2	0	8	2	2	2	0	2	0
C	0	0	0	0	0	0	2	10	4	0	4	0	0
T	0	2	0	2	0	2	0	4	6	0	0	0	2
G	0	0	0	0	0	0	4	0	6	8	2	2	0
G	0	0	0	0	0	0	2	0	2	8	4	4	0
C	0	0	0	0	0	0	0	4	0	2	10	4	0
G	0	0	0	0	0	0	2	0	6	2	4	12	6

Where is my best match?

When do I stop?

Local Alignment: Example

Traceback from largest value

Stop when 0

	ε	T	A	T	A	T	G	C	G	G	C	G	T
ε	0	0	0	0	0	0	0	0	0	0	0	0	0
G	0	0	0	0	0	0	2	0	2	2	0	2	0
G	0	0	0	0	0	0	2	0	2	4	0	2	0
T	0	2	0	2	0	2	0	0	0	0	0	0	4
A	0	0	4	0	4	0	0	0	0	0	0	0	0
T	0	2	0	6	0	4	0	0	0	0	0	0	2
G	0	0	0	0	2	0	2	2	2	2	0	2	0
C	0	0	0	0	0	0	2	4	4	0	4	0	0
T	0	2	0	2	0	2	0	4	6	0	0	0	0
G	0	0	0	0	0	0	4	0	6	8	2	2	0
G	0	0	0	0	0	0	2	0	2	4	4	4	0
C	0	0	0	0	0	0	0	4	0	2	4	4	0
G	0	0	0	0	0	0	2	0	6	2	4	6	6

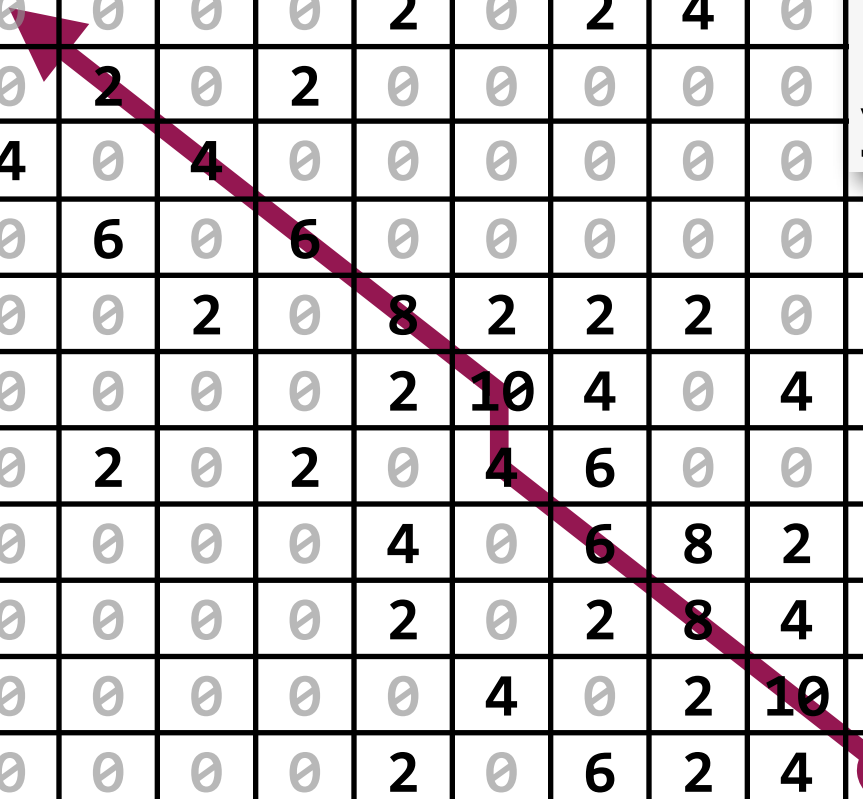
```

x : G G T A T G C T G G C G C T A
      | | | | | | | |
y : T A T A T G C - G G C G T T T
  
```



	ε	T	A	T	A	T	G	C	G	G	C	G	T	T	T
ε	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
G	0	0	0	0	0	0	2	0	2	2	0	0	0	0	0
G	0	0	0	0	0	0	2	0	2	4	0	0	0	0	0
T	0	2	0	2	0	2	0	0	0	0	0	0	0	0	0
A	0	0	4	0	4	0	0	0	0	0	0	0	0	0	0
T	0	2	0	6	0	6	0	0	0	0	0	0	2	2	2
G	0	0	0	0	2	0	8	2	2	2	0	2	0	0	0
C	0	0	0	0	0	0	2	10	4	0	4	0	0	0	0
T	0	2	0	2	0	2	0	4	6	0	0	0	2	2	2
G	0	0	0	0	0	0	4	0	6	8	2	2	0	0	0
G	0	0	0	0	0	0	2	0	2	8	4	4	0	0	0
C	0	0	0	0	0	0	0	4	0	2	10	4	0	0	0
G	0	0	0	0	0	0	2	0	6	2	4	12	6	0	0
C	0	0	0	0	0	0	0	4	0	2	4	6	8	2	0
T	0	2	0	2	0	2	0	0	0	0	0	0	8	10	4
A	0	0	4	0	4	0	0	0	0	0	0	0	2	4	6

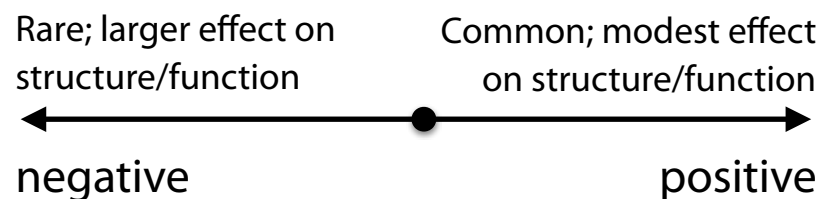
x : G G T A T G C T G G C G C T A
 | | | | | | | | | |
y : T A T A T G C - G G C G T T T



BLOSUM62

Some amino acid substitutions have a smaller impact on structure & function than others. BLOSUM62 elements are, roughly speaking, log-odds of observing these substitutions between two related proteins

Ala	4																			
Arg	-1	5																		
Asn	-2	0	6																	
Asp	-2	-2	1	6																
Cys	0	-3	-3	-3	9															
Gln	-1	1	0	0	-3	5														
Glu	-1	0	0	2	-4	2	5													
Gly	0	-2	0	-1	-3	-2	-2	6												
His	-2	0	1	-1	-3	0	0	-2	8											
Ile	-1	-3	-3	-3	-1	-3	-3	-4	-3	4										
Leu	-1	-2	-3	-4	-1	-2	-3	-4	-3	2	4									
Lys	-1	2	0	-1	-3	1	1	-2	-1	-3	-2	5								
Met	-1	-1	-2	-3	-1	0	-2	-3	-2	1	2	-1	5							
Phe	-2	-3	-3	-3	-2	-3	-3	-3	-1	0	0	-3	0	6						
Pro	-1	-2	-2	-1	-3	-1	-1	-2	-2	-3	-3	-1	-2	-4	7					
Ser	1	-1	1	0	-1	0	0	0	-1	-2	-2	0	-1	-2	-1	4				
Thr	0	-1	0	-1	-1	-1	-1	-2	-2	-1	-1	-1	-1	-2	-1	1	5			
Trp	-3	-3	-4	-4	-2	-2	-3	-2	-2	-3	-2	-3	-1	1	-4	-3	-2	11		
Tyr	-2	-2	-2	-3	-2	-1	-2	-3	2	-1	-1	-2	-1	3	-3	-2	-2	2	7	
Val	0	-3	-3	-3	-1	-2	-2	-3	-3	3	1	-2	1	-1	-2	-2	0	-3	-1	4
	Ala	Arg	Asn	Asp	Cys	Gln	Glu	Gly	His	Ile	Leu	Lys	Met	Phe	Pro	Ser	Thr	Trp	Tyr	Val



Matrix is symmetric

Amino acids