Chapter 7: Rapid alignment methods: FASTA and BLAST
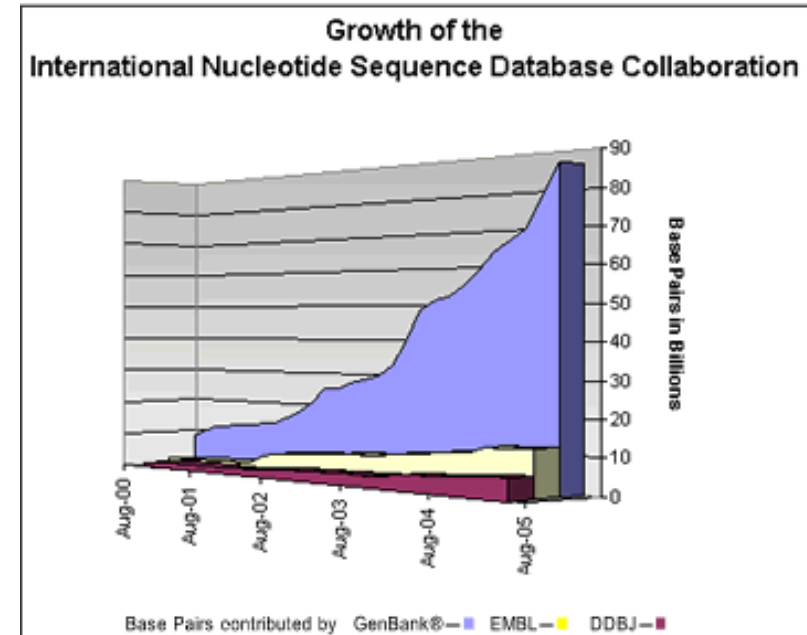
- *The biological problem*

- *Search strategies*

- FASTA

- BLAST

The biological problem

- Global and local alignment algoritms are slow in practice
- Consider the scenario of aligning a *query sequence* against a large database of sequences
  - New sequence with unknown function

- For instance, the size of NCBI GenBank in January 2007 was 65,369,091,950 bases (61,132,599 sequences)
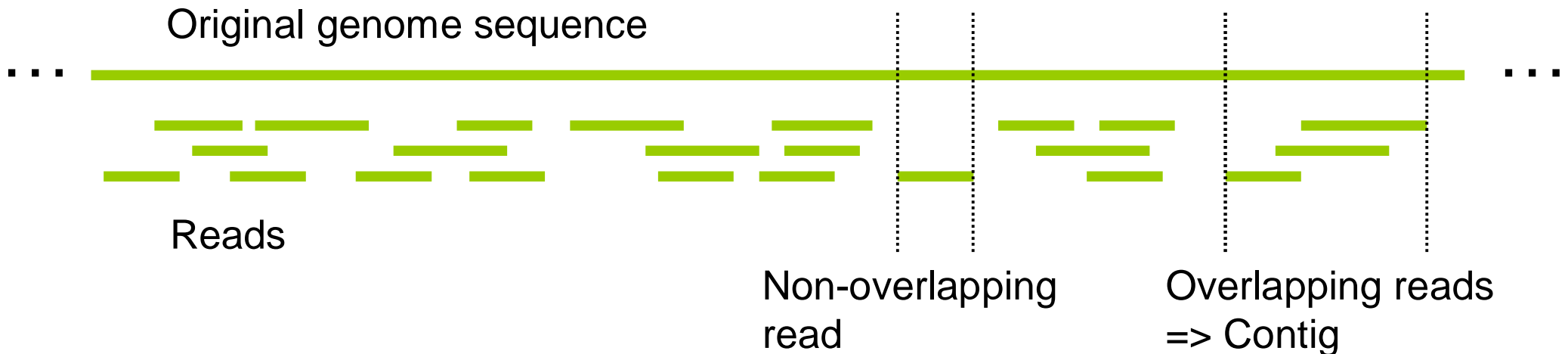
Problem with large amount of sequences

- Exponential growth in both number and total length of sequences

- Possible solution: Compare against model organisms only

- With large amount of sequences, changes are that matches occur by random

  - Need for statistical analysis
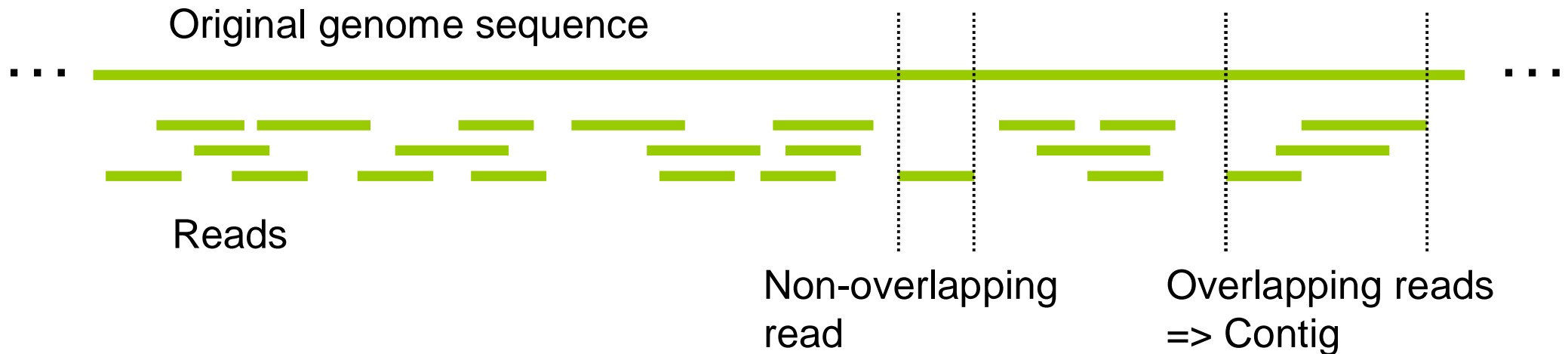
Application of sequence alignment: shotgun sequencing

- Shotgun sequencing is a method for sequencing whole-organism genomes

  - First, a large number of short sequences (~500-1000 bp), or *reads* are generated from the genome

  - Reads are contiguous subsequences (substrings) of the genome

  - Due to sequencing errors and repetitions in the reads, the genome has be covered multiple times by reads

Shotgun sequencing

Original genome sequence

... ——————————————————————————————————— ...

Reads

Non-overlapping read

Overlapping reads => Contig

- Ordering of the reads is initially unknown

- Overlaps resolved by aligning the reads

- In a 3×10^9 bp genome with 500 bp reads and 5x coverage, there are $\sim 10^7$ reads and $\sim 10^7(10^7-1)/2 = \sim 5 \times 10^{13}$ pairwise sequence comparisons

Shotgun sequencing

Original genome sequence

... ———————————————————— ...

Reads

Non-overlapping read

Overlapping reads => Contig

- ~5×10^{13} pairwise sequence comparisons

- Recall that local alignment takes $O(nm)$ time, where n and m are sequence lengths

- Already with n=m=500, the computation cost is prohibitive

Search strategies

- How to speed up the computation?

  - Find ways to limit the number of pairwise comparisons

- Compare the sequences at word level to find out common words

  - Word means here a k-tuple (or a k-word), a substring of length k

Analyzing the word content

- Example query string I: TGATGATGAAGACATCAG

- For k = 8, the set of k-tuples of I is

```
TGATGATG

  GATGATGA

    ATGATGAA

      TGATGAAG

          ...

              GACATCAG
```

Analyzing the word content

- There are n-k+1 k-tuples in a string of length n

- If at least one word of I is not found from another string J, we know that I differs from J

- Need to consider statistical significance: I and J might share words by chance only

- Let n=|I| and m=|J|

Word lists and comparison by content

- The k-words of I can be arranged into a table of word occurences $L_w(I)$

- Consider the k-words when k=2 and I=GCATCGGC:

GC, CA, AT, TC, CG, GG, GC

AT: 3
CA: 2
CG: 5
GC: 1, 7 $\longleftarrow$ Start indecies of k-word GC in I
GG: 6
TC: 4

Building $L_w(I)$ takes O(n) time

Common k-words

l Number of common k-words in I and J can be computed using $L_w(I)$ and $L_w(J)$

l For each word w in I, there are $|L_w(J)|$ occurences in J

l Therefore I and J have $\sum_w |L_w(I)||L_w(J)|$

common words

l This can be computed in $O(n + m + 4^k)$ time

- $O(n + m)$ time to build the lists

- $O(4^k)$ time to calculate the sum

Common k-words

- I = GCATCGGC

- J = CCATCGCCATCG

$L_w(I)$	$L_w(J)$	Common words
AT: 3	AT: 3, 9	2
CA: 2	CA: 2, 8	2
	CC: 1, 7	0
CG: 5	CG: 5, 11	2
GC: 1, 7	GC: 6	2
GG: 6		0
TC: 4	TC: 4, 10	2
		10 in total

Properties of the common word list

l Exact matches can be found using binary search (e.g., where TCGT occurs in I?)

- O($\log 4^k$) time

l For large k, the table size is too large to compute the common word count in the previous fashion

l Instead, an approach based on merge sort can be utilised (details skipped, see course book)

l The common k-word technique can be combined with the local alignment algorithm to yield a rapid alignment approach

Chapter 7: Rapid alignment methods: FASTA and BLAST

l The biological problem

l Search strategies

l *FASTA*

l BLAST

FASTA
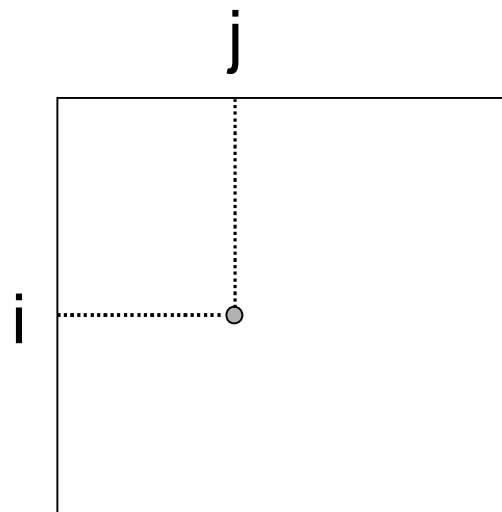
l FASTA is a multistep algorithm for sequence alignment (Wilbur and Lipman, 1983)

l The sequence file format used by the FASTA software is widely used by other sequence analysis software

l Main idea:

  – Choose regions of the two sequences that look promising (have some degree of similarity)

  – Compute local alignment using dynamic programming in these regions

FASTA outline

- FASTA algorithm has five steps:

  - 1. Identify common k-words between I and J

  - *2. Score diagonals with k-word matches, identify 10 best diagonals*

  - 3. Rescore initial regions with a substitution score matrix

  - 4. Join initial regions using gaps, penalise for gaps

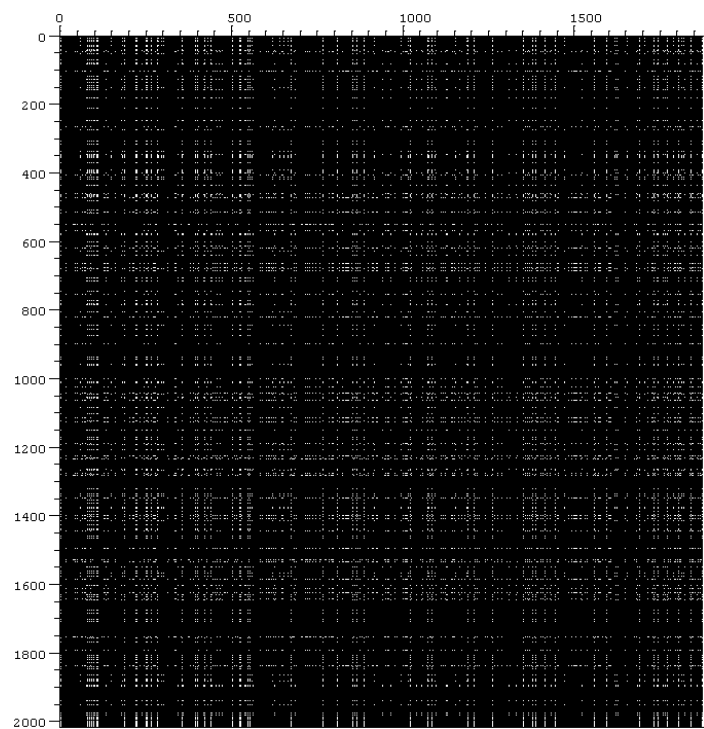 - 5. Perform dynamic programming to find final alignments

Dot matrix comparisons

- Word matches in two sequences I and J can be represented as a *dot matrix*

- Dot matrix element (i, j) has "a dot", if the word starting at position i in I is identical to the word starting at position j in J

- The dot matrix can be plotted for various k
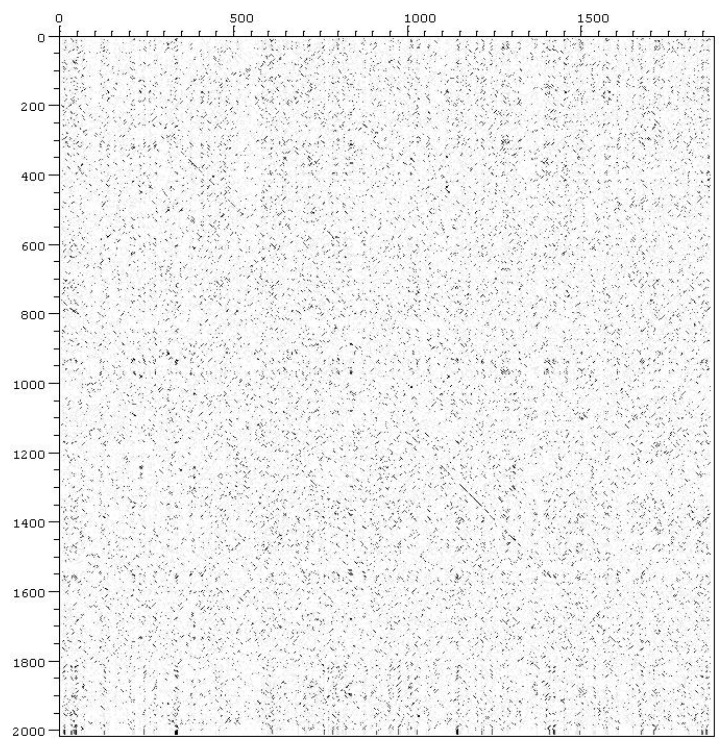
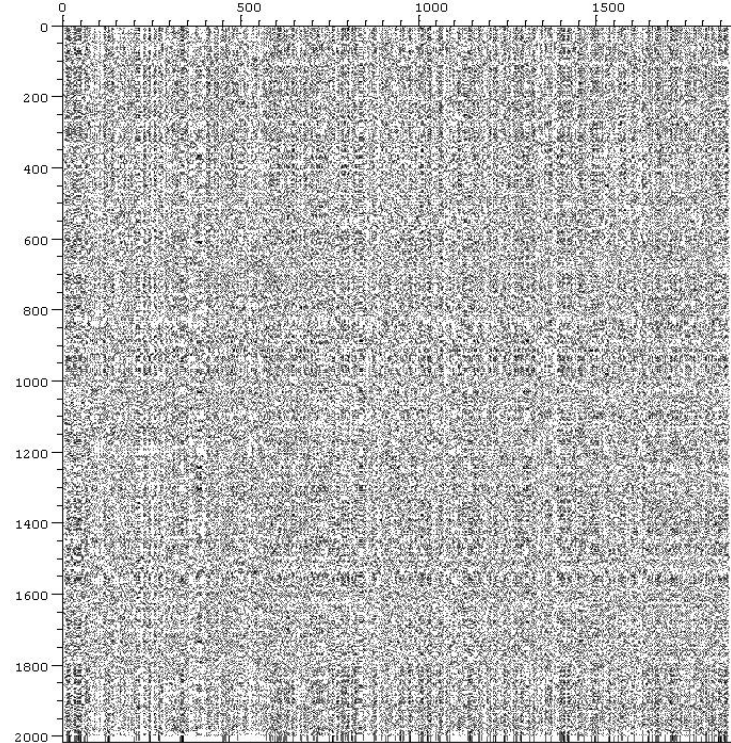```
I = … ATCGGATCA …
J = … TGGTGTCGC …
```

k=1

k=4

k=8
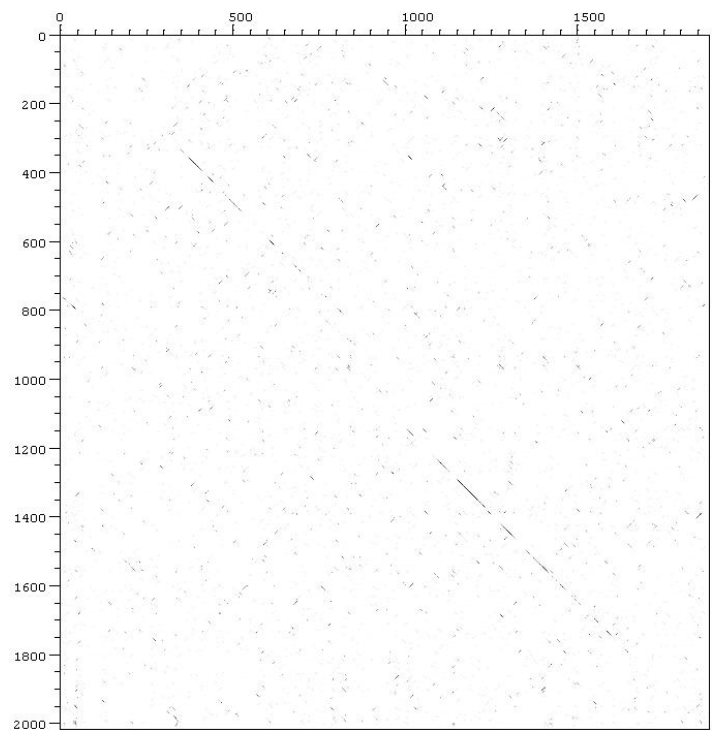
k=16

Dot matrix (k=1,4,8,16)
for two **DNA** sequences
X85973.1 (1875 bp)
Y11931.1 (2013 bp)

Introduction to bioinformatics, Autumn 2007

k=1

k=4

Dot matrix
(k=1,4,8,16) for two
**protein** sequences
CAB51201.1 (531 aa)
CAA72681.1 (588 aa)

k=8

k=16

Shading indicates
now the match score
according to a
score matrix
(Blosum62 here)

Introduction to bioinformatics, Autumn 2007

# Computing diagonal sums

l We would like to find high scoring diagonals of the dot matrix

l Lets index diagonals by the offset, $l = i - j$



Diagonal $l = i - j = -6$

# Computing diagonal sums
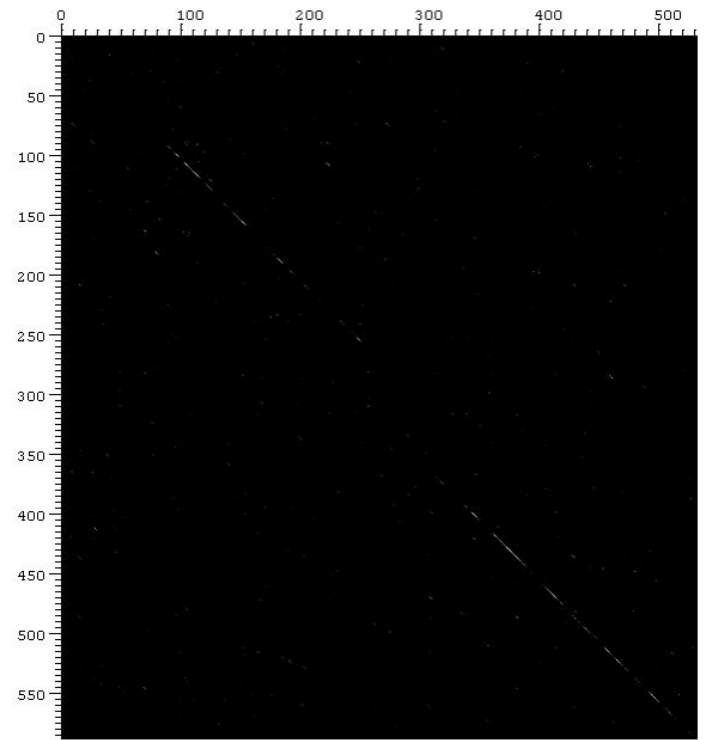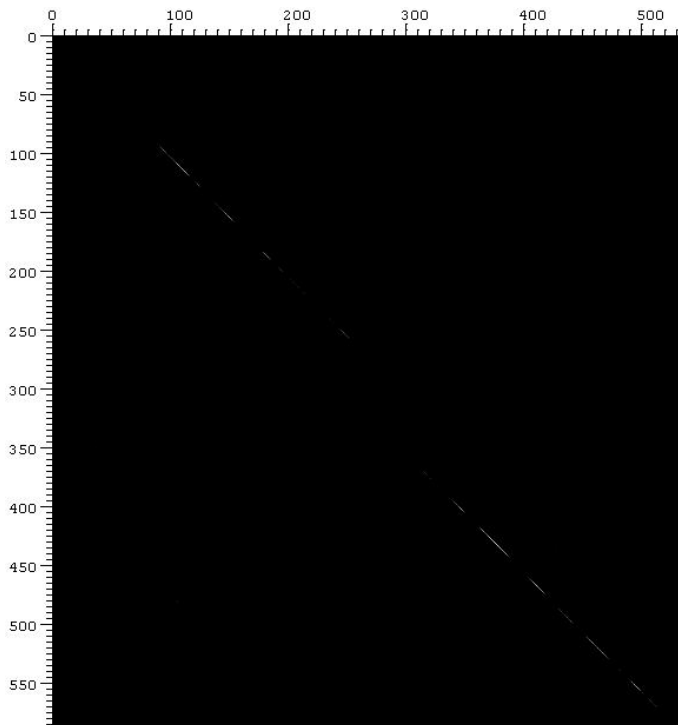
- As an example, lets compute diagonal sums for  I = GCATCGGC, J = CCATCGCCATCG, k = 2

- 1. Construct k-word list $L_w(J)$

- 2. Diagonal sums $S_l$ are computed into a table, indexed with the offset and initialised to zero

```
 l │ -10  -9  -8  -7  -6  -5  -4  -3  -2  -1  0  1  2  3  4  5  6
───┼───────────────────────────────────────────────────────────
S_l│   0   0   0   0   0   0   0   0   0   0  0  0  0  0  0  0  0
```

Computing diagonal sums

3. Go through k-words of I, look for matches in $L_w(J)$ and update diagonal sums

J

	C	C	A	T	C	G	C	C	A	T	C	G
G						*						
C		*						*				
A			*						*			
T				*						*		
C					*						*	
G												
G						*						
C												

I

For the first 2-word in I, GC, $L_{GC}(J) = \{6\}$.

We can then update the sum of diagonal $l = i - j = 1 - 6 = -5$ to $S_{-5} := S_{-5} + 1 = 0 + 1 = 1$

Computing diagonal sums

l 3. Go through k-words of I, look for matches in $L_w(J)$ and update diagonal sums

J

	C	C	A	T	C	G	C	C	A	T	C	G
G						*						
C		*						*				
A			*						*			
T				*						*		
C					*						*	
G												
G					*							
C												

I

Next 2-word in I is CA, for which $L_{CA}(J) = \{2, 8\}$.

Two diagonal sums are updated:
$l = i - j = 2 - 2 = 0$
$S_0 := S_0 + 1 = 0 + 1 = 1$

$l = i - j = 2 - 8 = -6$
$S_{-6} := S_{-6} + 1 = 0 + 1 = 1$

Computing diagonal sums

- 3. Go through k-words of I, look for matches in $L_w(J)$ and update diagonal sums

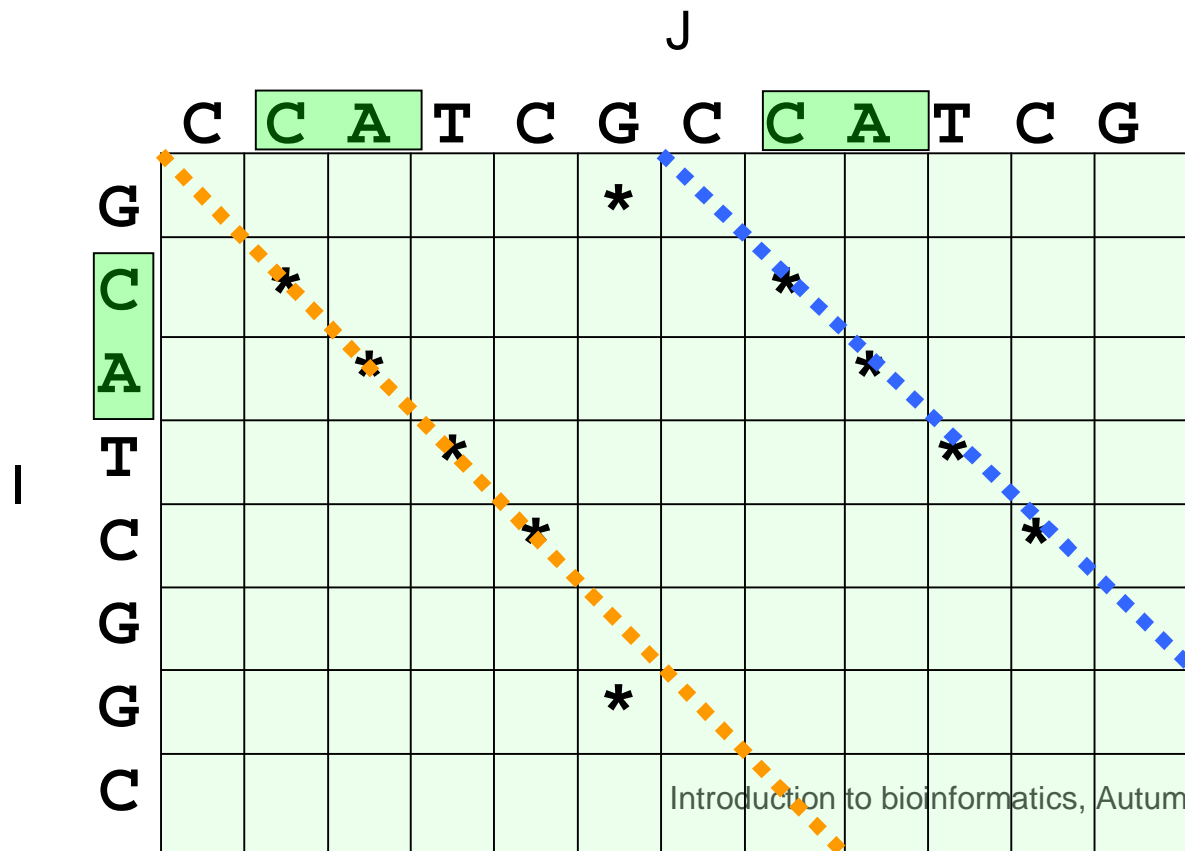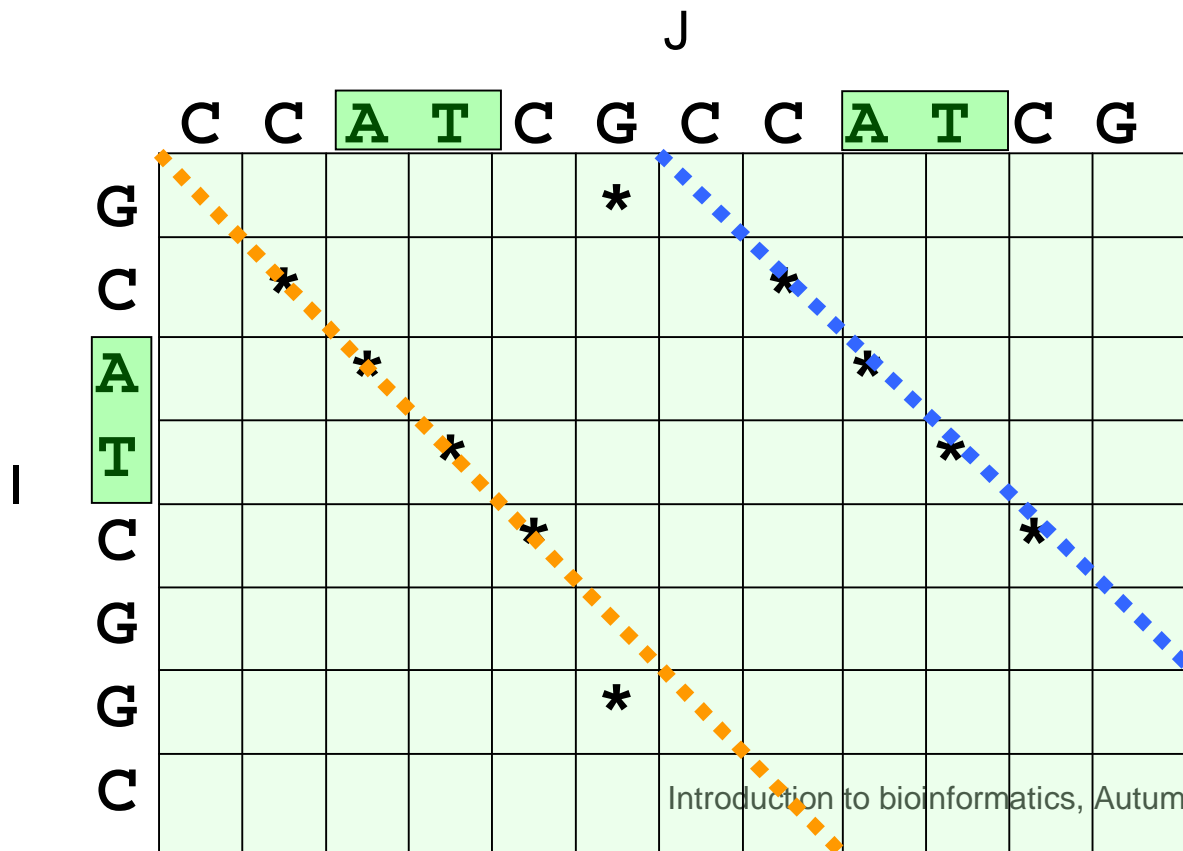Next 2-word in I is AT, for which $L_{AT}(J) = \{3, 9\}$.

Two diagonal sums are updated:

$l = i - j = 3 - 3 = 0$

$S_0 := S_0 + 1 = 1 + 1 = 2$

$l = i - j = 3 - 9 = -6$

$S_{-6} := S_{-6} + 1 = 1 + 1 = 2$

Computing diagonal sums

After going through the k-words of I, the result is:

l	-10	-9	-8	-7	-6	-5	-4	-3	-2	-1	0	1	2	3	4	5	6
S_l	0	0	0	0	4	1	0	0	0	0	4	1	0	0	0	0	0

J

	C	C	A	T	C	G	C	C	A	T	C	G
G						*						
C		*						*				
A			*						*			
T				*						*		
C					*						*	
G												
G						*						
C												

I

Algorithm for computing diagonal sum of scores

$S_l := 0$ for all $1 - m \leq l \leq n - 1$

Compute $L_w(J)$ for all words w

for $i := 1$ to $n - k - 1$ do

    $w := l_i l_{i+1} \ldots l_{i+k-1}$

    for $j \in L_w(J)$ do

        $l := i - j$

        $S_l := S_l + 1$ ←——————   Match score is here 1

    end

end

FASTA outline

- FASTA algorithm has five steps:
  - 1. Identify common k-words between I and J
  - 2. Score diagonals with k-word matches, identify 10 best diagonals
  - *3. Rescore initial regions with a substitution score matrix*
  - *4. Join initial regions using gaps, penalise for gaps*
  - 5. Perform dynamic programming to find final alignments

Rescoring initial regions

l Each high-scoring diagonal chosen in the previous step is rescored according to a score matrix

l This is done to find subregions with identities shorter than k

l Non-matching ends of the diagonal are trimmed

```
I:  C  C  A  T  C  G  C  C  A  T  C  G
J:  C  C  A  A  C  G  C  A  A  T  C  A
```
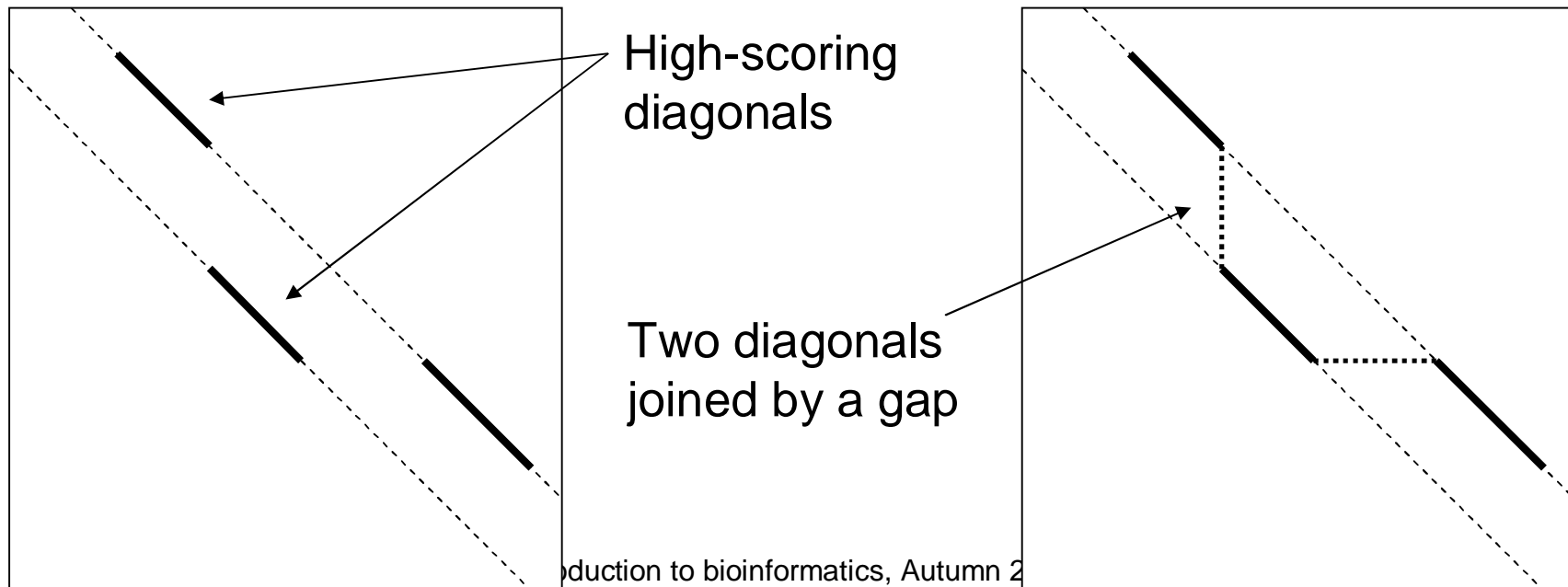75% identity, no 4-word identities

```
I': C  C  A  T  C  G  C  C  A  T  C  G
J': A  C  A  T  C  A  A  A  T  A  A  A
```
33% identity, one 4-word identity

Joining diagonals

- Two offset diagonals can be joined with a gap, if the resulting alignment has a higher score

- Separate gap open and extension are used

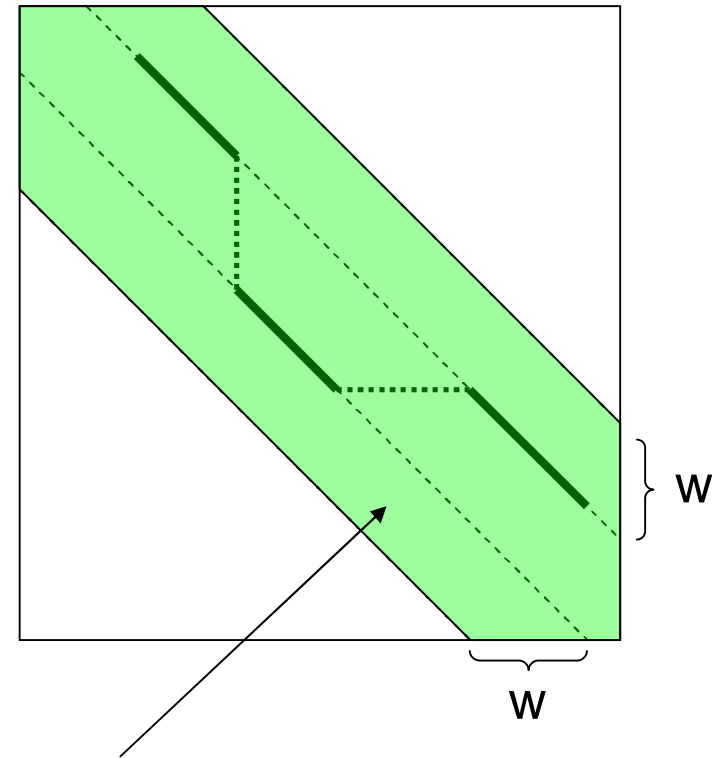- Find the best-scoring combination of diagonals

High-scoring diagonals

Two diagonals joined by a gap

FASTA outline

- FASTA algorithm has five steps:

  - 1. Identify common k-words between I and J

  - 2. Score diagonals with k-word matches, identify 10 best diagonals

  - 3. Rescore initial regions with a substitution score matrix

  - 4. Join initial regions using gaps, penalise for gaps

  - *5. Perform dynamic programming to find final alignments*

Local alignment in the highest-scoring region

- Last step of FASTA: perform local alignment using dynamic programming around the highest-scoring

- Region to be aligned covers –w and +w offset diagonal to the highest-scoring diagonals

- With long sequences, this region is typically very small compared to the whole n x m matrix

w

w

Dynamic programming matrix M filled only for the green region

Properties of FASTA

- Fast compared to local alignment using dynamic programming only

  - Only a narrow region of the full matrix is aligned

- Increasing parameter k decreases the number of hits: increases specificity, decreases sensitivity

- FASTA can be very specific when identifying long regions of low similarity

  - Specific method does not produce many incorrect results

  - Sensitive method produces many of the correct results

Properties of FASTA

- FASTA looks for initial exact matches to query sequence

  - Two proteins can have very different amino acid sequences and still be biologically similar

  - This may lead into a lack of sensitivity with diverged sequences

Demonstration of FASTA at EBI

- http://www.ebi.ac.uk/fasta/

- Note that parameter ktup in the software corresponds to parameter k in lectures