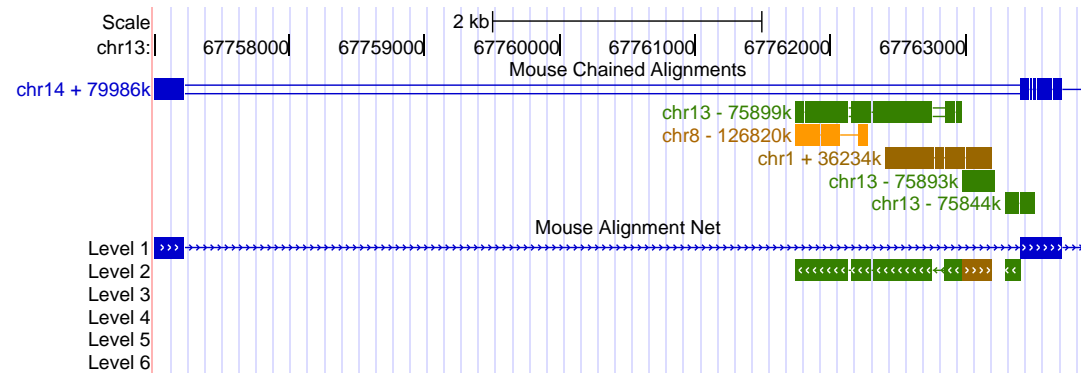


Announcements

- Submit your preferences for journal club papers using the form at the website until next Wednesday, Oct. 20 22:00
- Homework 1 will be published on the website, submit until Tuesday November 9 22:00 (pdf via Moodle, guests by e-mail)
- You are allowed to discuss homework questions with classmates, but do not take notes during discussions and do not show your solutions to others. Everybody should write their homework submission independently, do not copy from classmates or other sources.
- Please use MS Teams for questions regarding homeworks, quizzes and the course in general.
- However, any questions involving your ideas about solving the questions should be sent privately to instructors by email.

Sequence alignment 2/2

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October 14 2021



Summary from the last lecture

- **Global and local alignment problem**

Input: sequences $X = x_1x_2 \dots x_n$ and $Y = y_1y_2 \dots y_m$.

Output: alignment of X and Y with the highest score
or alignment of **substrings** $x_i \dots x_j$ a $y_k \dots y_\ell$ with the highest score

- **Correct algorithms** using dynamic programming
- **Realistic scoring schemes**

We have dynamic programming, what else do we need?

Running time: $O(n^2)$ on two sequences of length n

How much is that in practice?

(simple implementation, random sequences, desktop computer)

n	time
100	0.0008s
1,000	0.08s
10,000	8s
100,000	13m (*)
1,000,000	22h (*)
10,000,000	3months (*)
100,000,000	25years (*)

We need a more efficient algorithm, particularly for comparative genomics

Memory: basic implementation $O(n^2)$, but can be done in $O(n)$

Heuristic alignment

- Trade sensitivity for speed (some alignments not found)
- Reduce the search to “promising” parts of the matrix

Heuristic local alignment

BLASTN [Altschul et al 1990], FASTA [Pearson 1988]

- Find short exact matches of length w (**seeds**)
- Extend hits along diagonals to ungapped alignments
- Connect alignments on nearby diagonals to gapped alignment
- Possibly optimize by dynamic programming

How to find short exact matches?

- Create a **dictionary** of short substrings of length w from the first sequence.
- Search for all substring from the second sequence in the dictionary

Exmple: CAGTCCTAGA vs CATGTCATA

Dictionary:

AG 2, 8

CA 1

CC 5

CT 6

GA 9

GT 3

TA 7

TC 4

Search for:

CA → 1

AT → -

TG → -

GT → 3

TC → 4

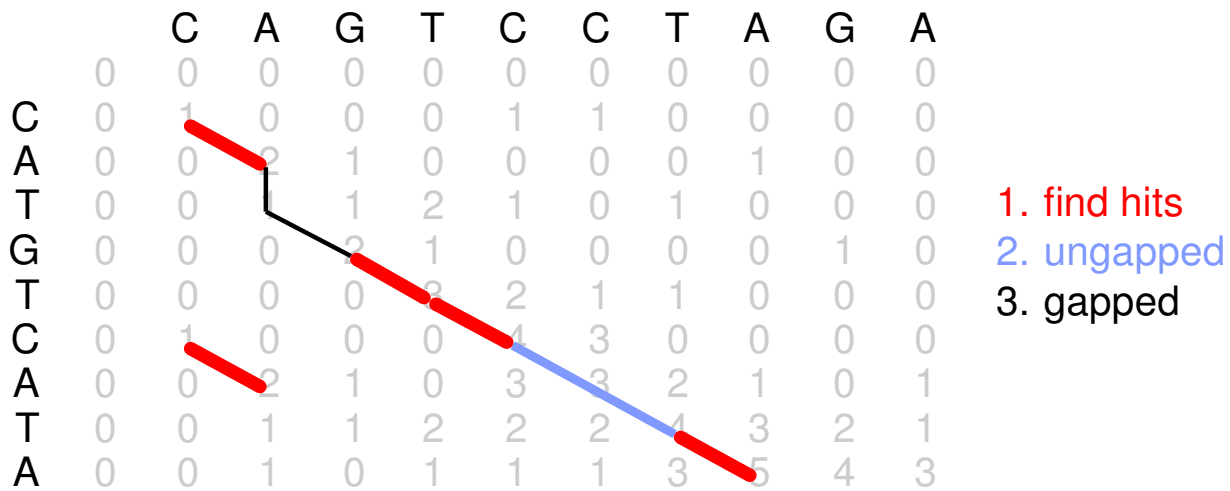
CA → 1

AT → -

TA → 7

Heuristic local alignment

Example: start from **seeds** of length $w = 2$
 (in practice we would use $w = 11$ or more)



Running time of heuristic local alignment

Algorithm

- Find seeds (short exact matches of length w)
- **Expensive step:** extend/connect seeds to longer alignments

Random seeds of length w : not part of any high-scoring alignment. These are filtered in the extension step, but they slow down the program

How many random hits?

Two unrelated nucleotides match with probability $1/4$

We have w matches in a row with probability 4^{-w}

Expected number of false positives roughly $nm4^{-w}$

Increase of w by 1 means cca 4-fold decrease of spurious seeds

Sensitivity of heuristic local alignment

Algorithm

- Find seeds (short exact matches of length w)
- **Expensive step:** extend/connect seeds to longer alignments

Some alignments not found: high score but **no seed of length w**

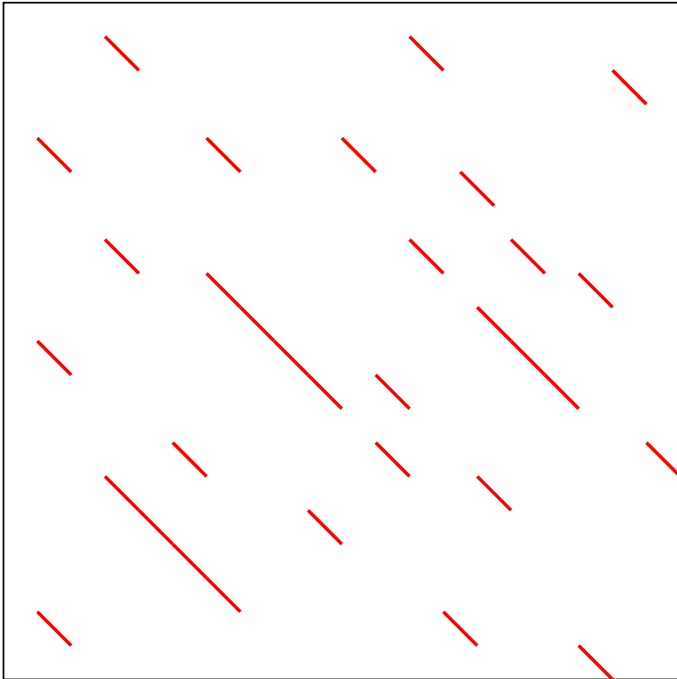
Example: CA-GTCCTA no seed of length $w \geq 4$
 CATGTCATA

Sensitivity: fraction of **real alignments** containing a seed of length w

Sensitivity vs. running time

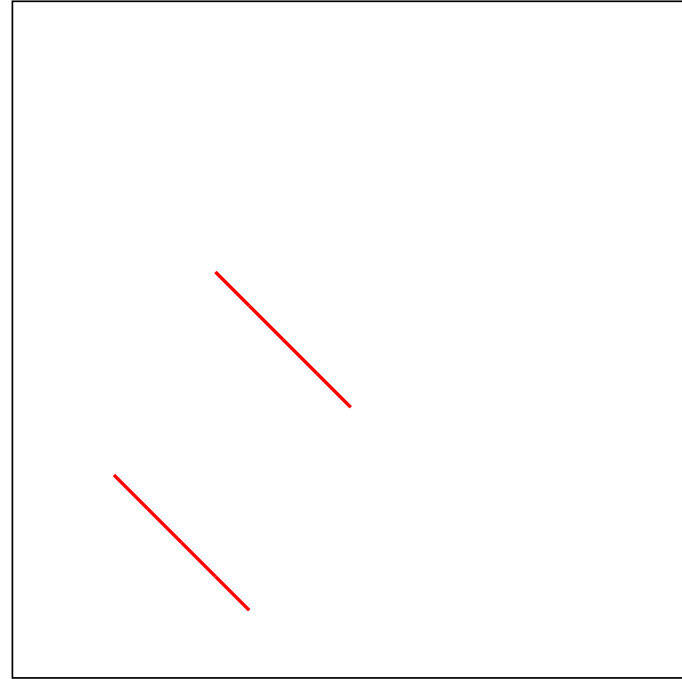
Small w

many spurious seeds, slow



Large w

many alignments not found

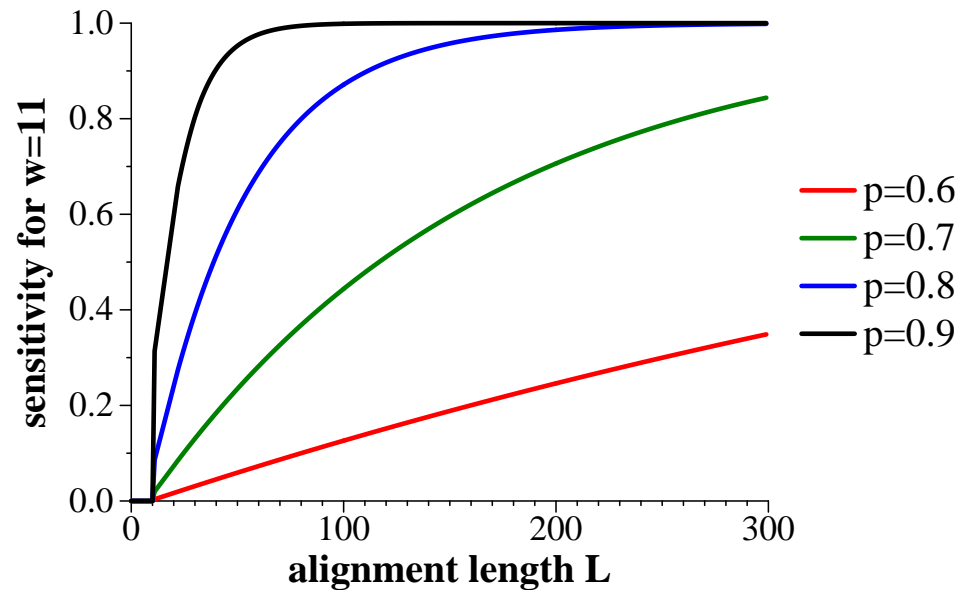


Can we estimate the sensitivity?

Assume random ungapped alignment of length L

Every position match with probability p

Sensitivity $f(L, p) = \Pr(\text{alignment contains } w \text{ consecutive matches})$



(human-mouse: $p \approx 0.7$)

Protein BLAST

BLOSUM62 scoring matrix for proteins

	A	R	N	D	C	Q	E	G	H	I	...
A	4	-1	-2	-2	0	-1	-1	0	-2	-1	
R	-1	5	0	-2	-3	1	0	-2	0	-3	
N	-2	0	6	1	-3	0	0	0	1	-3	
D	-2	-2	1	6	-3	0	2	-1	-1	-3	
C	0	-3	-3	-3	9	-3	-4	-3	-3	-1	
Q	-1	1	0	0	-3	5	2	-2	0	-3	
E	-1	0	0	2	-4	2	5	-2	0	-3	
G	0	-2	0	-1	-3	-2	-2	6	-2	-4	
H	-2	0	1	-1	-3	0	0	-2	8	-3	
I	-1	-3	-3	-3	-1	-3	-3	-4	-3	4	

Instead of exact match of length w , protein BLAST requires 3 amino acids with score at least 13

Hit: N I R
 N L R

 6+2+5=13

Not a hit: A I L
 A I L

 4+4+4=12

Examples of software tools for various tasks

NCBI BLAST: `blastn` for DNA/RNA, `blastp` for proteins, `tblastx` translates DNA to proteins and uses `blastp`

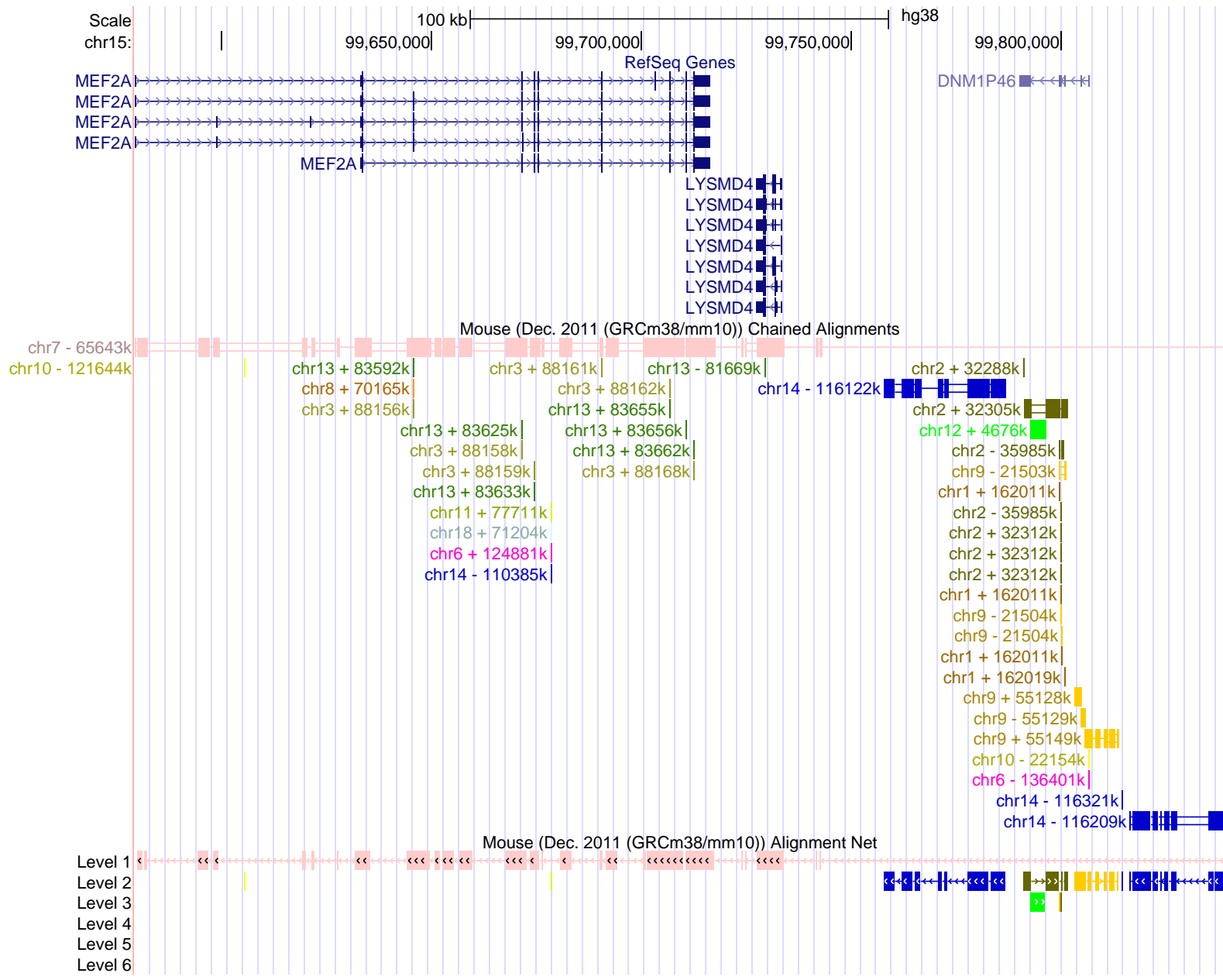
UCSC Blat: very fast search for very similar sequences, i.e. aligning sequencing reads to the genome

- uses very large values of w
- can split alignments with big gaps (aligning transcripts with introns)

Whole-genome alignments

For each section of human genome find closest section from mouse, dog, chicken, etc. (see e.g. UCSC genome browser)

- Local alignments will cover protein coding exons and other conserved parts
- Sections that diverged too much cannot be aligned
- If there was a duplication, we need to decide which pairs belong together
- **Synteny principle:** if two similar sections (local alignments) are present in the same order and orientation in two genomes, they likely evolved from the same common ancestor (orthologs)



Multiple sequence alignment

Running time: $O(2^k n^k)$ for k sequences of length n

For general k NP-hard.

```
Human  ctccatagcaatgt-cagagatagggcagagcggat-----ggtggtgac
Rhesus ctccatggcaatgt-cagagatagggcagagcggat-----gctggtgac
Mouse  ttt--tgacaaca--tagagac-tgagatagaaaat-----atgctgac
Dog    -tccccgctaatagtacaaagatggggcag-gaaga--a----tgtgctgaa
Horse  -tccacggcaatac-tggagatggggcagagcaga--agat-ggtgatgaa
Armadillo ctgcatagaaatct-cagagatgggggaaagcaga-----agacattcat
Opossum atccatggaaacat-cagaagtgggagaaatagaaga----tggcaatga-
Platypus acccggggaagggg-aagaggaaggggccggccg-----
```

Heuristic algorithms, e.g. CLUSTAL-W [Higgins et al., 1996], MUSCLE [Edgar, 2004] and TBA [Blanchette et al., 2004].

Sequences producing significant alignments:			Score (Bits)	E Value	
ref XP_002345317.1 	PREDICTED: similar to protein tyrosine ph...	28.2	108	UG	
ref XP_001726210.1 	PREDICTED: similar to protein tyrosine ph...	28.2	108	G	
ref ZP_03264973.1 	isocitrate dehydrogenase, NADP-dependent [...]	27.4	194		
ref XP_001225150.1 	hypothetical protein CHGG_07494 [Chaetomi...	27.4	194	G	
ref YP_002967336.1 	hypothetical protein MexAM1_META2p1254 [M...	26.9	261	G	
ref ZP_03013307.1 	hypothetical protein BACINT_00864 [Bactero...	26.9	261		
ref YP_001834672.1 	phospholipid/glycerol acyltransferase [Be...	26.9	261	G	
ref ZP_04426281.1 	NADH dehydrogenase subunit L [Planctomyces...	26.1	469		
ref YP_003129642.1 	putative exonuclease RecJ [Halorhabdus ut...	26.1	469	G	
ref ZP_02926313.1 	multidrug efflux pump, AcrB/AcrD/AcrF fami...	26.1	469		
ref ZP_02044690.1 	hypothetical protein ACTODO_01565 [Actinom...	26.1	469		
ref XP_001153320.1 	PREDICTED: similar to tyrosine phosphatas...	26.1	469	G	
ref YP_001958968.1 	inner-membrane translocator [Chlorobium p...	26.1	469	G	
ref YP_003133865.1 	hypothetical protein Svir_20200 [Saccharo...	25.7	630	G	

<http://blast.ncbi.nlm.nih.gov/Blast.cgi>

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Alignments
 Select All
 [Get selected sequences](#)
[Distance tree of results](#)
[Multiple alignment](#) NEW

[ref|XP_002345317.1|](#) **UG** PREDICTED: similar to protein tyrosine phosphatase 4a1 isoform 2 [Homo sapiens]
 Length=139

[GENE ID: 730167 LOC730167](#) | similar to protein tyrosine phosphatase 4a1 [Homo sapiens]

Score = 28.2 bits (59), Expect = 108
 Identities = 9/10 (90%), Positives = 10/10 (100%), Gaps = 0/10 (0%)

Query 1 VIVALASVEG 10
 V+VALASVEG
 Sbjct 79 VLVALASVEG 88

[ref|XP_001726210.1|](#) **G** PREDICTED: similar to protein tyrosine phosphatase 4a1 isoform 1 [Homo sapiens]
 Length=170

[GENE ID: 730167 LOC730167](#) | similar to protein tyrosine phosphatase 4a1 [Homo sapiens]

Score = 28.2 bits (59), Expect = 108
 Identities = 9/10 (90%), Positives = 10/10 (100%), Gaps = 0/10 (0%)

Query 1 VIVALASVEG 10
 V+VALASVEG
 Sbjct 110 VLVALASVEG 119

How to distinguish when the alignment is “real”?

Query length m . Database length n .

Alignment with score S .

P -value: Probability that a **random query** of length m in a **random database** of length n yields alignment of score at least S

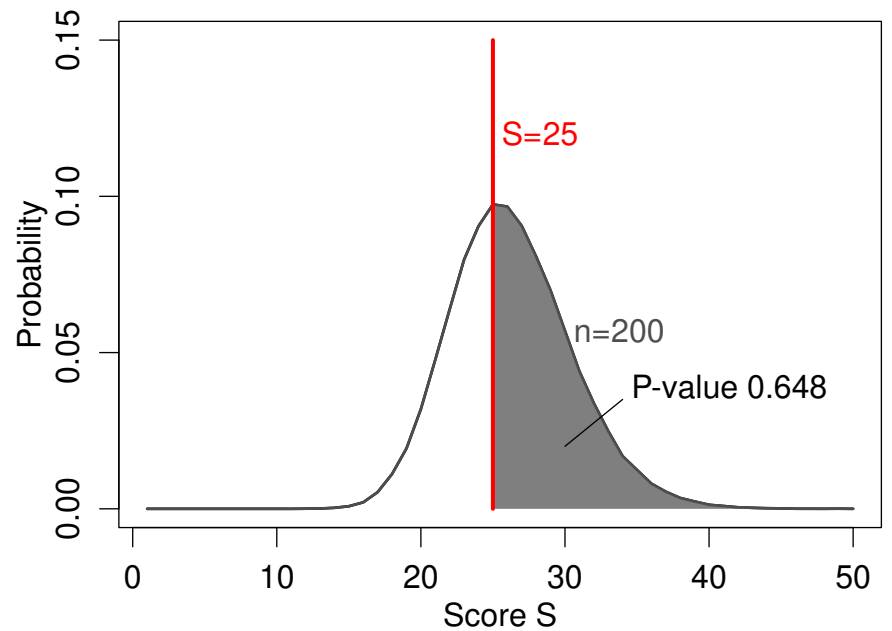
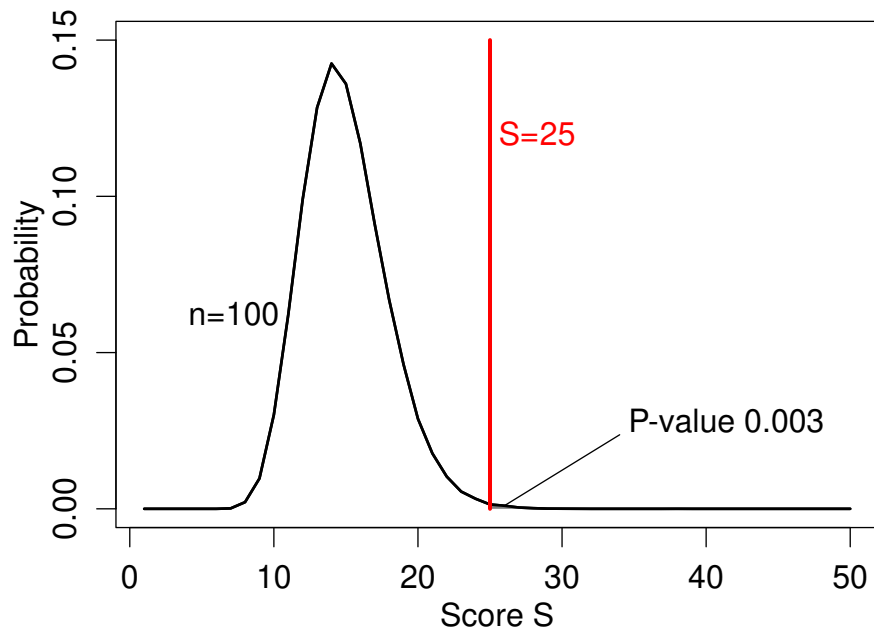
E -value: Expected number of alignments with the score of at least S when searching for a **random query** of length m in a **random database** of length n

Note: $P = 1 - e^{-E} \Rightarrow$ for very small values of E , $P \approx E$

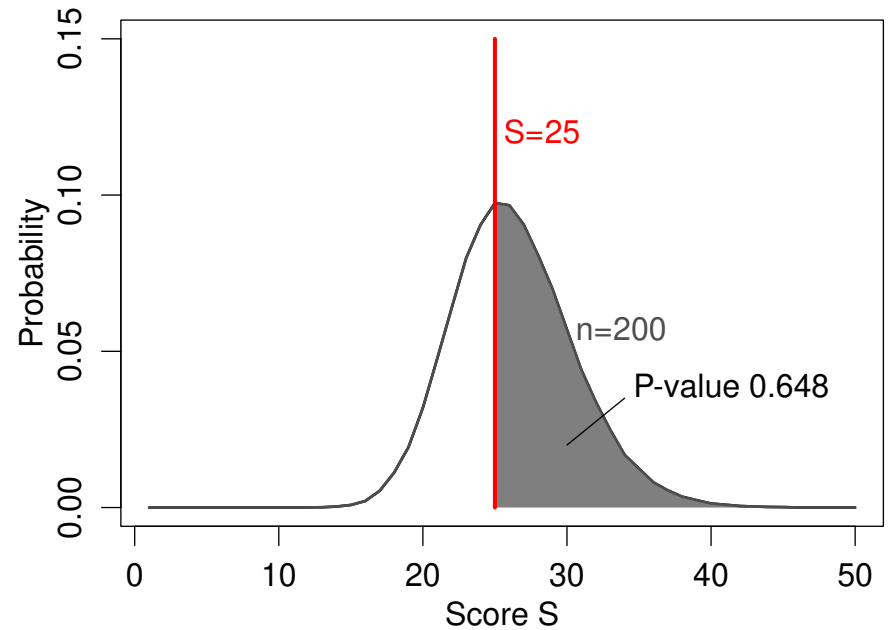
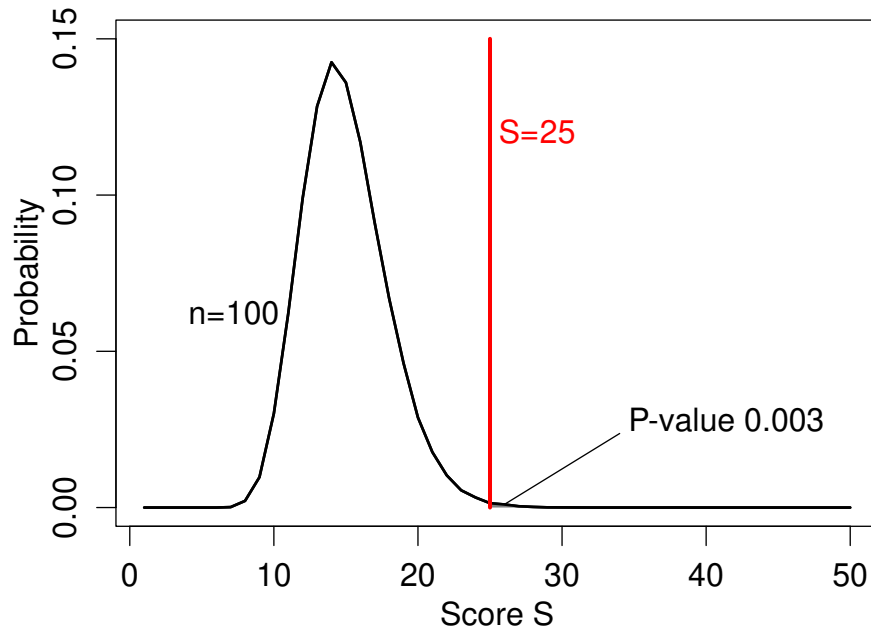
[Karlin and Altschul, 1990, Dembo et al., 1994]

Computing P -values by simulation

- Generate a random query and a random database of length n
- Compute best local alignment (+1/-1 scheme)
- Record the resulting score
- Repeat many times



Computing P -values by simulation (cont)



P-value for score 25:

How many alignments have score 25 or higher?

(In practice, simulations are slow, but we have mathematical estimates of how these distributions look like.)

Summary

- Sequence alignment is the essential bioinformatics tool
- Problem formulation: defining a scoring scheme
- Problem solution: either slow and exact algorithms, or fast heuristics that can miss some alignments
- There are specialized tools for various tasks related to the sequence alignment
- Estimation of statistical significance (P -values) is an important tool in distinguishing real alignments from those that occur just by chance