

# PAM and BLAST

**Ulf Leser** 

### Klausurtermin

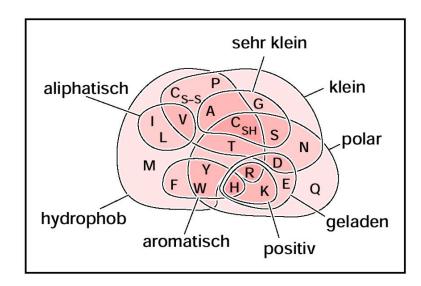
- Montag, 27.8.2018, 11-14 (11.30 13.30) Uhr
- Raum: 3.001
- Keine Hilfsmittel erlaubt
- Anmelden
- Übungsschein
- Mündliche Prüfungen

### This Lecture

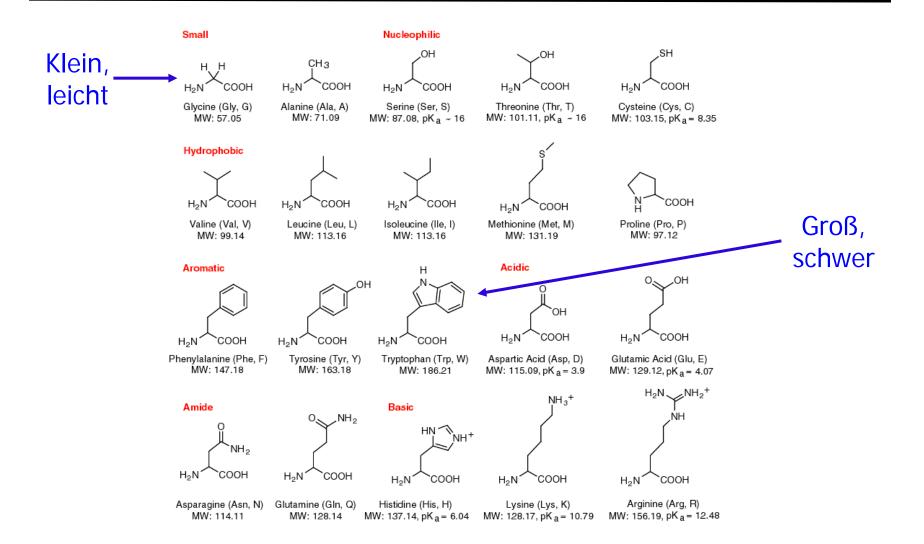
- Substitution Matrices
  - PAM distance
  - PAM matrices
- Scaling up Local Alignments
  - BLAST

#### **Substitution Matrices**

- Recall
  - A scoring function (substitution matrix) is a function s:  $\Sigma' x \Sigma' \rightarrow N$
- DNA: symmetric, simple matrices
- Protein sequences are different
  - Different AA have very different properties
  - Substitutions may change the 3D structure completely or just a little bit or not at all

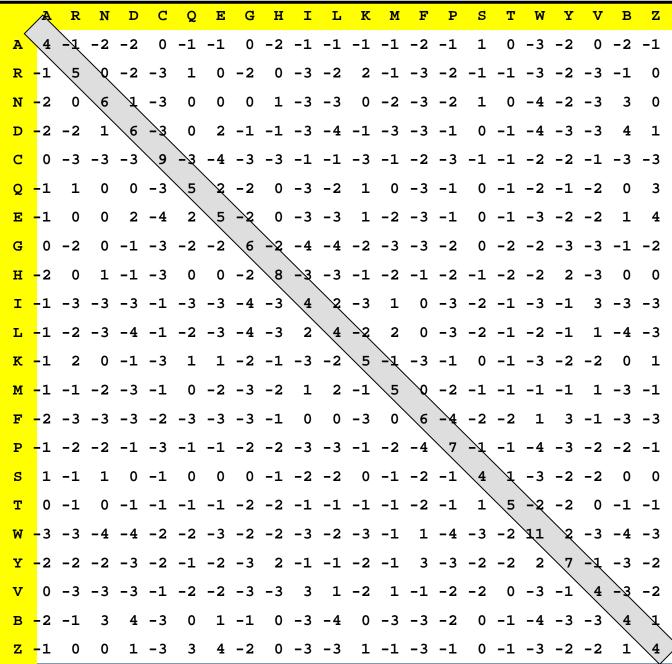


### **Amino Acids**



# Example

Where do all these numbers come from?



## Is it Really Necessary?

Code	Häufig- keit	Mutier- barkeit
L	0.091	54
A	0.077	100
G	0.074	50
S	0.069	117
V	0.066	98
E	0.062	77
K	0.059	72
T	0.059	107
I	0.053	103
D	0.052	86
P	0.051	58
R	0.051	83
N	0.043	104
Q	0.041	84
F	0.040	51
Y	0.032	50
M	0.024	93
Н	0.023	91
C	0.020	44
W	0.014	25

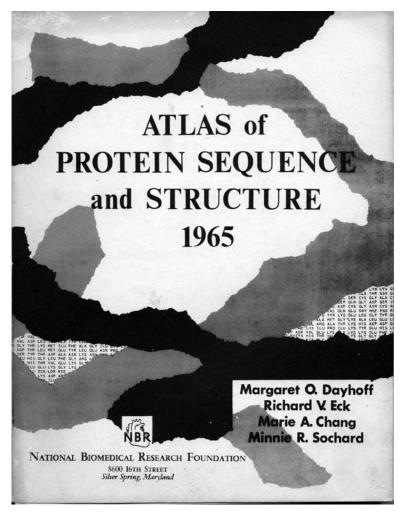
- We count how often a particular AA was replaced by any other AA
  - Using "sure" sequence alignments
- Replacement rate of Alanin (A) := 100%
- Obviously no equal distribution
- Even if we assume that mutations happen more or the less at the same rate, they obviously don't survive at the same rate
  - Mutations are suppressed to different degrees
  - W (Tryptophan): Strong suppression
  - S (Serin): Little suppression

# Filling a Substitution Matrix for Protein Sequences

- We need app. (20\*20)/2=200 values
  - Scoring functions should be symmetric
- Possibility 1: Analytical
  - Capture weight, polarity, size, ...
  - Find a scoring scheme to measure the difference between two AA
  - Needs to produce a single value per AA pair
  - Not used in practice
- Possibility 2: Empirical
  - Count which substitutions survived at which frequency in reality
  - Needs true alignments: Pairs of homologues and aligned sequences

## Margaret O. Dayhoff

- Goal: "Deduce evolutionary relationships of the biological kingdoms, phyla, and other taxa from sequence evidence"
- Collection of all known protein sequences
  - First edition: 65 proteins
  - Several releases followed
  - Resulted in the Protein
     Information Resource (PIR)



Thanks to Antje Krause

### PAM: Point-Accepted Mutations

- Dayhoff, M. O., R. V. Eck, C. M. Park. (1972)
   A model of evolutionary change in proteins.
   in M. O. Dayhoff (ed.), Atlas of Protein Sequence and Structure Vol. 5.
- PAM has two meanings
  - 1 PAM Unit for measuring the similarity of two AA sequences
  - PAM-X matrix Substitution matrix to use when aligning two sequences that are X PAM distant

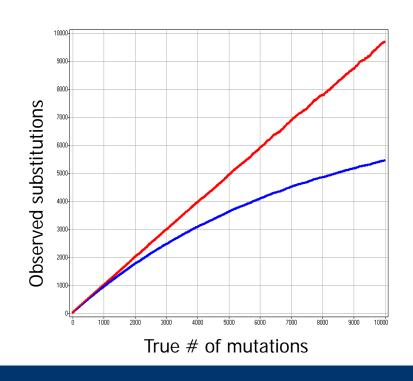
#### PAM as Distance Measure

#### Definition

Let  $S_1$ ,  $S_2$  be two protein sequences with  $|S_1| = |S_2|$ . We say  $S_1$  and  $S_2$  are x PAM distant, iff  $S_1$  most probably was produced from  $S_2$  with x mutations per 100 AAs

#### Remarks

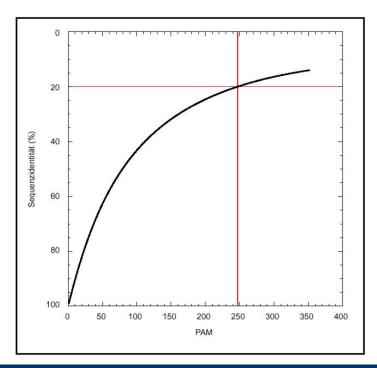
- PAM is motivated by evolution
- Assumptions: Mutations happen with the same rate at every position of a sequence
- If mutation rate is high or time is long, mutations will occur at the same positions
- PAM ≠ %-sequence-identity



### PAM as Distance Measure

- No InDels, only replacements
- The PAM distance d of two DNA sequences can be derived analytically from their %-sequence-diversity p
  - d = -3/4\*ln(1-4/3\*p)
  - p = 1-"sequence identity"
  - Derivation skipped
- Pairs with PAM >250 are probably not homologues
  - %-sequence-identity < 20%</p>
  - Twilight zone
  - Which %-sequence-identity will two random protein sequences have?

(Jukes-Cantor model)



#### **PAM Matrices**

- The PAM-X matrix contains measures for the probability that a AA (column) was replaced by another AA (row) in two sequences that are x PAM distant
- Estimated from data
  - Let  $(S_{1,1}, S_{2,1}), ..., (S_{1,n}, S_{2,n})$  be n x-PAM distant pairs of aligned sequences
  - Compute f(i), the relative frequency of AA A<sub>i</sub> in all pairs
  - Compute f(i,j), the relative substitution frequency of A<sub>i</sub> and A<sub>j</sub>
    - Number of positions k in any of the aligned pairs with  $S_{1,z}{}^{i}[k]=A_{i}$  and  $S_{2,z}{}^{i}[k]=A_{i}$  or vice versa
  - Then

$$M_x(i, j) = \log \left( \frac{f(i, j)}{f(i) * f(j)} \right)$$

### Some Explanations

### Log-likelihood ratio combining

- Expectation: chances to generate this mutation by chance given the relative frequencies of the two involved AAs
- Observation: observed frequency of this mutation

$$M_x(i, j) = \log \left( \frac{f(i, j)}{f(i) * f(j)} \right)$$

### Meaning

- M(i,j) = 0: No selection
- M(i,j) < 0: Negative selection, suppression of mutation</li>
- M(i,j) > 0: Positive selection, mutation is favored

## Example

 $S_{1,1}$ : ACGTGAC

 $S_{2,1}$ : AGGTGCC

 $S_{1,2}$ : GTTAGTA

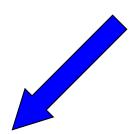
 $S_{2,2}$ : TTTAGTA

 $S_{1,3}$ : GGTCA

 $S_{2,3}$ : AGTCA

### Relative frequencies

A: 10/38 C: 6/38 G: 11/38 T: 11/38



#### Mutation rates

	А	С	G	T
Α	4/19	1/19	1/19	0/19
С		2/19	1/19	0/19
G			4/19	1/19
Т				5/19

#### Matrix

	Α	С	G	T
Α	0,48	0,10	-0,16	•
С		0,63	0,06	-
G			0,40	-0,20
Т				0,50

#### **Problems**

- Depends on predefined alignments
- We need a substitution matrix to find optimal alignments
  - A hen-egg problem
  - Alternative: Do it manually using experience, 3D-structure, ...
- Makes several assumptions
  - Mutations are equally likely at every position in a sequence
  - Mutations are equally likely independent from AA neighbors

**–** ...

### **Real Substitution Matrices**

- PAM requires large n for each x to adequately capture rare mutations
- Dirty trick: Molecular clock assumption
  - Assume that mutations appear with equal rate over time
  - Then the frequencies of PAM-x mutations depend on the frequencies of PAM-1 mutations
  - PAM-x matrices are computed by repeated matrix multiplication of PAM-1 with itself (assuming a linear relationship)
- The complete (highly heuristic) procedure
  - Choose set of n pairs with small PAM distance and align manually
  - Use these alignments to compute M<sub>1</sub>
  - Compute  $M_x = (M_1)^x$

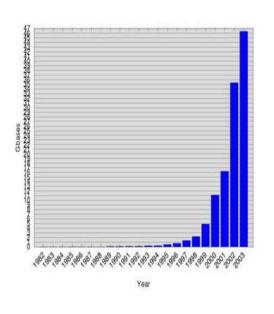
#### **BLOSUM**

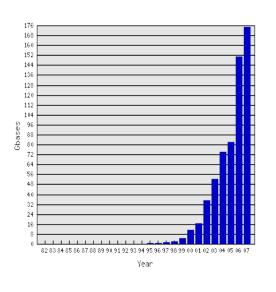
- PAM is a bit old-fashioned
- BLOSUM: BLOcks SUbstitution Matrix
  - Henikoff and Henikoff, 1993
  - Removes assumption of equal mutation rates across each sequence position by considering conserved blocks
  - Direct estimation for different PAM distances instead of errorpropagating self multiplication

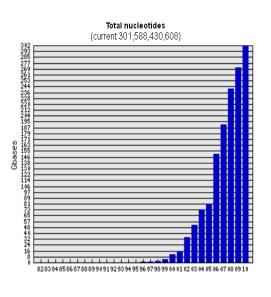
### This Lecture

- Substitution Matrices
  - PAM distance
  - PAM matrices
- Scaling up Local Alignments
  - BLAST

# Scaling Up Local Alignment







- Searching similar sequences (with a high local alignment score) is a fundamental operation in Bioinformatics
- Sequence databases grow exponentially
- We need faster algorithms, even if they sometimes fail

## Similarity Search Problems and their Accuracy

- Task: Given a sequence s and a database D, find all sequences T in D that are sufficiently local-similar to s
  - Often, exactly computing T is not feasible and not necessary (think of the WWW and search engines)

Prediction

- Assume a method that finds a set X of answers for s
- How good is this method?
  - Some sequences will be in X and T true positives
  - Some will be in X but not T false positives
    - Also called Type I error

Some will be in T but not X – false negatives Reality

- Also called Type II error
- Some will be neither
   in X nor T true negatives

### **Precision and Recall**

- Precision = TP/(TP+FP)
  - What is the fraction of correct answers in X?

Prediction

	Reality	
	+	•
+	TruePositive	FalsePositive
	(TP)	(FP)
-	FalseNegative	TrueNegative
	(FN)	(TN)

- Related to specificity
- Recall = TP/(TP+FN)
  - Which fraction of correct answers from T are also in X?
  - Also called sensitivity
- Trade-Offs
  - Usual methods compute a score per element of D
  - All sequences with a score above a threshold t are returned as X
  - Increasing t : higher precision, lower recall
  - Lowering t: lower precision, higher recall
  - if the score correlates with correctness ...

### Example

• Let |DB| = 1000, |X| = 15, |T| = 20,  $|X \cap T| = 9$ 

	Real: Positive	Real: Negative
Alg: Positive	TP = 9	FP = 6
Alg: Negative	FN = 11	TN= 974

- Precision = TP/(TP+FP) = 9/15 = 60%
- Recall = TP/(TP+FN) = 9/20 = 45%
- Assume we increase t: |X|=10, |X∩T|=7

	Real: Positive	Real: Negative
Alg: Positive	TP = 7	FP = 3
Alg: Negative	FN = 13	

– Precision: 70%, recall = 35%

#### **BLAST**

- Altschul, Gish, Miller, Myers, Lipman: "Basic Local Alignment Search Tool", J Mol Bio, 1990
  - A heuristic algorithm for sequence similarity search
  - Very fast, high recall, not perfect
  - Very successful: You "blast" a sequence
  - NCBI runs thousands of BLAST searches every day
- A family of tools
  - Gapped-BLAST, PSI-BLAST, MegaBlast, BLAST-ALL, PATHBLAST, Name-BLAST, ...
  - BLAST for DNA, protein, DNA-protein, protein-DNA, ...
  - We only look at the simple DNA-DNA version
  - We skip several heuristic and domain-specific tricks

#### Fundamental Idea

- Fundamental idea: If two sequences have a good local alignment, then the matching area contains, with very high probability, a sub-area where the match is even better (or even exact)
- These sub-areas are called seeds

TTGACTCGATTATAGTCGCGGATATACTATCG CCTATCACAAGAATATAGTCCCTGATCCAGC

TTGACTC GATTATAGTCGCGGAT ATACTATCG
CCTATCACAA GAATATAGTCCCTGAT CCAGC

TTGACTC GATTATAGTCGCGGAT ATACTATCG
CCTATCACAA GAATATAGTCCCTGAT CCAGC

## Algorithm

- Given query sequence s and sequence database D={d<sub>i</sub>}
- 1. Compute all substrings s<sub>i</sub> of s of length q
  - Also called q-grams
  - How many?
- 2. Find all approximate occurrences of all s<sub>i</sub> in all d<sub>i</sub>
  - Gap-free alignment with matrix; score must be above threshold t
  - Hits are called seeds approx. occurrences of some s<sub>i</sub> in d<sub>i</sub>
- 3. Extend seeds to left and right in s<sub>i</sub> and d<sub>j</sub> until
  - [Constantly update the similarity score]
  - ... the score drops sharply
  - $\dots s_i$  or  $d_i$  ends
  - ... the score gets too bad compared to other hits found earlier

### Example

q=5, t= 3, Matrix: M=+1, R=-1
s=ACGTGATA
d=GATTGACGTGACTGCTAGTGATACTATAT



 $s_1$ =ACGTG  $s_2$ =CGTGA  $s_3$ =GTGAT  $s_4$ =TGATA

GATTGACGTGACTGCTAGTGATACTATAT
GATTGACGTGACTGCTAGTGATACTATAT
GATTGACGTGACTGCTAGTGATACTATAT
GATTGACGTGACTGCTAGTGATACTATAT





GATTGACGTGACTGCAAGTGATACTATAT	
ACGTGATA	5
ACGTGATA	5+1=6
ACGTGATA	6-1=5
• • •	• • •

### **Properties**

- Finding seeds efficiently requires more work
  - Pre-compute all q-grams of all d<sub>i</sub>
  - Group by q-gram
  - Called a hash-index (should be kept in main memory)
  - Lookup: Given s, find all matching q-grams (as seeds)

#### Exclusion method

- Vast majority of all sequences in DB are never looked at because they do not contain a seed
- The "seed" idea is the basis of nearly all fast alignment methods

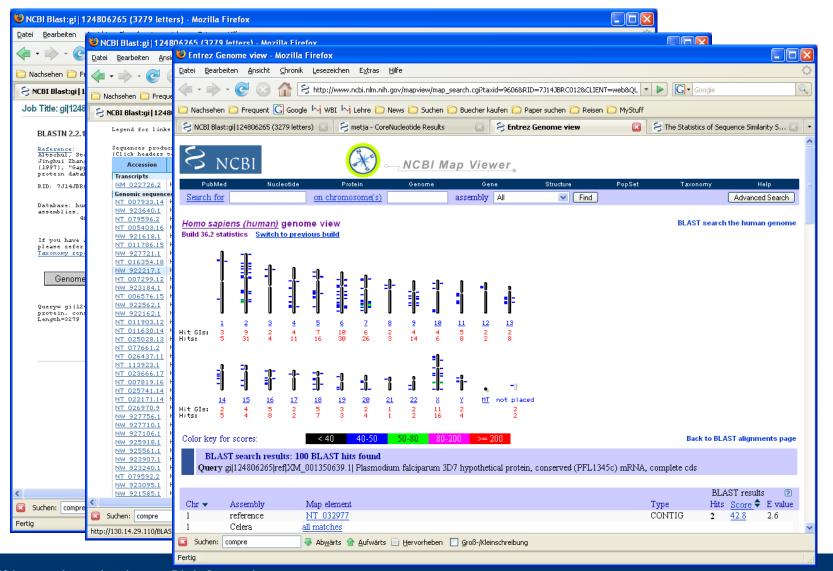
#### Where it fails

- Sensitive to t: Too high missing hits; too low slow
- Does not consider gaps

## Speed – Precision - Recall

- Increasing t
  - Higher requirements for any seed
  - Less seeds, less extensions
  - Lower recall, higher speed, precision stays
- Increasing q (and adapting t)
  - Higher requirements for any seed
  - Less seeds, less extensions
  - Lower recall, higher speed, precision stays

#### **BLAST Screenshots**



#### BLAST-2

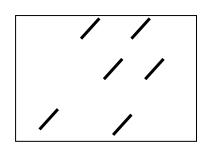
 Altschul, Madden, Schaffer, Zhang, Zhang, Miller, Lipman: "Gapped BLAST and PSI-BLAST: a new generation of protein database search programs", NAR, 1997

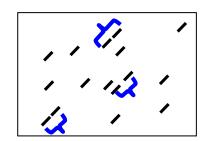


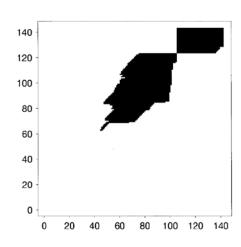
- BLAST: 90% of time spend in extensions
- BLAST2: Two seeds in short distance
  - Needs a decrease in t

### Higher recall

- BLAST didn't even consider gaps in the extension phase
- BLAST2: Full local alignment starting from seeds
  - Allows an increase of t







# **Further Reading**

- Substitution matrixes: Krane & Raymer, Chapter 3
- BLAST, BLAST2: Merkl & Waack, Chapter 12