

Administration

Sprache ?

- zu verhandeln

Selection of topics

- Proteins / DNA / RNA

Two themes (Semesterhälfte), 14 Termine

- Torda: larger molecules, proteins
- Rarey: Chemoinformatics, Wirkstoffentwurf

Extra Lectures

- Do. 30. April + 25. Jun, 18:15 – 19:45 (es tut mir leid)

Administration

Who are we ? (Torda parts, 1st half of semester)

- Andrew Torda
- + Björn Hansen, Iryna Bondarenko,
- admin - Annette Schade (schade@zbh.uni-hamburg.de)

Where am I

- 42838 7331
- ZBH 1st floor (Bundesstr. 43)

Background

- numerical simulations

Übungen

- Mo 14:15
- Mo 16:15
- wenig Flexibilität mit den Gruppen
- Sie können sich untereinander austauschen
- Konten für unseren Rechnerpool (ZBH)

Prüfungen

Schriftliche Klausur

- 90 Minuten
- Letzter Freitag des Semesters - 10. Juli
(wegen eines Biologie-Praktikums)

Bücher

- nicht nötig
- "Understanding Bioinformatics"
Zvelebil, M. & Baum, J.O.
in Bibliothek – leider auf Englisch

Goals, why are we here ?

Overall

- Given some sequence data, what can we find out about it ?
- Some references to structures

Situations – you have

- DNA sequence data (very common)
- corresponding protein sequence (rather common)
 - splice variants, pre/pro-versions
- structure of a protein (much, much less often)
- If you have the structure of a protein, you know the sequence

The Plan

Sequence alignments

- detour protein modelling

Multiple sequence alignments

My examples

Usually well-studied proteins

- myoglobin, haemoglobin, DNA gyrase, metabolic enzymes

In my examples

- you have sequence and structure
- pretend you are trying to work something out from the sequence

In real world

- you may have only sequence information

Data

Structures

- protein data bank (10^5 files)

Sequences

- genbank, NCBI (6.4×10^7 sequences)

Freely available

- downloadable
- searchable via web
- requirement for publication (often)

Not much private/secret data

Predictions

- what shape is this molecule ?
- will this small molecule inhibit some enzyme ?
- will this molecule be broken down in the body quickly ?
- ...

Predictions – different approaches

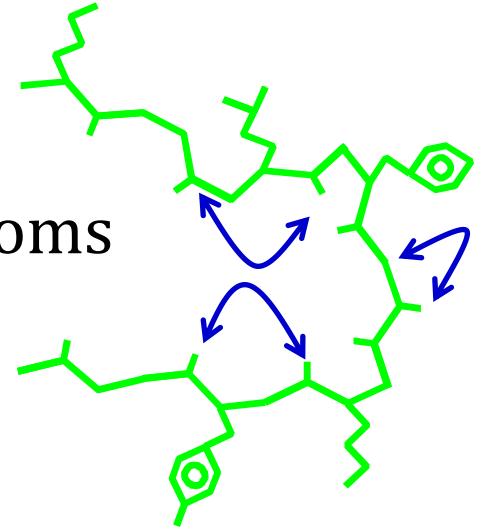
- First principles (physics, chemistry)
- Finding patterns (underlying principles not known)
- Similarity

... explanation

First principles prediction

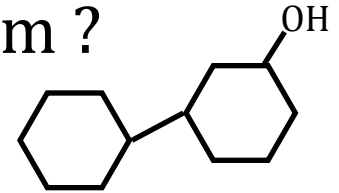
Protein structure example

- a protein molecule = set of atoms in space
- I know all the interactions between the atoms
- should be able to predict the 3D structure



Quantum chemistry

- I have a model for electron wave functions
- can I predict electron density around each atom ?
- predict pK_a for this molecule ?
- ...



Maybe best method

- elegant, expensive, needs good models

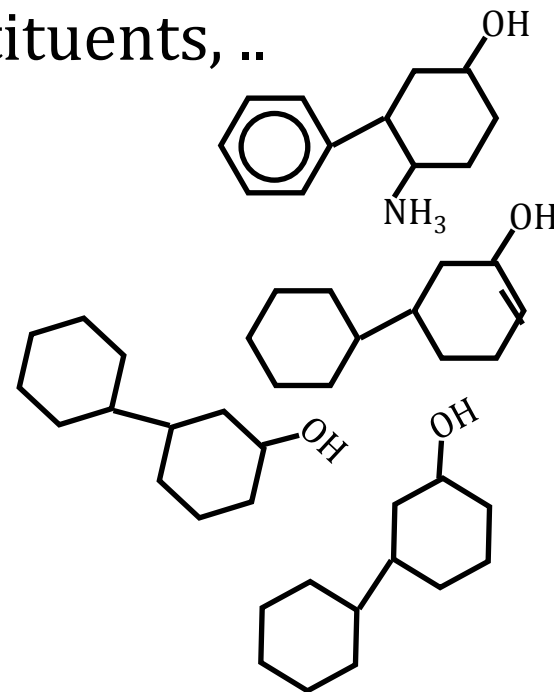
Finding patterns

Take known data – collect properties, look for correlations

- look at mol wt, aromatic/aliphatic, substituents, ..
- for each molecule collect pK_a
- hope patterns can be found

Gene regulator recognition example

- take known examples
 - look at GC content
 - proximity to protein
 - sizes ...



Field of "data mining", machine learning

- little understanding of problem / chemistry
- often works

Similarity – answer many questions

DNA

- is this region coding ?
- where does the reading frame start ?
- is this region involved in regulator binding ?

Protein sequence

- can one guess the structure ?
- is this membrane bound ?
- does it have a certain activity (kinase, transferase, ..) ?

Protein structure

- what is a likely function ?

Prediction by similarity

Need databases

- DNA sequences, protein sequences
- therapeutics with known properties (2nd half of semester)

For some queries / your sequence.. Is your

- protein sequence similar to a known structure ?
- DNA sequence similar to a known regulatory region ?
- RNA similar to some RNase ?

Experiments are slow, expensive, dangerous, smelly

Similarity in sequences

Protein / nucleotide

- same ideas, differences later

Questions

- are two sequences similar ?
- suspected similarity
 - how reliable is it ?
- detailed alignments (modelling, important residues, ..)

Plan

- generalities
- alignment methods
- DNA versions
- Protein versions
- differences

Alignments and Similarities

Problem

.	.	.	A	C	A	C	T	G	A	C	T	A	.	.	.
.	.	.	.	.	A	T	T	G	A	G	T	A	.	.	.
.	.	.	.	.	1	0	1	1	1	0	1	1	.	.	.

- 6 of 8 positions match

Implicitly

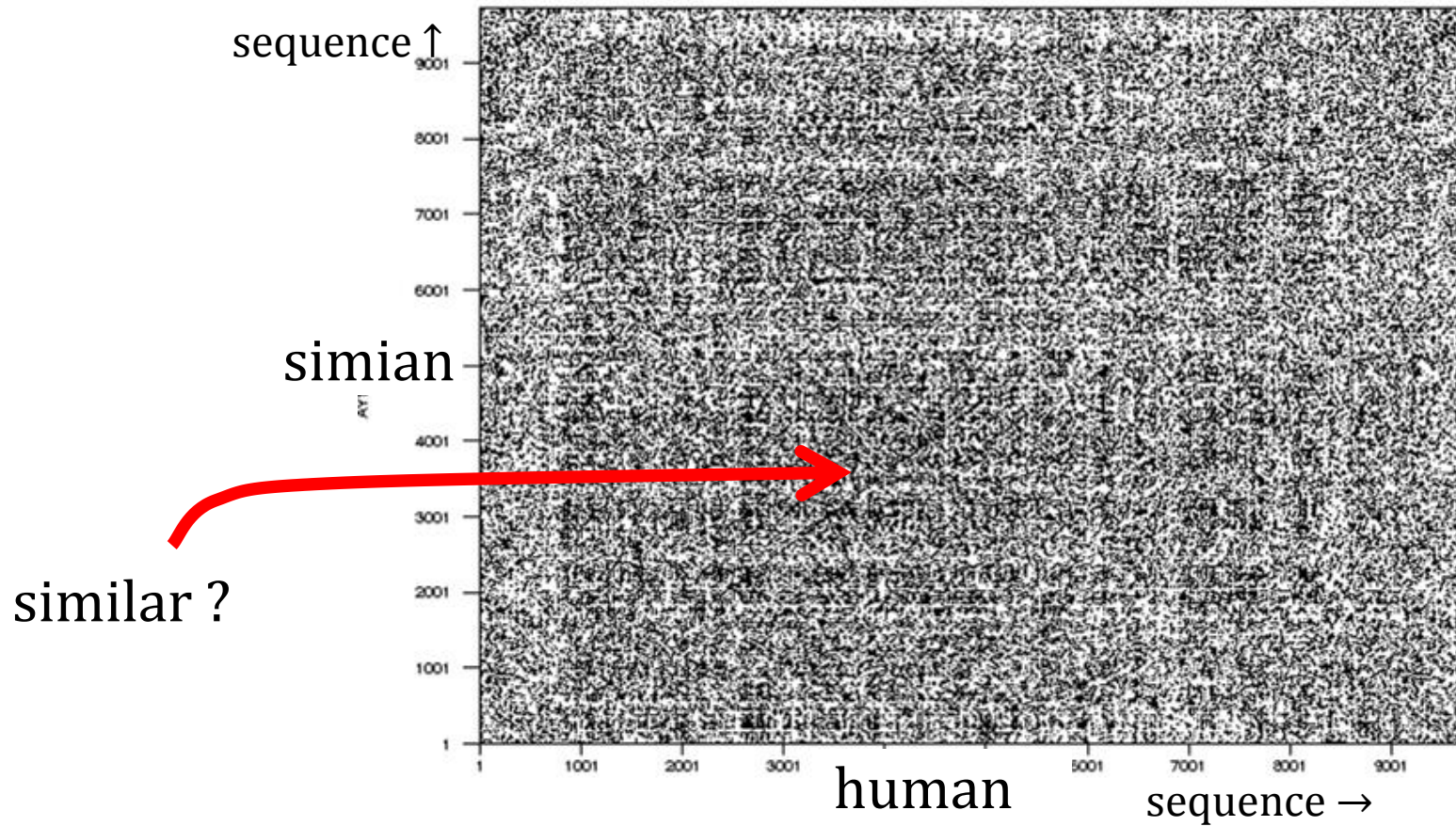
- I have already moved second sequence over the first
- gaps

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

- alignment not so obvious (gaps anywhere)
 - quick look

dot plot

Human and simian HIV



dot plot filtered

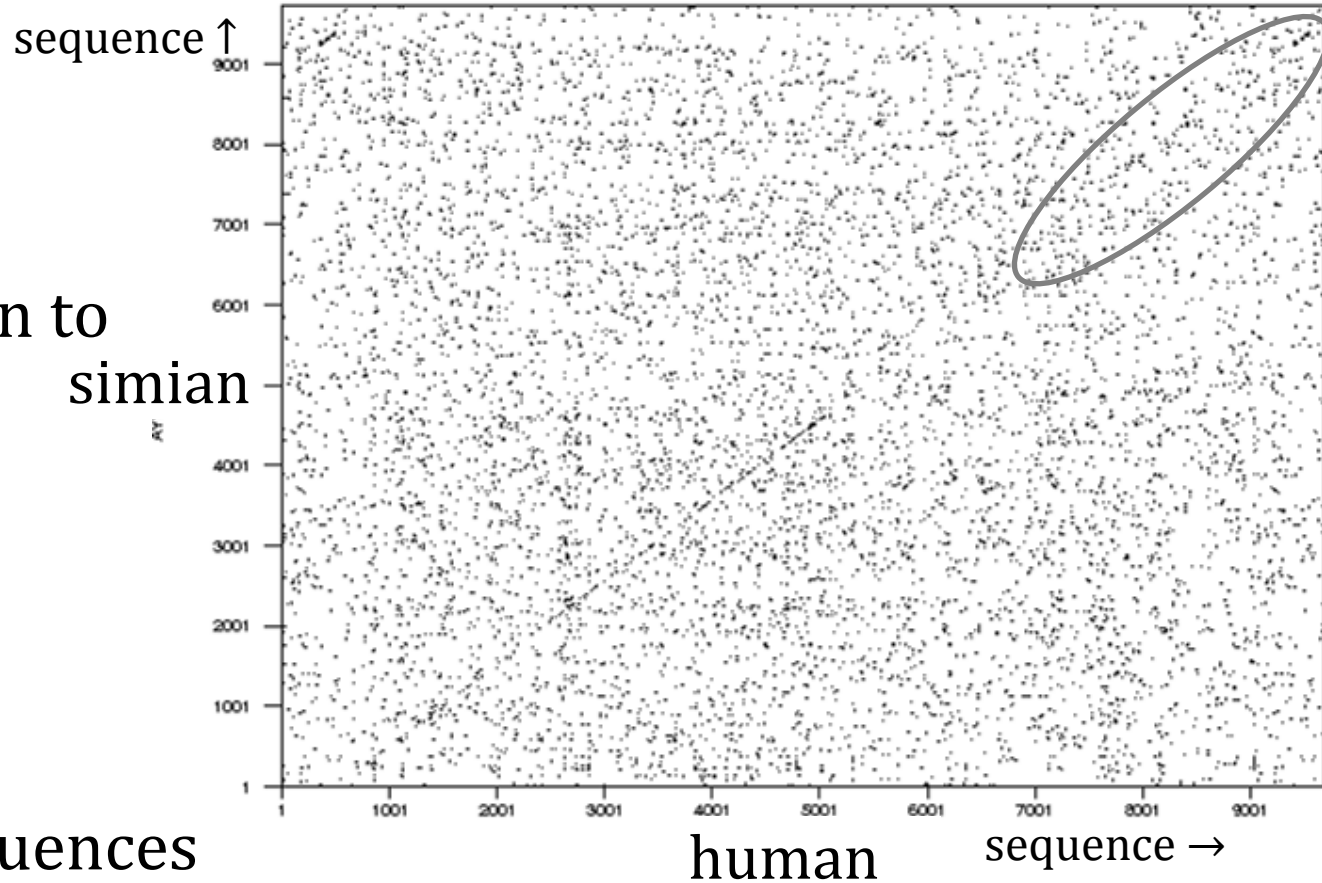
Similarity up to about 5200

Circled region ?

- not so clear
- easy for a human to recognise
- not so easy to automate

Worse case ...

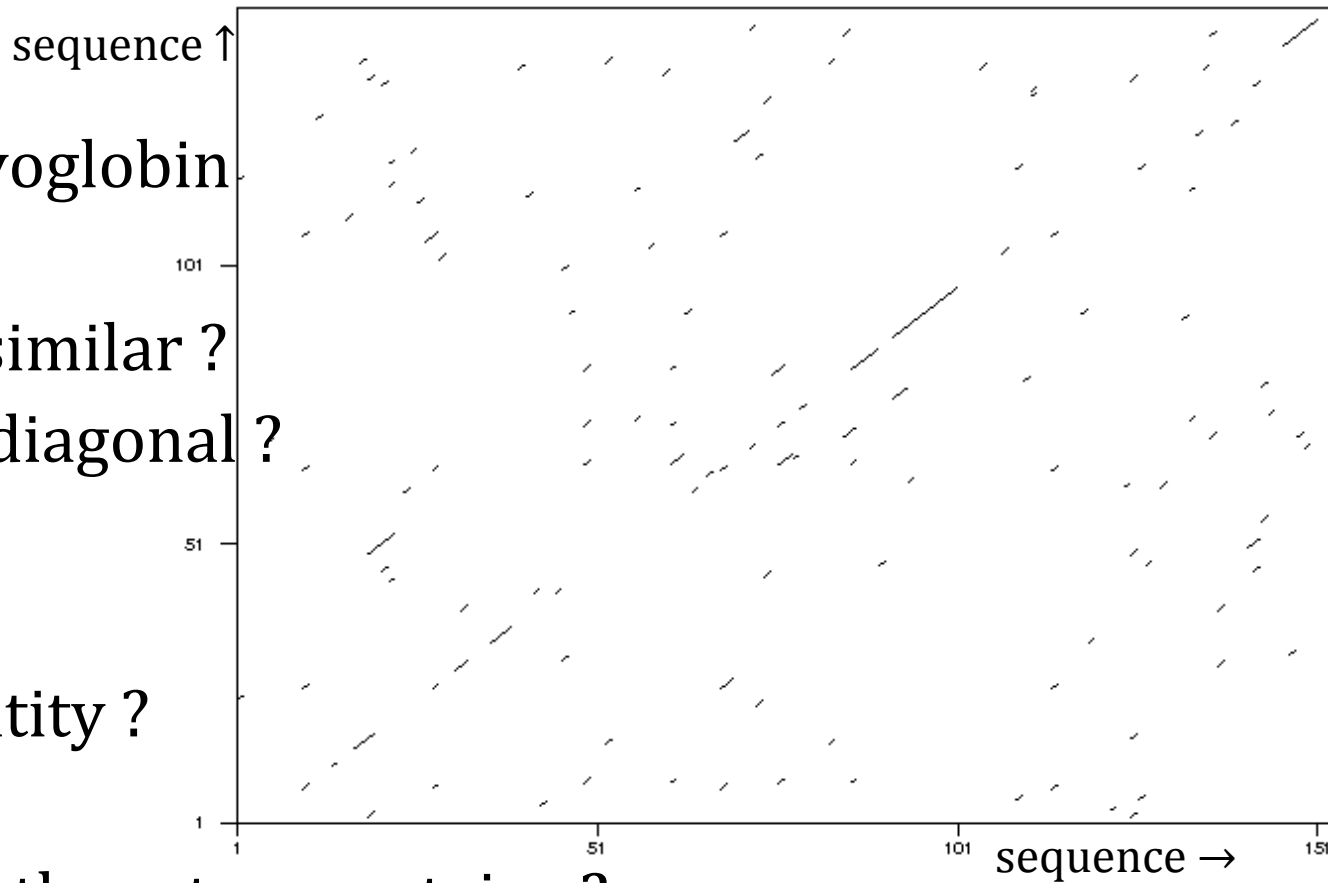
- two protein sequences



protein dot plot

2 proteins

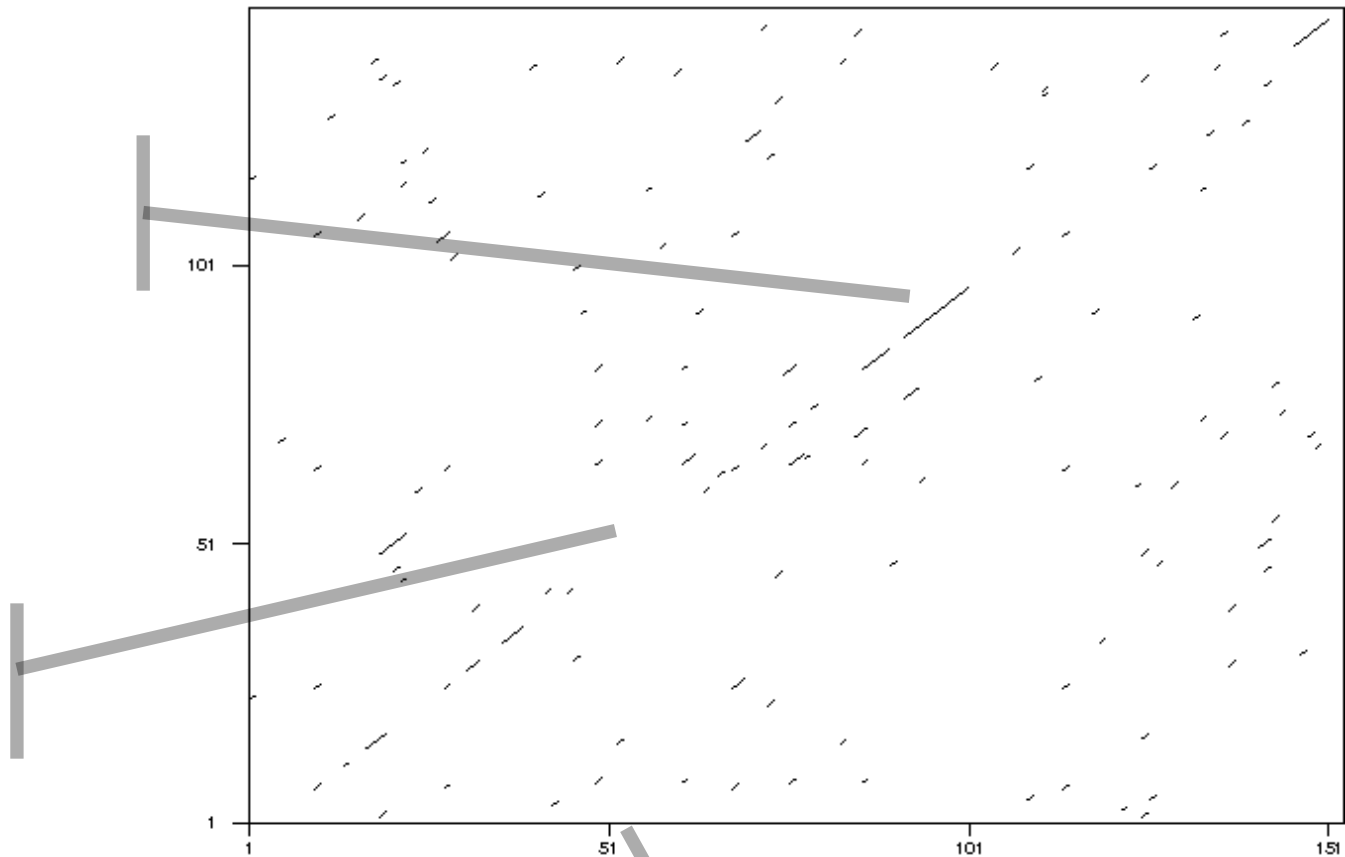
- 2nrl, 2o58
- tuna / horse myoglobin
- are they really similar ?
- how real is the diagonal ?
- what is the identity ?
 - $\approx 45\%$
- how similar are these two proteins ?
- is there a "correct alignment" ? Physical interpretation ?



Properties of alignment ?

Is the alignment
above diagonal ?

What is
happening here ?



Here we know the answer

- look at structures

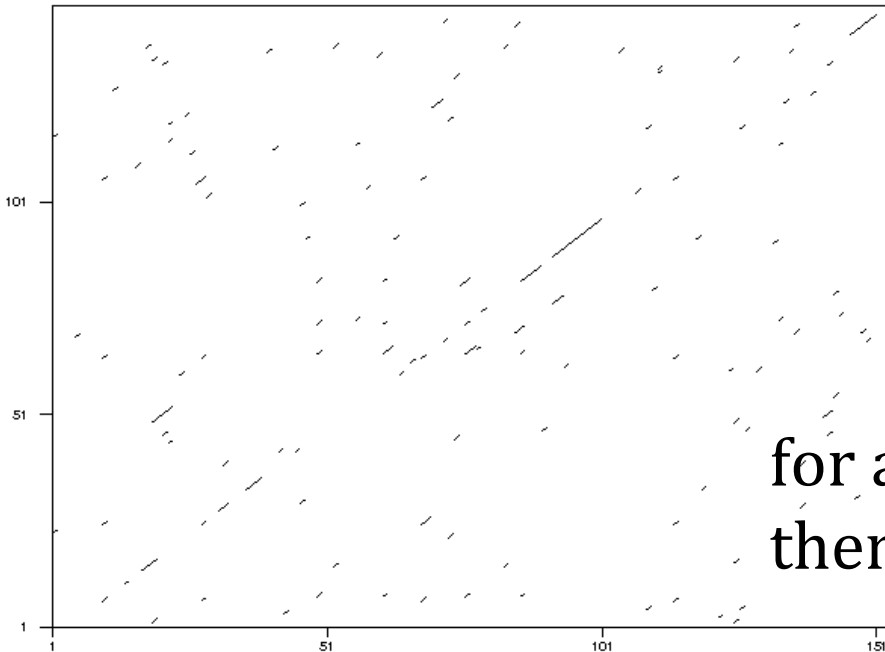
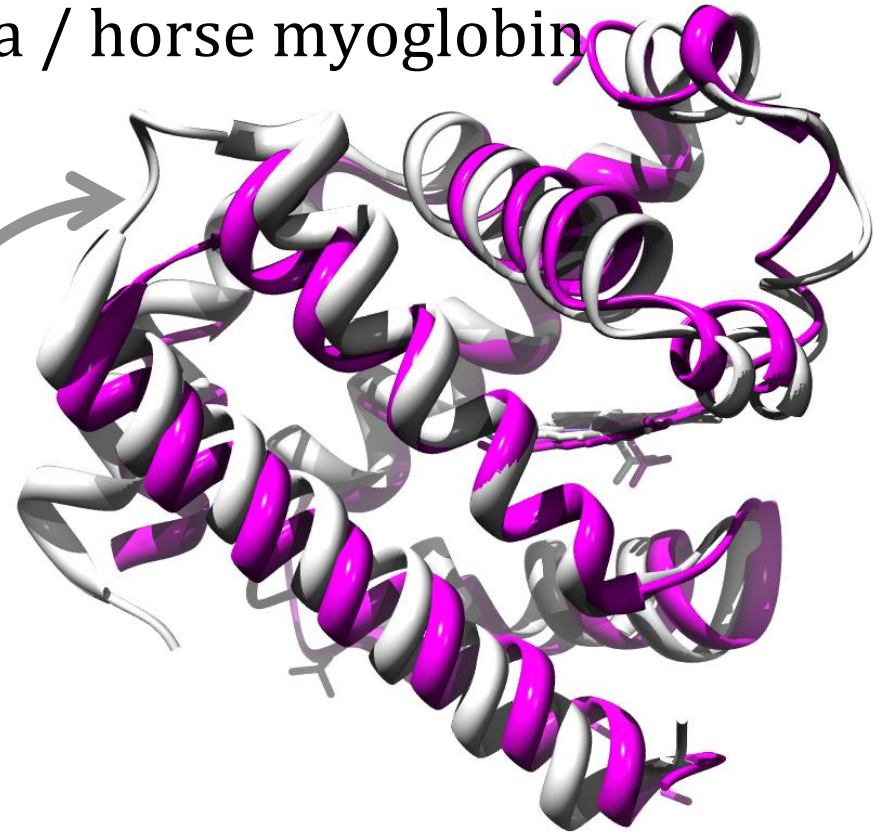
What is aligned to
residue 51 ?

If one knew the structure

The same proteins as before: tuna / horse myoglobin

There are no holes ?

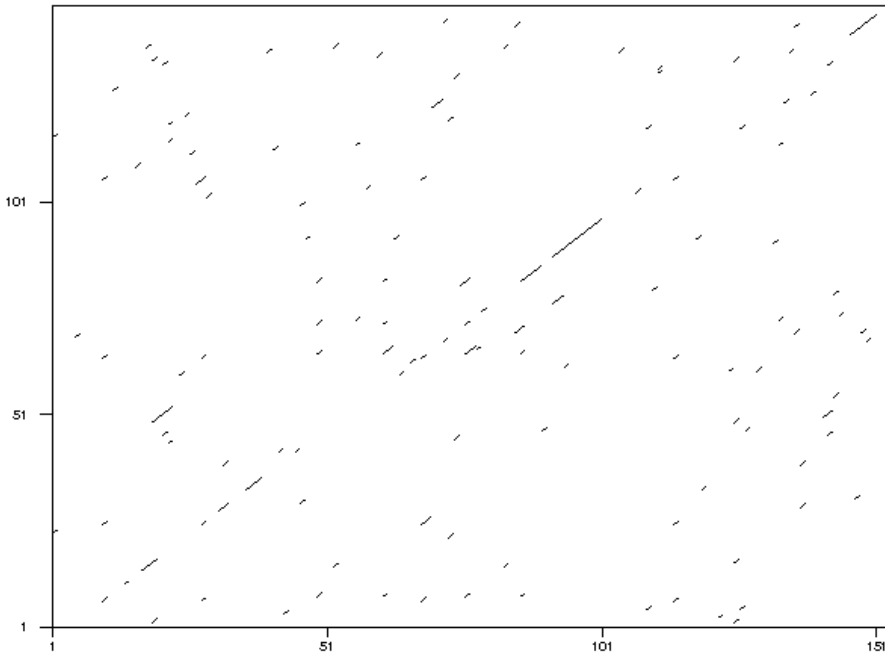
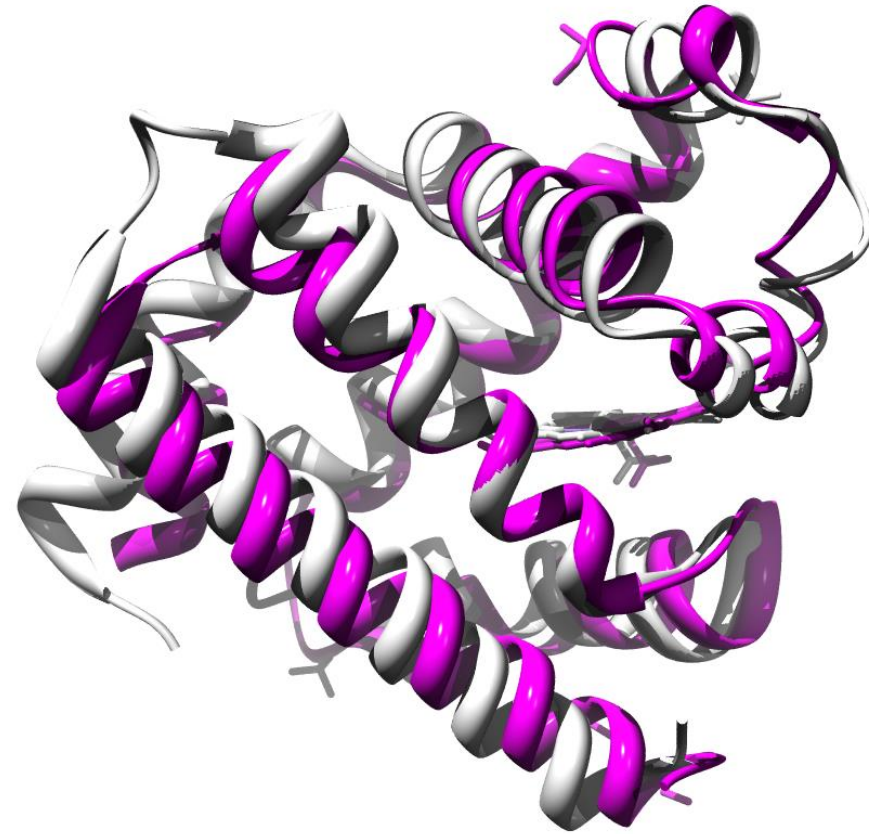
- there are some differences
 - some bits are longer



for almost every pink residue,
there is a corresponding grey residue

If one knew the structure..

Would you have recognised this from dotplot ?



Look at residue 51 in dot plot

- aligned residue not clear

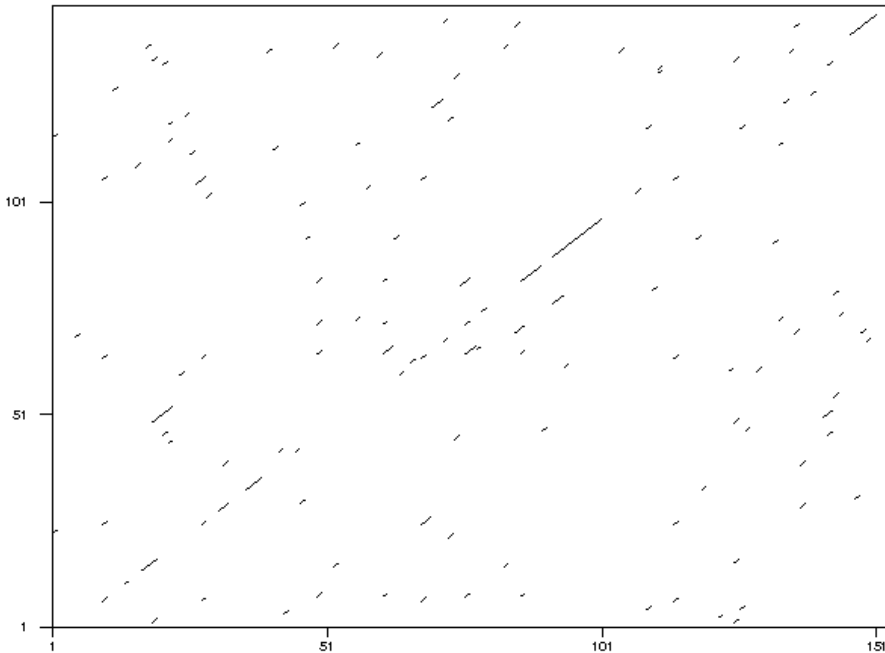
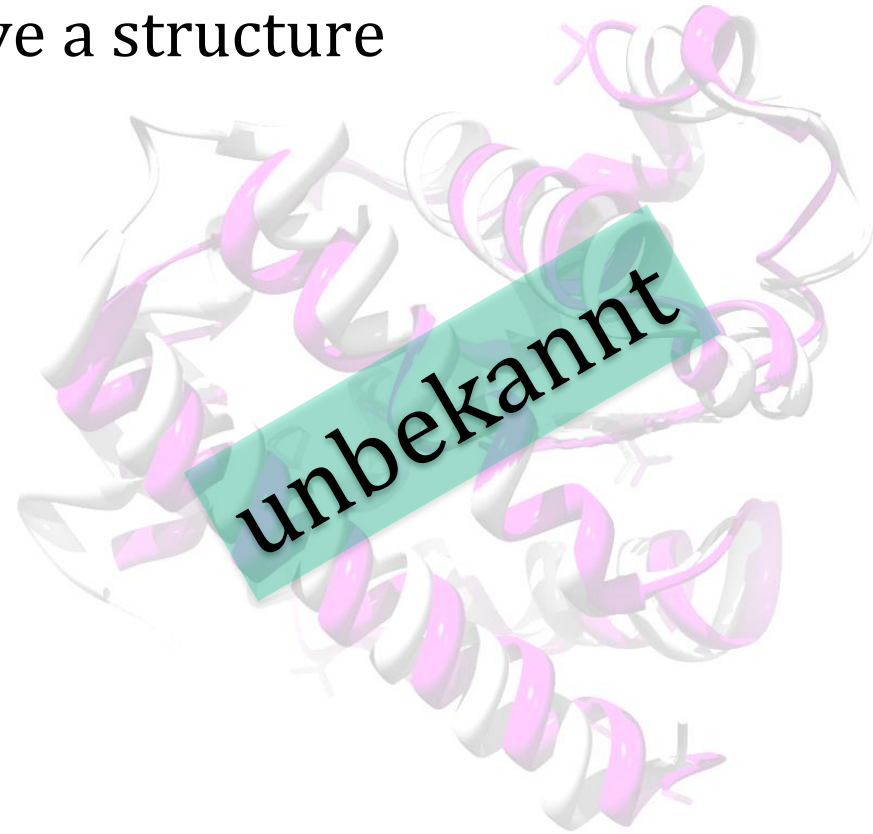
Look in structure

- aligned residues clear

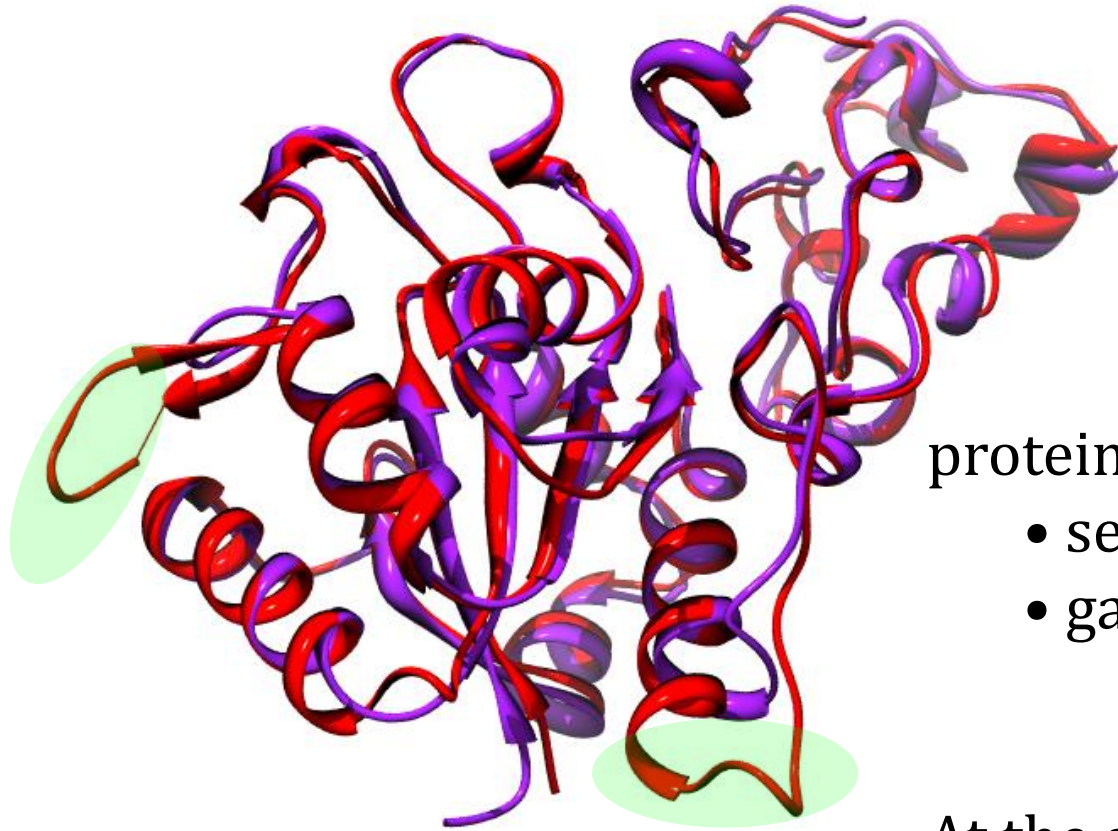
If one knew the structure..

In the first lectures, we do not have a structure

- we get as far as we can from sequence



Clearer Example



hydrogenases

- 40 % sequence identity
- 2frvG & 1cc1S

proteins – obviously similar

- sequence identity OK
- gaps and insertions

At the sequence level ?

Seq ID 40.6 % (103 / 254) in 280 total including gaps

:	1	:	2	:	3	:	4	:	5	:	6
:	0	:	0	:	0	:	0	:	0	:	0

kkapviwvqggctgcsvsllnavhprikeilldvislefhptvmasegemalahmyeia
krpsvvyllhnaectgcsesvlrtvdpvdelildvismdyhetlmagaghaveea-1-he

:	1	:	2	:	3	:	4	:	5	:
:	0	:	0	:	0	:	0	:	0	:

:	0	:	0	:	0	:	1	:	1	:	1
:	7	:	8	:	9	:	0	:	1	:	2
:	0	:	0	:	0	:	0	:	0	:	0

ekfngnffllvegaiptakegrycivgeakahhhevtmmelirdlapkslatvavgtcsa
aikg-dfvcvieggipmgdggwygk-----vggrnmydicaevapakaviaigtcat

0	:	0	:	0	:	0	:	1	:	1
6	:	7	:	8	:	9	:	0	:	1
0	:	0	:	0	:	0	:	0	:	0

:	1	:	1	:	1	:	1	:	1	:	1
:	3	:	4	:	5	:	6	:	7	:	8
:	0	:	0	:	0	:	0	:	0	:	0

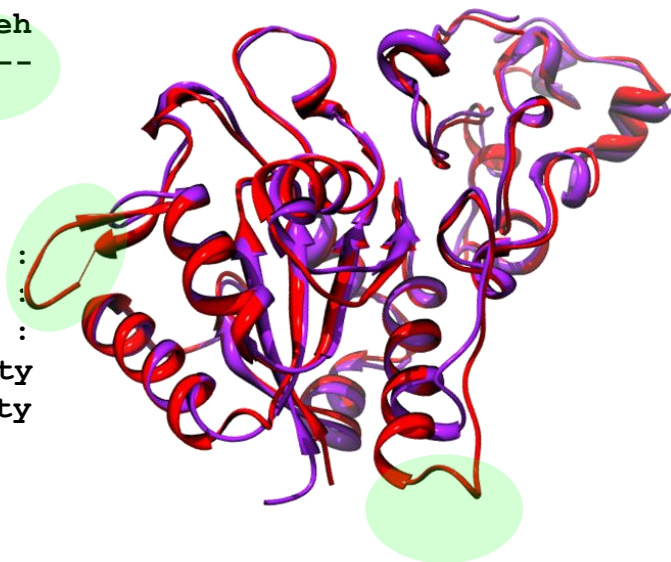
yggipaaegnvtgsksvrddffadekiekllvnvpgcphpdwmvgtlvaawshvlnpteh
yggvqaakpnptgtvgvnealgklgvkai--niagcppnmpnfvgtv--vhlltk-----

:	1	:	1	:	1	:	1	:	1	:
:	2	:	3	:	4	:	5	:	6	:
:	0	:	0	:	0	:	0	:	0	:

:	1	:	2	:	2	:	2	:	2	:
:	9	:	0	:	1	:	2	:	3	:
:	0	:	0	:	0	:	0	:	0	:

plpeldddgrplllffgdnihcnpylldkynsefaetftkpg-----ckaelgckgksty
gmpelddkqgrpvmffgetvhdncprlkhfeagefatsfgspeakkgyclyelgckgpdty

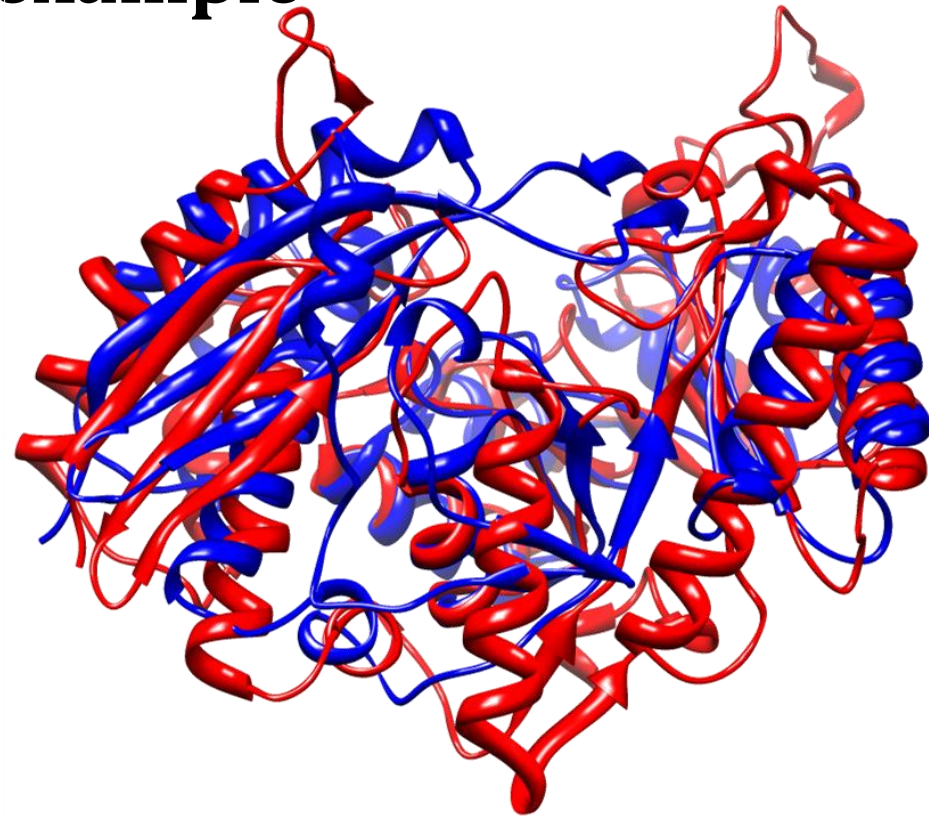
:	1	:	1	:	1	:	2	:	2	:	2
:	7	:	8	:	9	:	0	:	1	:	2
:	0	:	0	:	0	:	0	:	0	:	0



Another example

Structures known

- can easily get sequence alignment



2mnr mandelate racemase
4enl enolase

Seq ID 25.1 % (81 / 323) in 373 total including gaps

```

      1      2      3      4      5
      :      :      :      :      :
      0      0      0      0      0
sktyavlglnngghafaaylalkgqsv--lawdidaqr-----ikeiqdrgaiiaegpg
svehimrdv-nggwa-mryihangaslflavyyihifrglyygsykapreitwivgmviy
0      0      1      1      1
8      :      9      :      1      :      2      :      3      :
0      :      0      :      0      :      0      :      0      :

      :      0      :      0      :      0      :      1      :
      :      6      :      7      :      8      :      9      :      0      :
      :      0      :      0      :      0      :      0      :      0      :
la--gtahpdltsdiglavkdadvilivvpaihhasiaaniaisyiseqgli---ilnpg
llmmgtafmggyvlpwgqmsfwgatvitglfgaipg--igpsiqawllggpavdnatlnrf
1      1      1      1      1      1      1      1      1
4      :      5      :      6      :      7      :      8      :      9      :
0      :      0      :      0      :      0      :      0      :      0      :

      1      :      1      :      1      :      1      :      1      :
      1      :      2      :      3      :      4      :      5      :      6      :
      0      :      0      :      0      :      0      :      0      :      0      :
atggalefrkilrengapevtigetssmlftcrserpgqvtnaikgamdfaclpaakag
fslhyllpf-viaalvaihiwafhttggnnptgvevrrtskadaekdtlpfwpvfvikdl
:      2      :      2      :      2      :      2      :      2      :      2      :
:      0      :      1      :      2      :      3      :      4      :      5      :
:      0      :      0      :      0      :      0      :      0      :      0      :

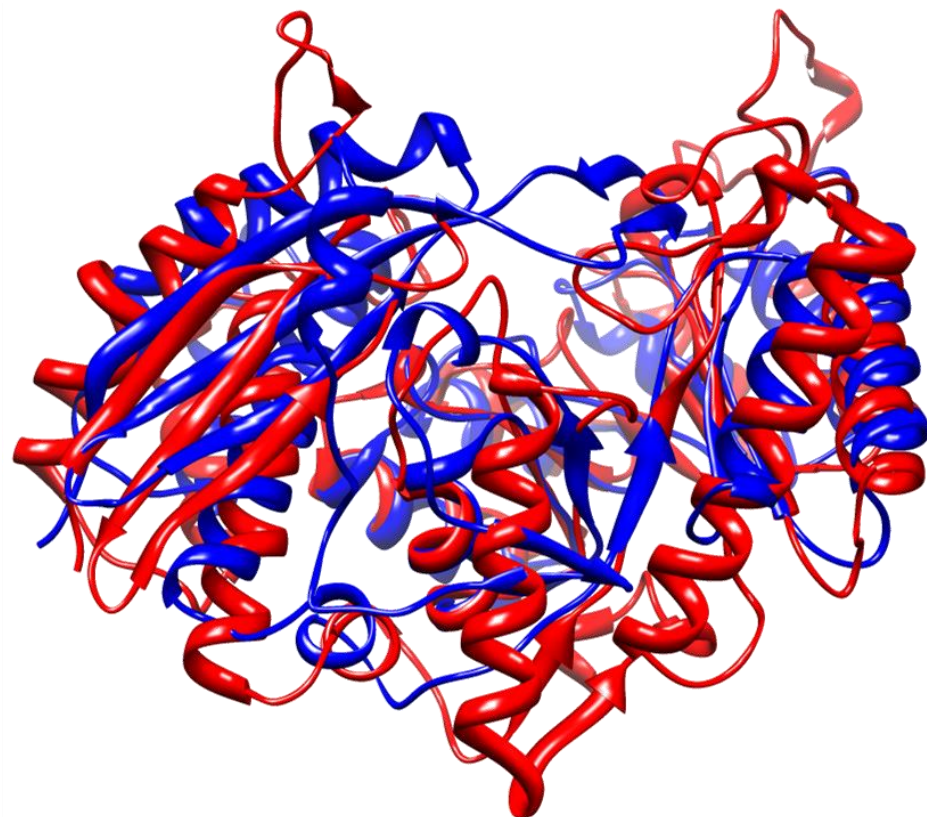
      1      :      1      :      1      :      2      :      2      :      2      :
      7      :      8      :      9      :      0      :      1      :      2      :
      0      :      0      :      0      :      0      :      0      :      0      :
waleqigsvlpqyvavenvlhtsltnv-navm-hplptllnaarcesgtpf----qyyl-
fala-l--vllgffavvaympnylghpdnyvqanplstpahivpewyflpfyailrafaa
:      2      :      2      :      2      :      2      :      3      :
:      6      :      7      :      8      :      9      :      0      :
:      0      :      0      :      0      :      0      :      0      :

      :      2      :      2      :      2      :      2      :      2      :
      :      3      :      4      :      5      :      6      :      7      :
      :      0      :      0      :      0      :      0      :      0      :
-egitpsv-gslaekvdaeriaiakafdlnvpsvcewypatiyeavqgnpayrgiagpin
dvwvvilvdgltfgivdakffgviamfga-i-avmalapw-ldtskvrsgayr---pkf
3      :      3      :      3      :      3      :      3      :
1      :      2      :      3      :      4      :      5      :      6      :
0      :      0      :      0      :      0      :      0      :      0      :

      2      :      2      :      3      :      3      :      3      :
      8      :      9      :      0      :      1      :      2      :      3      :
      0      :      0      :      0      :      0      :      0      :      0      :
lntryffedvstglvplselgravnvptplidavldlisslidtdfrkegrtleklglsg
---rmwfwflvldfvvltwvg-a--m--pt-eypydwis-liastywfay-flvilpllg
:      3      :      3      :      3      :      4      :      4      :
:      7      :      8      :      9      :      0      :      1      :
:      0      :      0      :      0      :      0      :      0      :

      3      :
      4      :
      0      :
ltaag--irsave
atekpepipasie
:      4
:      2
:      0

```



Getting alignments

Do we always know the structure ?

- if so, we would not do these lectures
- sequences are cheap
- structures are expensive

Usually one only has the sequence

Mission for today ?

- how does one find the best alignment based on sequence ?

Alignment methods

Best alignment not obvious

.	.	.	.	.	.	.	C	C	A	T	C	C	G	C	.	.	
.	.	.	C	G	A	T	C	C	-	T	C	C	T	C	.	.	.

6 matches or

.	.	.	.	.	.	.	C	C	A	T	C	C	G	C	.	.	.	.	
.	.	.	.	.	.	.	C	G	A	T	C	C	T	C	C	T	C	.	.

also 6 matches

Can we invent some rules to say which is best ?

How many alignments ?

For two sequences of length 10, how many alignments could I generate ?

. A B C D E F G H I J

Q R S T U V W X Y Z

. Q R S T U V W X Y Z + more

with gaps

Q R S T U V W X Y - Z

Q R S T U V W X - Y Z then with

gap 2

Q R S T U V W X Y - - Z

. . .

- then with multiple gaps ... combinatorial explosion
- do not tackle the problem directly

Mission

For DNA, protein, RNA

- develop some scoring scheme
- maximize matches and similarities

Algorithm

- allow some gaps, not too many
- must be much faster than brute force
- these methods apply to proteins and nucleotides

What is coming

- simple scoring – DNA
- full alignment algorithm (Needleman and Wunsch)
- better scoring – proteins

Scoring for DNA

Sensible scheme

- matched pairs 2
- mismatch -3
- gaps -2

A	C	T	G	-	A	T	T	C	G	A
A	C	-	G	C	A	-	T	C	T	A
2	2	-2	2	-2	2	-2	2	2	-3	2

more sophisticated..

- gap opening costs - 2
- gap widening costs - 1
- so $cost = cost_{open} + (n_{gap} - 1)cost_{widen}$

Representing alignments

- sequences GATTCAGGTTA and GGATCGA

		g	g	a	t	c	g	a	
g									
a									
t									
t									
c									
a									
g									
g									
t									
t									
a									

- would mean
GGAT-CGA-----
-GATTC-AGGTTA
- notes...

Representing alignments

GGAT-CGA-----
 -GATTC-AGGTTA

		g	g	a	t	c	g	a	
g									
a									
t									
t									
c									
a									
g									
g									
t									
t									
a									

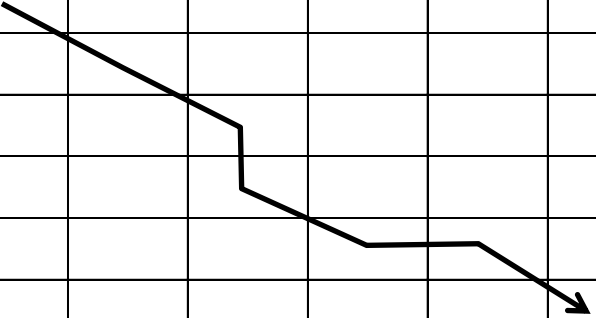
- alignment does not have to go to first / last row or column
- which is x and y is arbitrary
- gaps = row or column is skipped
- work ↘ or ↙ does not matter
- direction must be consistent
 - we only go → ↓

- make sure this is clear

Representing alignments with mismatches

- sequences GCTTCAGGTTA and GGATCGA

		g	g	a	t	c	g	a	
g									
c									
t									
t									
c									
a									
g									
g									
t									
t									
a									



- would mean
GGAT-CGA-----
-GCTTC-AGGTTA

Calculating alignment - steps

Needleman and Wunsch algorithm

1. fill score matrix
2. find best score possible in each cell
3. traceback

fill score matrix

- For convenience, add some zeroes to the ends

		g	g	a	t	c	g	a	
	0	0	0	0	0	0	0	0	0
g	0								0
a	0								0
t	0								0
t	0								0
c	0								0
a	0								0
g	0								0
g	0								0
t	0								0
t	0								0
a	0								0
	0	0	0	0	0	0	0	0	0

Mission

- find path through this matrix with best score
- account for gaps

fill score matrix

- For convenience, add some zeroes to the ends
- Add in match, mismatch scores

		g	g	a	t	c	g	a	
	0	0	0	0	0	0	0	0	0
g	0	2	2	-3	-3	-3	2	-3	0
a	0	-3	-3	2	-3	-3	-3	2	0
t	0	-3	-3	-3	2	-3	-3	-3	0
t	0	-3	-3	-3	2	-3	-3	-3	0
c	0	-3	-3	-3	-3	2	-3	-3	0
a	0	-3	-3	2	-3	-3	2	2	0
g	0	2	2	-3	-3	-3	2	-3	0
g	0	2	2	-3	-3	-3	2	-3	0
t	0	-3	-3	-3	2	-3	-3	2	0
t	0	-3	-3	-3	2	-3	-3	-3	0
a	0	-3	-3	2	-3	-3	-3	2	0
	0	0	0	0	0	0	0	0	0

Mission

- find path through this matrix with best score
- account for gaps

Summing the elements

- start at top left
- move right, then next line
- at each cell
 - find best score it could possibly have

		g	g	a	t	c	g	a	
	0	0	0	0	0	0	0	0	0
g	0	2	2	-3	-3	-3	2	-3	0
a	0	-3	-1	4	-3	-4	-5	4	0
t	0	-3	-3	-3	6	-1	-2	-3	4
t	0	-3	-4	-4	4	3	1	0	2
c	0	-3	-5	-5	-2	6	0	-2	1
a	0	-3	-5	-6	-3	0	3	6	3
g	0	2	0	-6	-4	-1	6	0	6
g	0	2	4	-3	-4	-2	5	3	4
t	0	-3	-1	1	4	-2	-1	2	3
t	0	-3	-3	-1	3	1	-1	0	2
a	0	-3	-4	3	-4	0	-2	4	0
	0	0	-2	0	3	1	0	1	4

Diagonal (no gaps)

for each cell, 3 possible scores

1. **diagonal (no gap)**
2. best from preceding column
3. best from preceding row

		g	g	a	t	c	g	a	
	0	0	0	0	0	0	0	0	0
g	0	2	2	-3	-3	-3	2	-3	0
a	0	-3	-1	4	-3	-4	-5	4	0
t	0	-3	-3	-3	6	-1	-2	-3	4
t	0	-3	-4	-4	4	3	1	0	2
c	0	-3	-5	-5	-2	6	0	-2	1
a	0	-3	-5	-6	-3	0	3	6	3
g	0	2	0	-6	-4	-1	6	0	6
g	0	2	4	-3	-4	-2	5	3	4
t	0	-3	-1	1	4	-2	-1	2	3
t	0	-3	-3	-1	3	1	-1	0	2
a	0	-3	-4	3	-4	0	-2	4	0
	0	0	-2	0	3	1	0	1	4

GAT

GAT

GG

GG

preceding row (gap)

for each cell, 3 possible scores

1. diagonal (no gap)

2. **best from preceding row**

3. best from preceding column

		g	g	a	t	c	g	a	
	0	0	0	0	0	0	0	0	0
g	0	2	2	-3	-3	-3	2	-3	0
a	0	-3	-1	4	-3	-4	-5	4	0
t	0	-3	-3	-3	6	-1	-2	-3	4
t	0	-3	-4	-4	4	3	1	0	2
c	0	-3	-5	-5	-2	6	0	-2	1
a	0	-3	-5	-6	-3	0	3	6	3
g	0	2	0	-6	-4	-1	6	0	6
g	0	2	4	-3	-4	-2	5	3	4
t	0	-3	-1	1	4	-2	-1	2	3
t	0	-3	-3	-1	3	1	-1	0	2
a	0	-3	-4	3	-4	0	-2	4	0
	0	0	-2	0	3	1	0	1	4

GAT
G-T

preceding column (gap)

for each cell, 3 possible scores

1. diagonal (no gap)
2. best from preceding row
3. **best from preceding column**

		g	g	a	t	c	g	a	
	0	0	0	0	0	0	0	0	0
g	0	2	2	-3	-3	-3	2	-3	0
a	0	-3	-1	4	-3	-4	-5	4	0
t	0	-3	-3	-3	6	-1	-2	-3	4
t	0	-3	-4	-4	4	3	1	0	2
c	0	-3	-5	-5	-2	6	0	-2	1
a	0	-3	-5	-6	-3	0	3	6	3
g	0	2	0	-6	-4	-1	6	0	6
g	0	2	4	-3	-4	-2	5	3	4
t	0	-3	-1	1	4	-2	-1	2	3
t	0	-3	-3	-1	3	1	-1	0	2
a	0	-3	-4	3	-4	0	-2	4	0
	0	0	-2	0	3	1	0	1	4

T-C
TTC

The order of cells

- start at top left
- every cell has best score considering all possible routes
- at end, highest score is best path

		g	g	a	t	c	g	a	
	0	0	0	0	0	0	0	0	0
g	0	2	2	-3	-3	-3	2	-3	0
a	0	-3	-1	4	-3	-4	-5	4	0
t	0	-3	-3	-3	6	-1	-2	-3	4
t	0								
c	0								
a	0								
g	0								
g	0								
t	0								
t	0								
a	0								
	0								

- would also work if we went left and up

Reading the alignment

- find highest scoring cell (last row or column)
- how did we reach this cell ?
 - how did we reach preceding cell ?
 - ...

		g	g	a	t	c	g	a	
	0	0	0	0	0	0	0	0	0
g	0	2	2	-3	-3	-3	2	-3	0
a	0	-3	-1	4	-3	-4	-5	4	0
t	0	-3	-3	-3	6	-1	-2	-3	4
t	0	-3	-4	-4	4	3	1	0	2
c	0	-3	-5	-5	-2	6	0	-2	1
a	0	-3	-5	-6	-3	0	3	6	3
g	0	2	0	-6	-4	-1	6	0	6
g	0	2	4	-3	-4	-2	5	3	4
t	0	-3	-1	1	4	-2	-1	2	3
t	0	-3	-3	-1	3	1	-1	0	2
a	0	-3	-4	3	-4	0	-2	4	0
	0	0	-2	0	3	1	0	1	4

GGAT-CGA

-GATTC-AGGTTA

Trick with traceback

For each cell

- how did we reach it ? What was the preceding cell ?

		g	g	a	t	c	g	a	
	0	0	0	0	0	0	0	0	0
g	0	2	2	-3	-3	-3	2	-3	0
a	0	-3	-1	4	-3	-4	-5	4	0
t	0	-3	-3	-3	6	-1	-2	-3	4
t	0	-3	-4	-4	4	3	1	0	2
c	0	-3	-5	-5	-2	6	0	-2	1
a	0	-3	-5	-6	-3	0	3	6	3
g	0	2	0	-6	-4	-1	6	0	6
g	0	2	4	-3	-4	-2	5	3	4
t	0	-3	-1	1	4	-2	-1	2	3
t	0	-3	-3	-1	3	1	-1	0	2
a	0	-3	-4	3	-4	0	-2	4	0
	0	0	-2	0	3	1	0	1	4

GGAT-CGA

-GATTC-AGGTTA

Summary (Needleman and Wunsch)

- Alignments are paths through the matrix
- There is an astronomical number of possibilities (with gaps)
- This algorithm has visited all of them and found best
- allows for gap costs of form

$$cost = cost_{open} + (n_{gap} - 1)cost_{widen}$$

- best or only method ? wait..

Cost

- say both sequences are length n
- we have to visit n^2 cells in matrix
 - each time we have to look at a row or column of length $\approx n$
- total cost n^3 or worst cost $O(n^3)$
 - remember this for later

Smith and Waterman version

So far: global alignments

- best match, covers as much as possible

Imagine proteins with 3 domains

ABCDEABCDEABCDE

QRSTUVWXYZBCDEQRSTU

Want to see ...

ABCDEABCDEABCDE

||||

QRSTUVWXYZBCDEQRSTU

not worth trying to align

everything

Use “Smith and Waterman” method

- scoring scheme: matches positive, mismatches negative
- during traceback
 - do not just look for max score
 - start with positive score
 - stop if score goes negative
- result: “local alignments” – often most useful

Other alignment algorithms

Needleman and Wunsch / Smith Waterman

- for given problem – optimal results
- allow fancy gap penalties
- cost $O(n^3)$

Other methods

- $O(n^2)$ – very small limitation on gaps

Faster

- ...

Faster Seeded Methods (blast, fasta..)

Seeded

- idea: use seeds / fragments of length k
 - 11 - 28 for DNA
 - 2 - 3 for protein
- look for exact matches of query words in database
- extend if found
- time depends mainly on length $O(n)$
 - most of the time no matches
- slow extension when a match is found

Seed size

- very small = lots of unimportant matches (slow)
- too big – may miss a match if there are too many changes

Fast versus slow

2 sequences (protein or DNA)

- time not an issue
- 1 000 alignments ? Time still not an issue
- $10^3 \times 10^3$ alignments ? Your decision

Databases

- non-redundant protein sequence database
 - $\approx 6.4 \times 10^7$ sequences
 - $\approx 2.3 \times 10^{10}$ residues
- must be fast
- maybe occasionally miss a word
- alignments may not be optimal

Problems so far

- We can align DNA sequences – maybe proteins
- how biological are the alignments, gaps and costs ?
- Coding versus non-coding DNA
 - 3 base pairs → 1 residue
ACAG ... lots ... CGA...
 - AC-G ... lots ... CGA ... one base deletion
 - 100's bases are shifted – amino acids in protein all wrong
 - non-coding region (binding / regulation / tRNA / rRNA..)
 - may not be so bad
- General problem – degeneracy ..

Degeneracy and Scoring

CCU, CCC, CCA, CCG are all proline (3rd position degenerate)

CCC→CCA no problem

CCC→ACC pro → ala (you die)

exactly the same mutation at DNA level (C→A)

- our scoring scheme does not know about this

A rule..

- some mutations will have no effect
- some are drastic
- usually the third base in each codon is least important

Can one do better ?

Scoring protein alignments

Two aspects

- forget DNA
- account for amino acid similarity

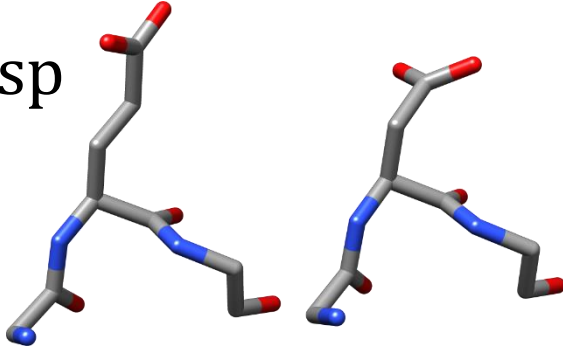
Instead of DNA – work directly with protein sequences

If our DNA is coding – easy to say

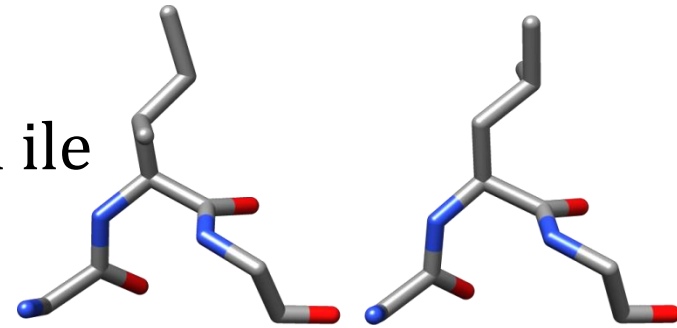
- CCUUCUUAU.. is pro-ser-tyr...
- immediate gain
 - CCC→CCA or similar will not be seen / affect alignment
- more subtle gain...

Amino acid similarities

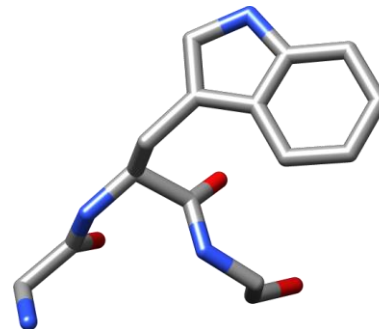
glu and asp



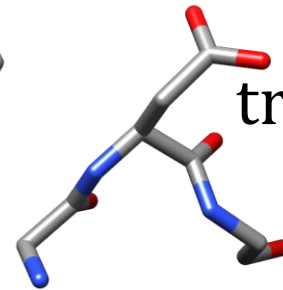
leu and ile



- many more similar examples
- glu → asp mutation, does it matter ? sometimes not
- trp → asp, big hydrophobic to small polar ? usually bad news
- relevance to alignments



trp and asp



Why we need better protein scoring

ANDREWANDRWANDRWW aligned to QNDRDW

ANDREWANDRWANDRWW

QNDRDW-----

ANDREWANDR-WANDRWW

-----QNDRDW-----

ANDREWANDRWANDRWW

-----QNDRDW

- one of which is biologically more likely ($E \rightarrow D$)
- how would we do it numerically ?

Substitution matrices

Earlier in DNA

- match = 2
- mismatch = -3

	A	C	G	T
A	2	-3	-3	-3
C	-3	2	-3	-3
G	-3	-3	2	-3
T	-3	-3	-3	2

We want a matrix that says

	D	E	W	...
D	10	5	-5	
E	5	10	-5	
W	-5	-5	15	
...				

A full matrix..

A serious protein similarity matrix

blosum62:

	A	R	N	D	C	Q	E	G	H	I	L	K	M	F	P	S	T	W	Y	V
A	4	-1	-2	-2	0	-1	-1	0	-2	-1	-1	-1	-1	-2	-1	1	0	-3	-2	0
R	-1	5	0	-2	-3	1	0	-2	0	-3	-2	2	-1	-3	-2	-1	-1	-3	-2	-3
N	-2	0	6	1	-3	0	0	0	1	-3	-3	0	-2	-3	-2	1	0	-4	-2	-3
D	-2	-2	1	6	-3	0	2	-1	-1	-3	-4	-1	-3	-3	-1	0	-1	-4	-3	-3
C	0	-3	-3	-3	9	-3	-4	-3	-3	-1	-1	-3	-1	-2	-3	-1	-1	-2	-2	-1
Q	-1	1	0	0	-3	5	2	-2	0	-3	-2	1	0	-3	-1	0	-1	-2	-1	-2
E	-1	0	0	2	-4	2	5	-2	0	-3	-3	1	-2	-3	-1	0	-1	-3	-2	-2
G	0	-2	0	-1	-3	-2	-2	6	-2	-4	-4	-2	-3	-3	-2	0	-2	-2	-3	-3
H	-2	0	1	-1	-3	0	0	-2	8	-3	-3	-1	-2	-1	-2	-1	-2	-2	2	-3
I	-1	-3	-3	-3	-1	-3	-3	-4	-3	4	2	-3	1	0	-3	-2	-1	-3	-1	3
L	-1	-2	-3	-4	-1	-2	-3	-4	-3	2	4	-2	2	0	-3	-2	-1	-2	-1	1
K	-1	2	0	-1	-3	1	1	-2	-1	-3	-2	5	-1	-3	-1	0	-1	-3	-2	-2
M	-1	-1	-2	-3	-1	0	-2	-3	-2	1	2	-1	5	0	-2	-1	-1	-1	-1	1
F	-2	-3	-3	-3	-2	-3	-3	-3	-1	0	0	-3	0	6	-4	-2	-2	1	3	-1
P	-1	-2	-2	-1	-3	-1	-1	-2	-2	-3	-3	-1	-2	-4	7	-1	-1	-4	-3	-2
S	1	-1	1	0	-1	0	0	0	-1	-2	-2	0	-1	-2	-1	4	1	-3	-2	-2
T	0	-1	0	-1	-1	-1	-1	-2	-2	-1	-1	-1	-1	-2	-1	1	5	-2	-2	0
W	-3	-3	-4	-4	-2	-2	-3	-2	-2	-3	-2	-3	-1	1	-4	-3	-2	11	2	-3
Y	-2	-2	-2	-3	-2	-1	-2	-3	2	-1	-1	-2	-1	3	-3	-2	-2	2	7	-1
V	0	-3	-3	-3	-1	-2	-2	-3	-3	3	1	-2	1	-1	-2	-2	0	-3	-1	4

some features

- diagonal
- similar
- different

Using the score matrix

Algorithm (global alignment, local, fast, ...)

- unchanged
- only scoring changes

If possible use the protein sequence rather than DNA

- not all DNA codes for proteins
- regulators, tRNA, catalytic RNA, sRNA, ..
- not possible for genomic comparisons

Automatically includes codons, amino acid similarity, ..

Where does this kind of matrix come from ?

Substitution Matrices

Many

- PAM point accepted mutations
- BLOSUM blocks substitution matrix

Philosophy

- if two amino acids are similar, we will see mutations often

To quantify this..

- Take some very similar proteins (lots)

parts of some haemoglobins

HAHKLRVGPVNFKLLSHCLLVTLAAHLPAEFTPAVHASLDKFLASVSTVLTSK
HAHKLRVDPVNFKLLSHCLLSTLAVHLPNDFTPAVHASLDKFLSSVSTVLTSK
HAHKLRVDPVNFKLLSHCLLVTLAAHLPAEFTPAVHASLDKFLASVSTVLTSK
HAHKLRVDAVNFKLLSHCLLVTLAAHLPAEFTPAVHASLDKFLASVSTVLTSK
HAHKLRVDPVNFKLLSHCLLVTLAAHLPAEFTPAVHASLDKFLASVSTVLTSK
HAHKLRVDPVNFKLLSHCLLVTLAAHLPAEFTPAVHASLDKFLASVSTVLTSK
HAHKLRVDPVNFKLLSHCLLVTLAAHLPAEFTPAVHASLDKFLASVSTVLTSK
HAHKLRVDPVNFKLLSHCLLVTLAAHLPAEFTPAVHASLDKFLASVSTVLTSK
HAHKLRVDPVNFKLLSHCLLVTLAAHLPAEFTPAVHASLDKFLASVSTVLTSK
HAHKLRVDPVNFKLLSHCLLVTLAAHLPAEFTPAVHASLDKFLASVSTVLTSK
HAHKLRVDPVNFKLLSHCLLSTLAVHLPNDFTPAVHASLDKFLSSVSTVLTSK
HAHKLRVDPVNFKLLSHCLLSTLAVHLPNDFTPAVHASLDKFLSSVSTVLTSK
HAHKLRVDPVNFKLLSHCLLSTLAVHLPNDFTPAVHASLDKFLSSVSTVLTSK
HAHKLRVDPVNFKLLSHCLLSTLAVHLPNDFTPAVHASLDKFLSSVSTVLTSK
HAHKLRVDPVNFKLLSHCLLVTLAAHHPDDFNPSVHASLDKFLANVSTVLTSK
HAHKLRVNPVNFKLLSHSLLVTLASHLPTNFTPAVHANLNKFLANDSTVLTSK
HAYKLRVDPVNFKLLSHCLLVTLACHHPTEFTPAVHASLDKFFTAVSTVLTSK
HAQKLRVDPVNFKFLGHCFLVVVAIHHPSALTPEVHASLDKFLCAVGTVLTA
HAQKLRVDPVNFKFLGHCFLVVVAIHHPSALTAEVHASLDKFLCAVGTVLTA
HAQKLRVDPVNFKFLGHCFLVVVAIHHPSALTAEVHASLDKFLCAVGTVLTA
HAQKLRVDPVNFKLLGQCFLVVVAIHNPSALTPEAHASLDKFLCAVGLVLTA
HAYNLRVDPVNFKLLSQCIVVLAVHMGKDYPTEVHAADFDFKFLSAVSAVLAEK
HAYNLRVDPVNFKLLSHCFQVVLGAHLGREYTPQVQVAYDKFLAAVSAVLAEK
HAYILRVDPVNFKLLSHCLLVTLAARFPADFTAEAHAAWDKFLSVVSSVLTEK

parts of some haemoglobins

HAHKLRVGPVNFKLLSHCLLVTLA
HAHKLRVDPVNFKLLSHCLLSTLA
HAHKLRVDPVNFKLLSHCLLVTLA
HAHKLRVDAVNFKLLSHCLLVTLA
HAHKLRVDPVNFKLLSHCLLVTLA
HAHKLRVDPVNFKLLSHCLLVTLA
HAHKLRVDPVNFKLLSHCLLVTLA
HAHKLRVDPVNFKLLSHCLLVTLA
HAHKLRVDPVNFKLLSHCLLVTLA
HAHKLRVDPVNFKLLSHCLLVTLA
HAHKLRVDPVNFKLLSHCLLSTLA
HAHKLRVDPVNFKLLSHCLLSTLA
HAHKLRVDPVNFKLLSHCLLSTLA
HAHKLRVDPVNFKLLSHCLLSTLA
HAHKLRVDPVNFKLLSHCLLVTLA
HAHKLRVNPVNFKLLSHSLLVTLA
HAYKLRVDPVNFKLLSHCLLVTLA
HAQKLRVDPVNFKFLGHCFLVVVA
HAQKLRVDPVNFKFLGHCFLVVVA
HAQKLRVDPVNFKFLGHCFLVVVA
HAQKLRVDPVNFKLLGQCFLVVVA
HAYNLRVDPVNFKLLSQCIQVVL
HAYNLRVDPVNFKLLSHCFQVVL
HAYLLRVDPVNFKLLSHCLLVTLA

Consider an example column

- how many pairs do we have ?
1-2, 1-3, ... 2-3, 2-4, ...
 - gives n_{total}

For example his (H)

- count n_{HH} , n_{HY} , ..
- probability of a certain exchange
- $p_{HQ} = \frac{n_{HQ}}{n_{total}}$
- what would you expect ?

$$\frac{p_{HQ}}{n_{total}} \frac{1}{q_H q_Q} = 1$$

parts of some haemoglobins

HAHKLRVGPVNFKLLSHCLLVTLA

HAHKLRVDPVNFKLLSHCLLSTLA

HAHKLRVDPVNFKLLSHCLLVTLA

HAHKLRVDAVNFKLLSHCLLVTLA

HAHKLRVDPVNFKLLSHCLLVTLA

HAHKLRVDPVNFKLLSHCLLVTLA

HAHKLRVDPVNFKLLSHCLLVTLA

HAHKLRVDPVNFKLLSHCLLVTLA

HAHKLRVDPVNFKLLSHCLLVTLA

HAHKLRVDPVNFKLLSHCLLVTLA

HAHKLRVDPVNFKLLSHCLLSTLA

HAHKLRVDPVNFKLLSHCLLSTLÄVNLGNDFIFAVNPAJLDFKELSSVSVLISN

HAHKLRVDPVNFKLLSHCLLSTLAVHLPNDFTPAVHASLDKFLSSVSTVLTISK

HAHKLRVDPVNFKLLSHCLLSTLAVHLPNDFTPAVHASLDKFLSSVSTVLTISK

HAHKLRVDPVNFKLLSHCLLVTLAAHHPDDFNPSVHASLDKFLANVSTVLTISK

HAHKLRVNPVNFKLLSHSLVTLSASHTPTNFTPAVHANLNKFLEANDSTVLTSK

HA Y K L R V D P V N F K L L S H C L L V T L A C H H P T E F T P A V H A S L D K F F T A V S T V L T S K

HAQKLRVDPVNFKFLGHCFLVVVAIHHP SALTPEVHASLDKFLCAVGTVLTAK

HAOKLRVDPVNFKFLGHCFLVVVAIHHPALTAEVHASLDKFLCAVGTVLTAK

HAQKLRVDPVNFKFLGHCFLVVVAIHHPALTAEVHASLDKFLCAVGTVLTAK

HAOKLRVDPVNFKLLGQCFLVVVAIHNPSALTPEAHASLDKFLCAVGLVLTAK

HAYNLRVDPVNFKLLSQCIQVVLAVHMGKDYTPEVHAAFDKFLSAVSAVLAEK

HAYNLRVDPVNFKLLSHCFQVVLGAHLGREYTPQVQVAYDKFLAAVSAVLAEK

HAYILRVDPVNFKLLSHCLLVTLAARFPADFTAEAHAAWDKFLSVVSSVLTEK

positive log ? More than expected

negative log ? Less...

use

$$\log_2 \frac{p_{HQ}}{n_{total}} - \frac{1}{q_H q_Q}$$

Calculating a substitution matrix

- We have all the probabilities p_{AB} and p_{AA}
- matrix element AB is $\log_2(p_{AB})$ why $\log_2$?
- is my example enough ?
 - needs much more data so as to get good probabilities
 - real calculation does not use Hb – uses 100s of proteins

Different matrices

Degree of homology

- if two sequences are very similar most residues not changed
- longer evolutionary time – many things change

Longer evolutionary times

So far: probability of one mutation $A \rightarrow B$

- in longer evolutionary time
- $D \rightarrow E \rightarrow D \rightarrow W \rightarrow D \dots$
 - multiple mutations
 - probability of conservation is lower (diagonal elements)
 - all off-diagonal elements will be bigger
- more formally - long time p_{long} is $p \times p \times p \times \dots$

How to count for this ?

- take matrix (like blosum) and do matrix multiplication
M M M ...
- result: a set of matrices
 - PAM10, PAM20, ...
 - Blosum62, blosum80, ...

Are these different matrices useful ?

In principle, yes

- looking for similar proteins – use blosum80
- more remote ? – use blosum62
- ...
- in practice ?
- better way to find remote homologues (soon)

Now

- should you believe any of this ?

Are alignments correct ?

Is there a correct alignment ?

Are alignments correct ?

1. algorithmically
2. mathematically
3. is my model sufficient ?

Is there a correct alignment ?

1. structurally
2. biologically / evolutionarily

Details..

Are alignments correct ? (algorithm)

Yes

(Needleman and Wunsch, Smith-Waterman)

Nearly

(seeded methods – blast)

- We have a scoring scheme
(identity, blosum similarity matrix)
- We find the alignment which maximises the score

Are alignments correct ? (compare with nature)

not necessarily

- blosum matrix comes from counting mutations
 - we find the most likely substitutions

Align ANDREW and ANDRWQANDRKWSANDRWWC

ANDR-WQANDRKWSANDRWWC

ANDREW----- guess 1 [includes gap

-----ANDREW----- guess 2

-----ANDREW- guess 3

Claim – we can find the most likely

- often correct, sometimes wrong

Are alignments correct ? (numerical)

How good is "most likely" ?

- closely related sequences – very good
- distant relatives (tuna / horse myoglobin) – will have mistakes
- repetitive sequences

Are alignments correct ? (model)

my model for evolution

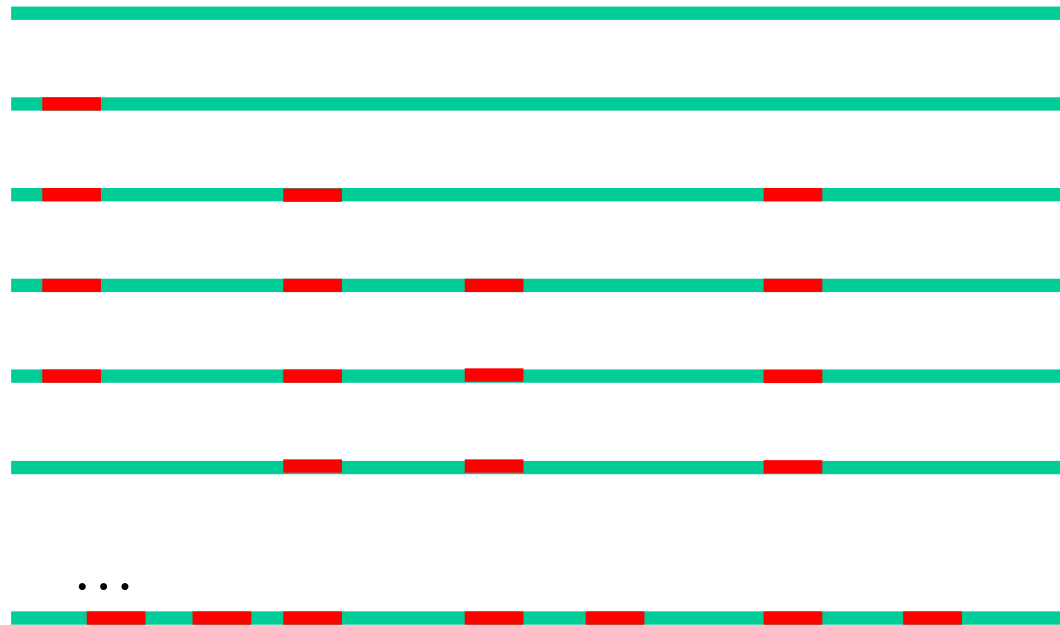
- mostly point mutations (DNA level)
- rarely a deletion/insertion

Is this enough ?

- not always
 - fusion, insertion of random DNA, splicing, duplications..

Is there a correct alignment ? evolutionary

Yes

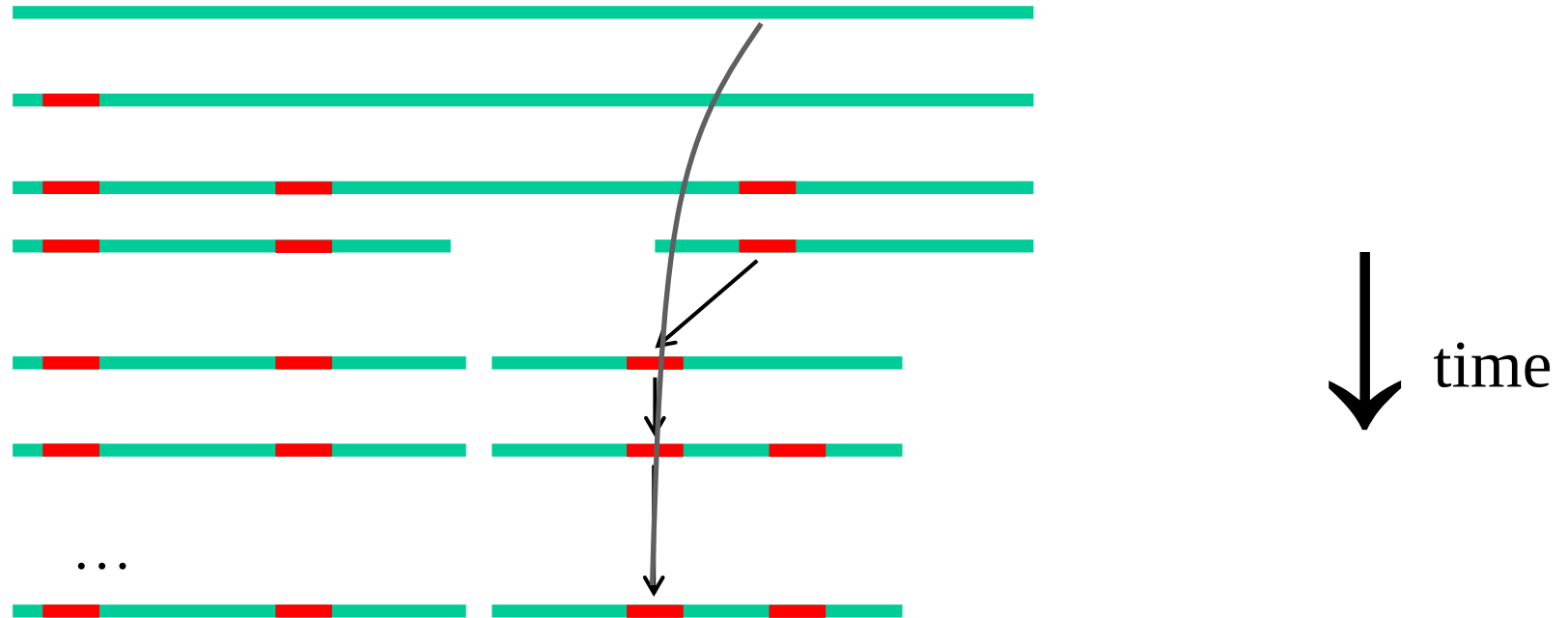


↓ time

There is a correct sequence alignment

- with gaps ?

Is there a correct alignment ? evolutionary



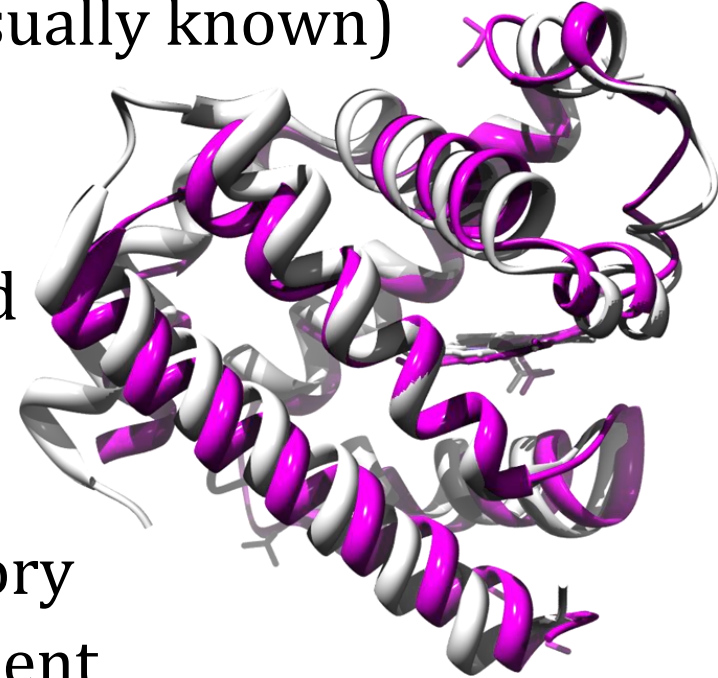
There is a correct alignment

- the evolutionary history of each residue / base

Is there a correct alignment ? (structure)

Yes

- every protein has a structure (not usually known)
- structures can be aligned
- corresponding residues can be found



This alignment may

- not correspond to evolutionary history
- not agree with best sequence alignment
(imagine deleting some residues on solvent exposed helix)

Alignments and searches

The practical goals - different problems

1. find the best alignment (modelling, predictions of important residues)
 - we care about alignment methods
2. find related known proteins (what kind of protein do I have ?)
 - we need sensitive search methods

Correct alignments - summary

There is something of a correct alignment

- evolutionary history or structure (both are good definitions)

The algorithm does maximise a score

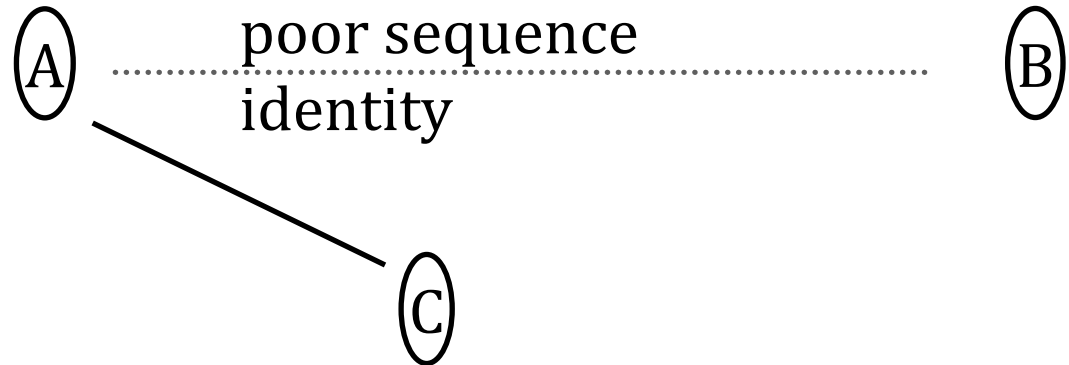
- it is numerically correct
- it will not always correspond to the correct alignment

iterated searches (psi-blast)

Search with protein A and find a very remote protein B

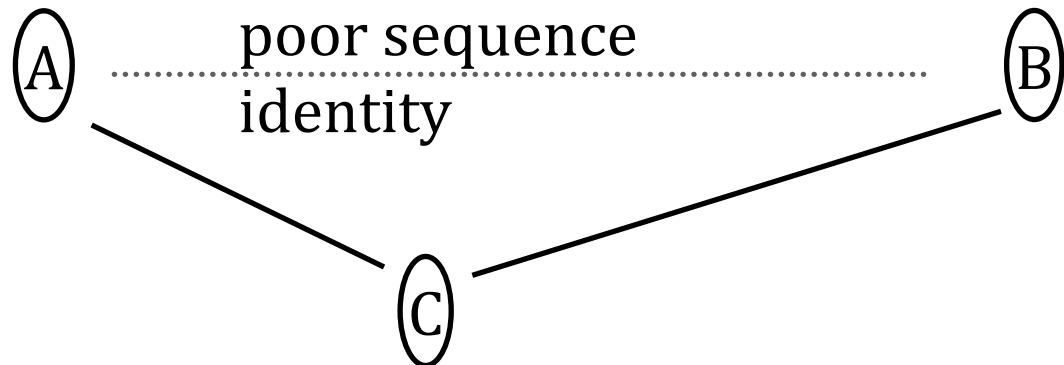


but there another protein C



search with C finds B

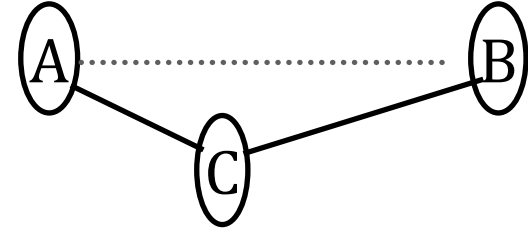
- the original AB relation is believable
- how to automate this ?



iterated searches (psi-blast)

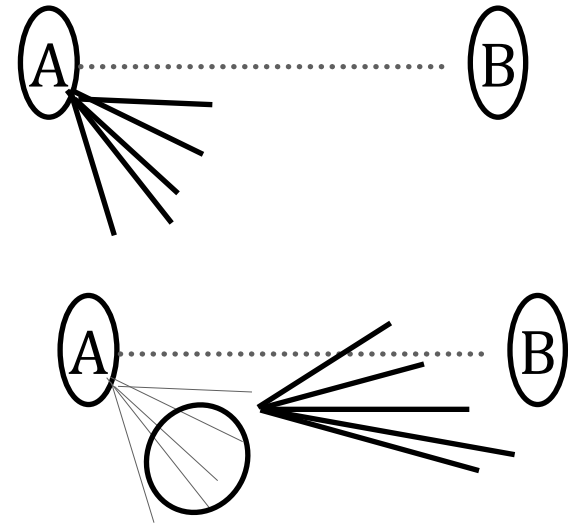
Searching with "A" finds lots of homologues

- cannot start a search with each



Alternative

- find all the homologues to A
- build an average sequence (profile)
- from this profile – repeat search
- build new average / repeat



Result

- at each step
- include reliable homologues
- eventually $A \rightarrow B$ may be found

iterated searches (psi-blast)

In practice

- really only one program (+ web page) NCBI blast / psiblast
- most significant advance in finding remote homologues in a decade

sequence identity / similarity / significance

Significance

- I find a homologue – is it evolutionarily related or just noise ?
 - probability estimations later
- how important is 10% sequence identity ? 90 % ?
- is 25 % identity in DNA as useful as in a protein ?

First principles DNA

- what would you expect by chance ?
GGATCGA
GATTCAGGTTA
 - At each position $\frac{1}{4}$ chance of a match
 - average 25 % sequence identity with random DNA
- ... very wrong

Naïve identity expectation – base usage

Two problems

1. uneven character frequency
2. gaps

Character frequency – what if I have a two letter alphabet ?

- a world with two bases

GCGGCGCGCCGCGCGCGCGCGC

average sequence identity 50 %

- a world with usually two bases - sometimes A or T

GCGACGCGTCGCGCGTTTCGCGC

average sequence identity : a bit less than 50 %

- a four letter alphabet

GCGACACGTCGTGAGTTCTTGC nearly 25 %

Naïve identity expectation – base usage

One extreme

- two types of base – 50 % sequence identity expected

Other extreme

- Random sequences – 25 % identity

Our world – somewhere between

How significant ?

- malaria is about $\frac{1}{3}$ GC (not $\frac{1}{2}$)
- *Streptomyces coelicolor* is 72 % GC
- GC differs between organisms, coding/non-coding regions

Consequence

- even randomly sampled sequences, will have > 25% sequence identity

Naïve identity expectation - gaps

- **ungapped:** 2 matches from 9 aligned (22 %)
GGATCGCAC
GACTGAGGTTA
- **one gap:** 3 matches 8 aligned (38 %)
GGATCGCAC
GACT-GAGGTTA
- **more gaps:** 4 matches from 6 positions (50 %)
GGATCGCAC
GACT-G-AGGTTA
- **more gaps:** 5 matches from 6 positions (83 %)
GGATC-GCAC
G-A-CTG-AGGTTA
- the more gaps one allows - the higher the identity
- One can make score arbitrarily good

Protein – random matches

20 amino acids

- naïve expectation – 5 %

%

ala 8.4

leu 8.3

gly 7.8

trp 1.5

cys 1.7

Proteins are not like a 20 character alphabet:

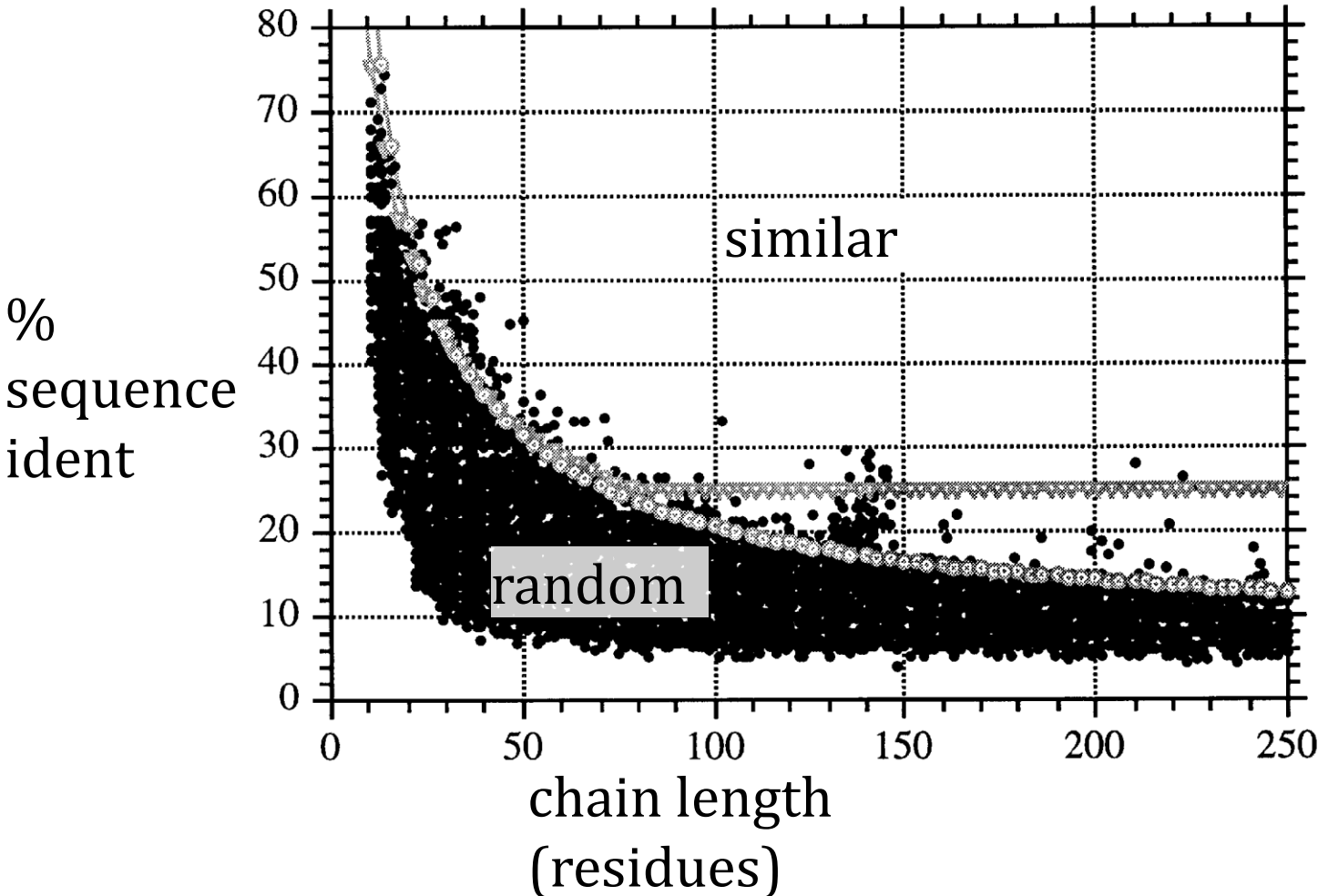
- varies between organisms
- varies between cell compartments, soluble, membrane bound...

Practical result - random sequences, realistic gaps

- 20 to 25 % identity by chance
- depends on length..

protein size and identity

- small proteins – need 30 % to believe they are related
- big proteins < 20 % , almost certainly related



Searching for significance – homologues

Try easy steps first

- simple searches first
- see if enough information is found
- gradually go to more sensitive methods (slightly more error prone)

Use the “least speculative” methods first

- accurate alignments – not seeded
- simple blast searches before iterated ones

Protein Modelling

- Where has all this been leading to ?
- Why worry about similarity ?

Mission

- You have a protein sequence
 - no structure known
- You would like to build a model for the atomic coordinates

What are the expectations ?

For easy sequences

- very good molecular models
- no doubt about function

Middle difficult

- reasonable models
 - enough to guide mutagenesis (which residues can be mutated safely)

Very difficult

- not even sure what class of proteins or what function
- may be able to suggest experiments most likely to be useful

Summarise problem and steps

Mission

- you have a protein sequence
 - no structure
 - maybe no biochemistry (substrates, binding targets, ..)
- find what you can
 - related proteins of known structure
 - related proteins with known function
- Is there
 - an answer ?
 - one set of steps ?

easy	98 % similar to protein of known function and structure
↕	
hard	weak possible similarity to a poorly characterised family

Why do protein modelling ?

- real structures (crystallography, NMR) are better
- crystallography
 - cost, crystallisation, phasing
 - think of membrane proteins
- NMR
 - limited in size, solubility
- what are the most important therapeutic targets ?
 - enzymes
 - receptors (where are they ?)
- crude models often used for crystallographic phasing

Overall scheme

For your sequence

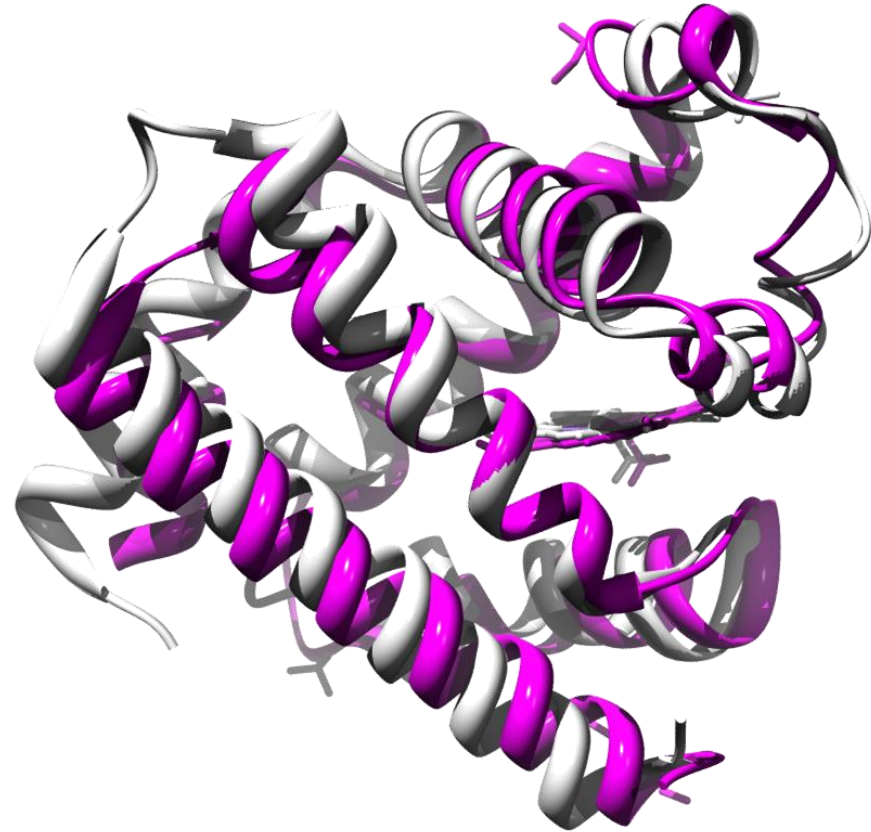
- find related proteins of known structure
 - gives you "template" structure
- sequence alignment
 - your sequence and sequence from template structure
- replace residues
 - where the residues are the same do not do much
 - where they differ, put your residues in place
- fix gaps, insertions
- fix side chains

will this work ?

What accuracy ? Examples

Tuna / horse myoglobin

- imagine you know the structure of tuna Mb
- align the sequences
- put residues from horse myoglobin onto tuna
- would make a good guess
- most atoms within 2 – 3 Å
- nasty case...



Accuracy – difficult example

Align sequences

seq id = 11 %

What would a model
look like ?

Alignment to 1v93A

Seq ID 11 % (25 / 227) in 268 total including gaps

```
: 5 : 6 : 7 : 8 : 9 :
: 0 : 0 : 0 : 0 : 0 :
afvsitygam-gstrersvawa-----qriqslglnplahlrtvaggsrkevaevlhrfv
rrpsvvyhlhnaectgcseavlrafepyidtlildtllsldyhetimaaagdaaeaaaleqav
: 1 : 2 : 3 : 4 : 5 : 6
: 0 : 0 : 0 : 0 : 0 : 0

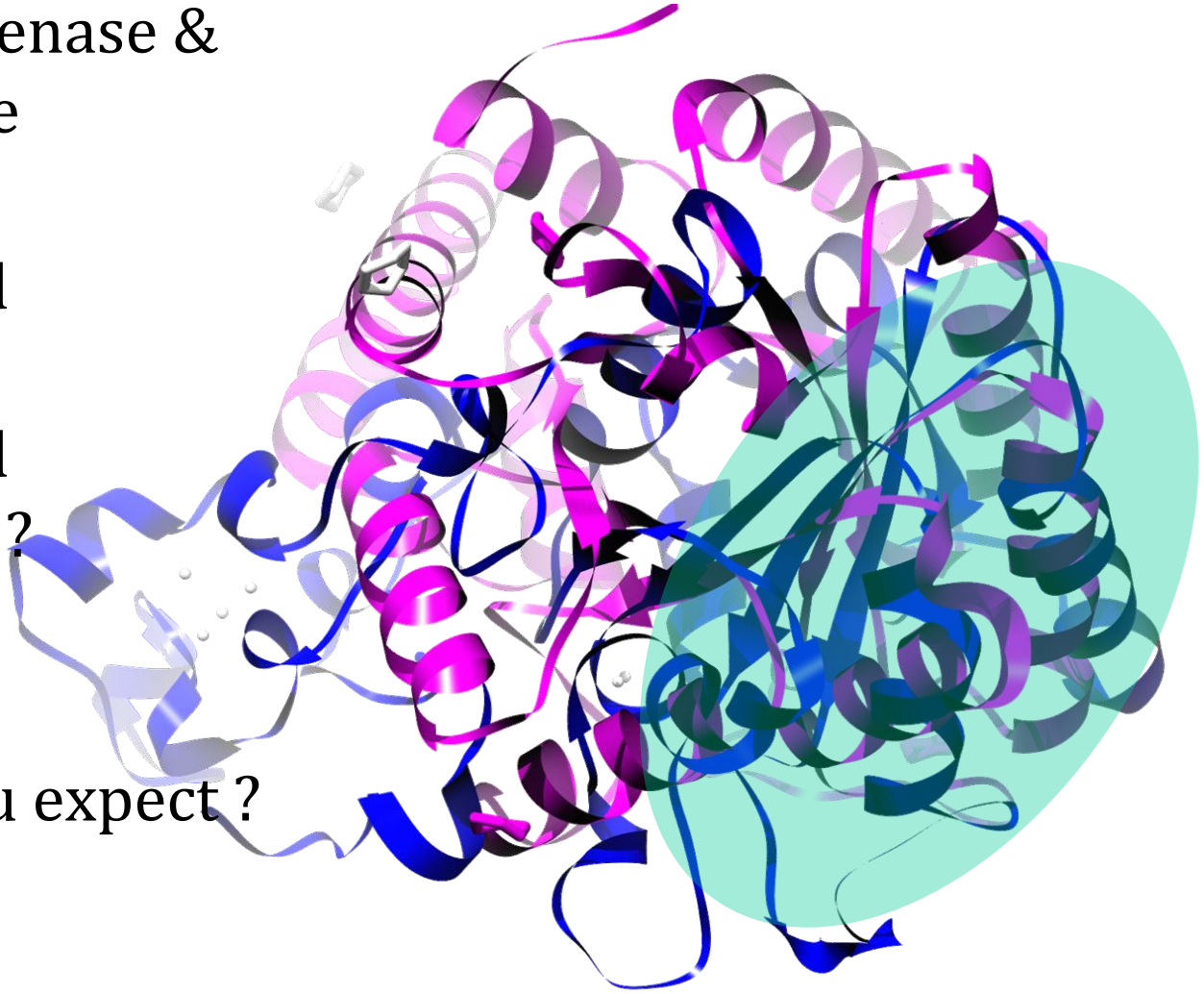
1 : 1 : 1 : 1 : 1 : 1 :
0 : 1 : 2 : 3 : 4 : 5 :
0 : 0 : 0 : 0 : 0 : 0 :
esgvenllalrgdpprgervfrphpegfryaaelvalirerygdrsvsggaaype-ghpe
nsphgfiavveggiptaangiygkvanh-tmldicsrilpka--qaviaygtcatfggvq
: 0 : 0 : 0 : 1 : 1 : 1
: 7 : 8 : 9 : 0 : 1 : 2
: 0 : 0 : 0 : 0 : 0 : 0

1 : 1 : 1 : 1 : 2 :
6 : 7 : 8 : 9 : 0 :
0 : 0 : 0 : 0 : 0 :
sesleadlr--hfkakveagldfa-itqlffnnahyfgflerarragigipil-----p
aakpnptgakgvndalkhlgvkainiagcppnpynlvgtivvylnkaapeldslnrptm
: 1 : 1 : 1 : 1 : 1 : 1
: 3 : 4 : 5 : 6 : 7 : 8
: 0 : 0 : 0 : 0 : 0 : 0

2 : 2 : 2 : 2 : 2 : 2 :
1 : 2 : 3 : 4 : 5 : 6 :
0 : 0 : 0 : 0 : 0 : 0 :
gimpvtsyrqlrrftevcgasipgpllaklerhqddpkavleigvehavrqvaelleagv
ffgqtvheqcprlphfdagefa-----psfeseeark-----gwclyelgc
: 1 : 2 : 2 : 2
: 9 : 0 : 1 : 2
: 0 : 0 : 0 : 0
```

Accuracy – difficult example

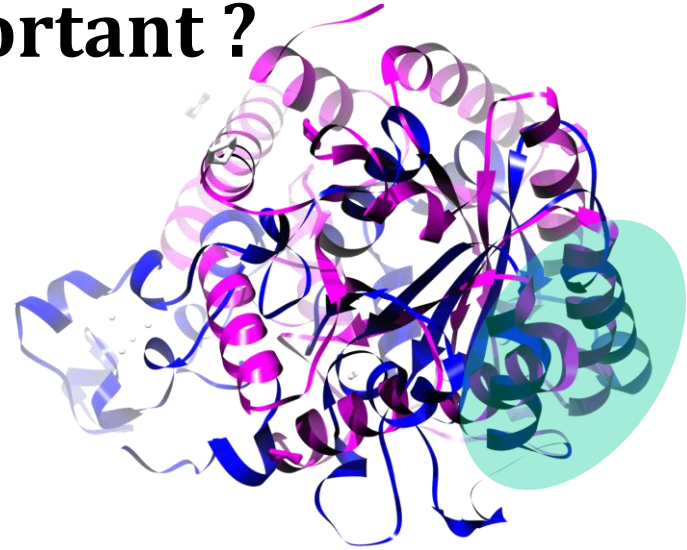
- 1ubr & 1v93
- Fe / Ni hydrogenase & oxidoreductase
- would you find this template ?
- would you find this alignment ?
- what could you expect ?



model quality – important ?

Do I need a good model ?

- design binding molecules ? – yes
- phasing in crystallography – no



Important question

- is this residue near the active site ? (chemical modification)
- accuracy not necessary
 - enzyme immobilization

Site directed mutagenesis + selection

- which are the critical residues to target ?

Expectations

	easy		hard
sequence identity	80-90 %		< 15 %
template	no problem	no problem	sometimes wrong
alignment	no errors	some parts wrong	some parts cannot be aligned
gaps / loops	very few		terrible
uses	designing ligands		predicting active sites
			mutagenesis

Relate to previous lectures

For your sequence – find a template

- if you cannot find it with blast / fasta – will be difficult

For many sequences – many templates equally good

Why all the talk about psi-blast / related sequences ?

- your protein may not have any close homologues

Template found - what next ?

alignment for modelling

Easy cases (sequence homologous to template)

- blast alignment OK
- any alignment OK

Harder cases

- use the best (slowest) alignment program
 - will not do any harm
- costs human time (computer time is insignificant)

insertions and gaps

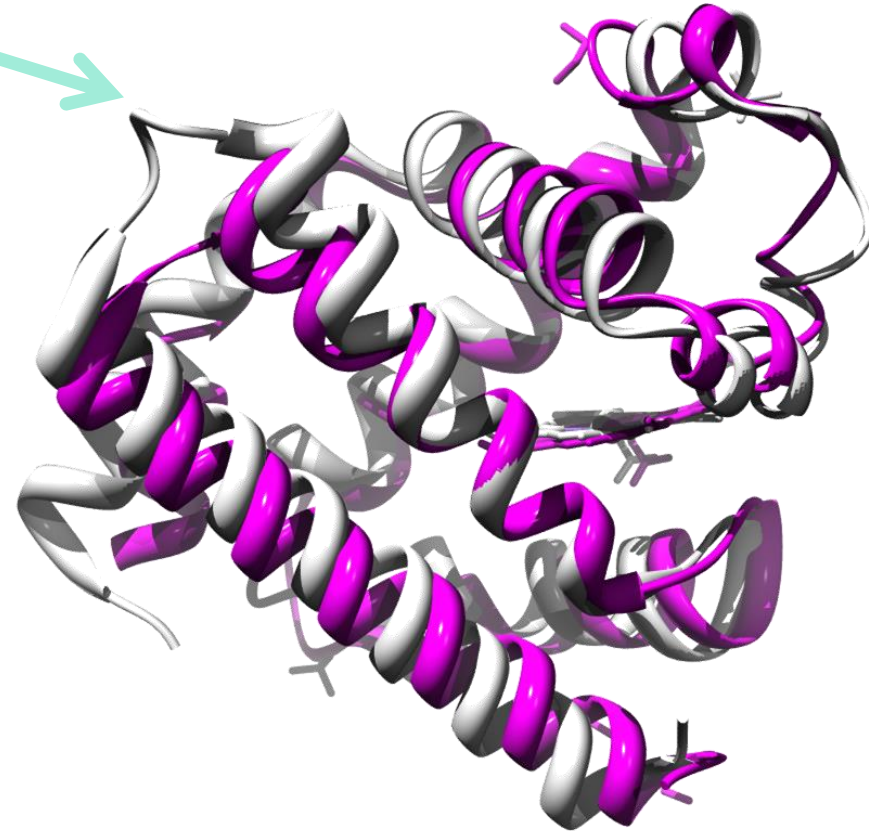
- dogma – gaps and insertions are less likely in regular secondary structure (α -helices, β -strands)
- more likely in "loops"

```

      5       6       7       8       9
      0       0       0       0       0
afvsitygam-gstrersvawa-----griqslglnplahltvaggsrkevaevlhrfv
rrpsvvylnhaectgcsesvlrafepyyidtlildtllsldyhetimaaagdaaeaaleqav
      1       2       3       4       5       6
      0       0       0       0       0       0

      1       1       1       1       1       1
      0       1       2       3       4       5
      0       0       0       0       0       0
esgvenllalrgdpprgervfrphpegfryaaelvalirerygdrvsvggaaype-ghpe
nspbgfiavveggiptaangiygkvanh-tmldicsrilpka--qaviaygtcatfggvq
      0       0       0       1       1       1
      7       8       9       0       1       2
      0       0       0       0       0       0

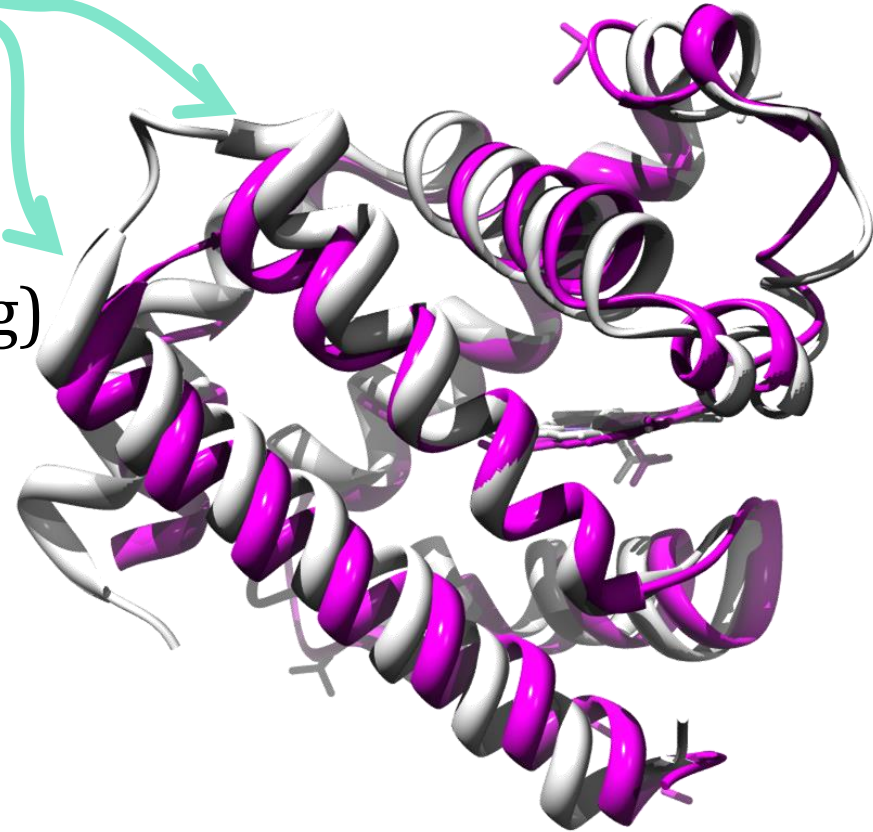
      1       1       1       1       2
      6       7       8       9       0
      0       0       0       0       0
sesleadlr--hfkakveagldfa-itqlffnnahyfgflerarragigipil-----p
aakpnptgakgvndalkhlgvkainiagcppnpynlvgtivyylnknaapeldslnrptm
      1       1       1       1       1       1
      3       4       5       6       7       8
      0       0       0       0       0       0
  
```



insertions and gaps

- imagine white is unknown, but pink is template
- where to put white loop residues ?

- fix end points
- join up backbone so as to keep reasonable geometry (bonds, angles)
 - distance geometry (in Übung)

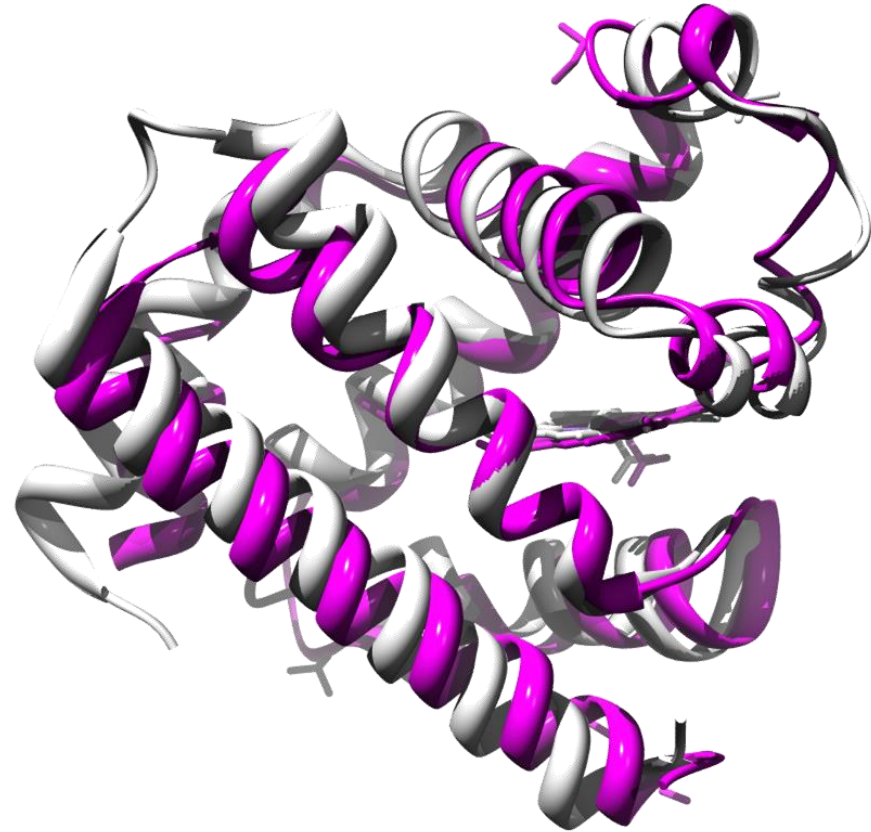


- OK ? Just a guess
- Better ?

insertions and gaps

Generate many (10^2 or 10^3) guesses for loop

Calculate energy of each guess



Sidechains

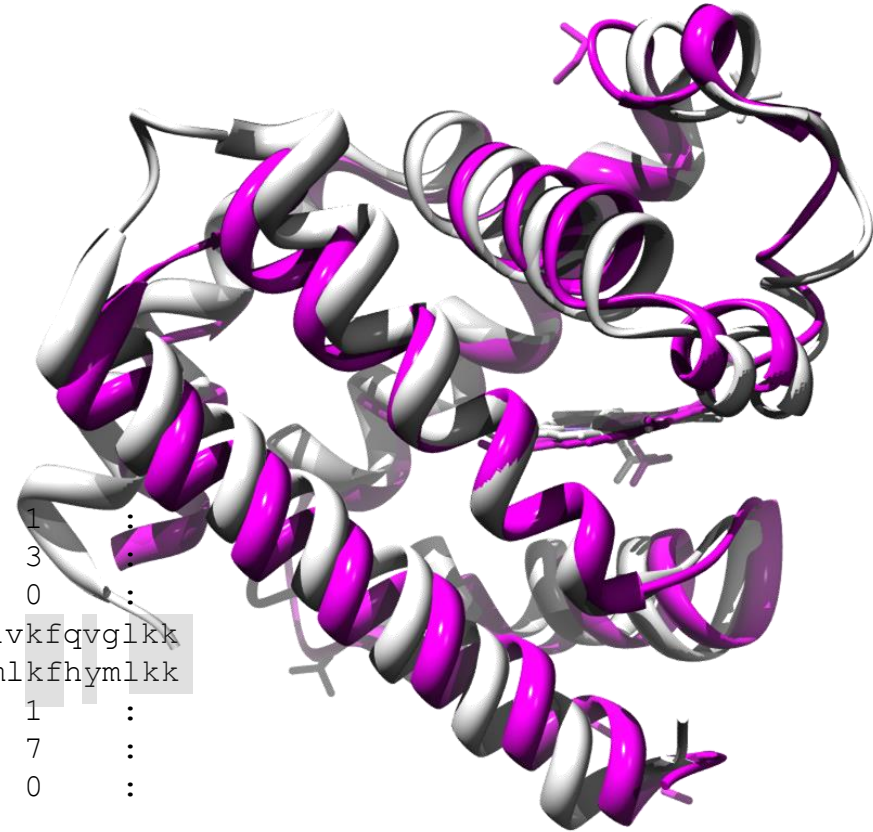
If my white one is the model

- where do we put sidechain atoms ?

Good strategy

- look at alignment
- find unchanged residues
- take sidechain coordinates
- rotate other sidechains to fit
 - energy minimization
(in Übung erwähnt)

0	:	0	:	1	:	1	:	1	:	1	:
8	:	9	:	0	:	1	:	2	:	3	:
0	:	0	:	0	:	0	:	0	:	0	:
lkssaieii	mlrs	nqsf	sled	mswsc	ggpdf	kycind	vtkag	htlelle	plvkf	qvglkk	
lkgaa	felcql	rntvf	naetg	twecg	---	rlsyc	ledta	ggfqq	lllep	mlkfh	ymllkk
:	1	:	1	:	1	:	1	:	1	:	
:	3	:	4	:	5	:	6	:	7	:	
:	0	:	0	:	0	:	0	:	0	:	



Summarise protein modelling

finding a template

wrong template – rest of procedure is wrong

alignments

usually some residues are not perfect

fixing gaps and insertions

really a guess as to coordinates

placing sidechains

wirkstoff Entwurf – vital

rough guide to essential residues – may not matter