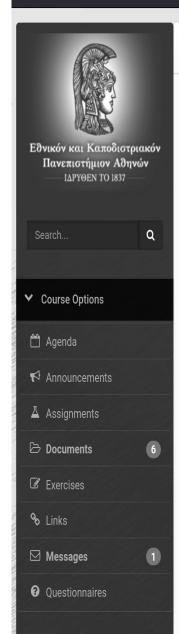
Bioinformatics overview + sequence alignment

Martin Reczko

Staff research scientist professor level
Institute for Fundamental Biomedical Science
Biomedical Sciences Research Center "Alexander Fleming"
Head of Node - ELIXIR-Greece

□ dev □ sci



♠ Portfolio / Introduction to Bioinformatics

Introduction to Bioinformatics (M413)

Martin Reczko - Alexandros Dimopoulos



Description



The course introduces students into the basic concepts of bioinformatics. It starts with a general overview of the various fields of bioinformatics and introduces dynamic programming as a solution to the sequence comparison problem (1). Next, a first introduction to the GNU / Linux operating system and the hands on use of basic command-line commands (CLI) as well as bash scripting is given. In addition, basic bioinformatics command line programs such as bedtools, vcftools, samtools, etc. are presented and used (2+3). Students are then familiarized with the programming language R, the use of IDE RStudio and the basic tools provided by the Bioconductor repository (4+5). Next, detailed examples of

NGS bioinformatics analysis and pipelines are explained for:

- RNAseq (quality control, gene expression analysis) (6),
- denovo assembly (both on the genome and transcriptome level) (7)
- ChipSeq, ClipSeq and (8)
- variant calling (exome sequencing example using GATK) (9)

Finally, the concept of flux

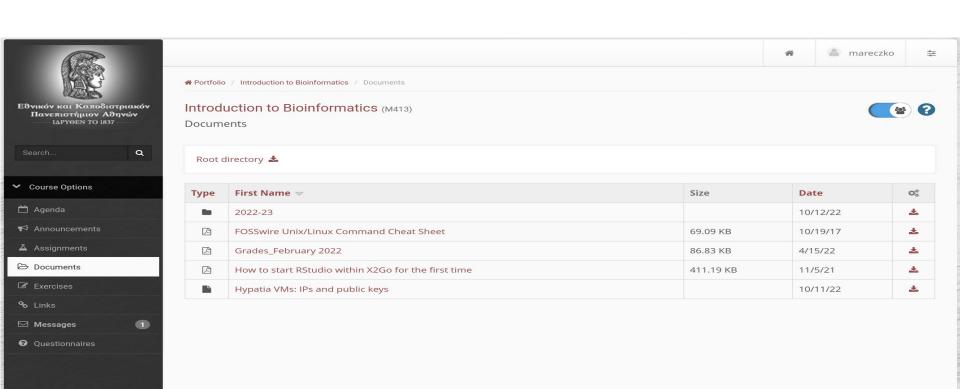




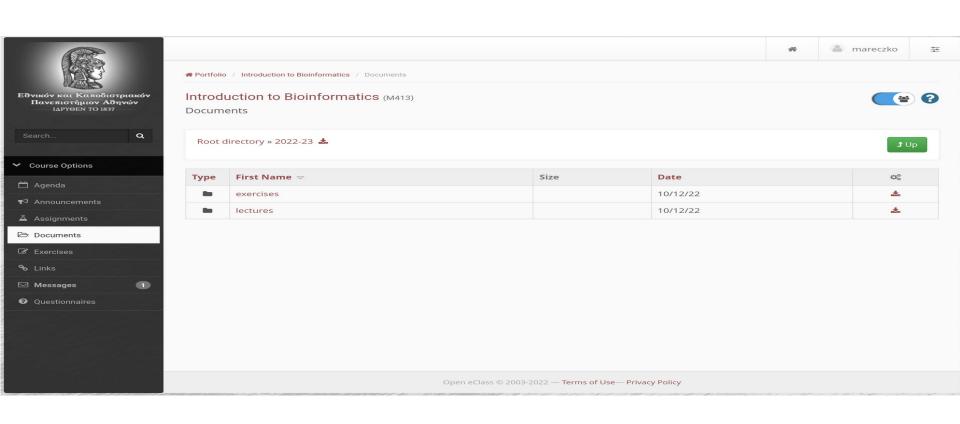








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Please verify name+email in participant list at

https://tinyurl.com/3hm8ze43

Enter all emails you might use (to get an account on the virtual machine from



28 CPUs, 242GB RAM, 800 GB disk shared for all



Syllabus and grading

#	Date	Short title	Lecturer	Other actions	Subject
1	Tuesday, October 10, 2023	introduction	MR		Overview of Bioinformatics, sequence alignment
2	Tuesday, October 17, 2023	Linux/shell/ssh	AD		Introduction to Linux and the command line, bash scripting and ssh
3	Tuesday, October 24, 2023	R (1)	AD		Introduction to the R programming language and Rstudio usage
4	Tuesday, October 31, 2023	QC+RNASeq	MR		Next generation sequencing: introduction, quality control and gene expression analysis for RNAseq
5	Tuesday, November 7, 2023	R (2)	AD		Advances R subjects, introduction to Bioconductor
6	Tuesday, November 14, 2023	bedtools/vcftools/samtools	AD		Command line tool usage: bedtools, vcftools, samtools etc.
7	Tuesday, November 21, 2023	Denovo	MR		NGS for denovo genome and transciptome assembly
8	Tuesday, November 28, 2023	ChipSeq/chirp	MR	assign presentations	NGS analysis for molecular interactions (ChipSeq, (Par-)Clip, structural sequencing, chromosome conformation capture (3C))
9	Tuesday, December 5, 2023	metabolomics	MR		Genome-scale models of metabolism and macromolecular expression, Biological applications of Transformers
10	Tuesday, December 12, 2023	Exome/SNP calling	AD	assign final projects	Pipelines for SNP calling, especially for exome sequencing using the GATK pipeline
11	Tuesday, December 19, 2023	presentations	MR+AD		Paper presentations by students
12	Tuesday, January 9, 2024	presentations	MR+AD		Paper presentations by students
13	Tuesday, January 16, 2024	final projects support	MR+AD		Support for the final project

Grade	100%		
Presentation	30%		
Exercises	20%		
Final Project	50%		

Subjects:

'Just enough' biology
Dynamic programming
Approximate string similarity
Bioinformatics fields
Recent machine learning results

An Eukaryotic Cell (biological view)

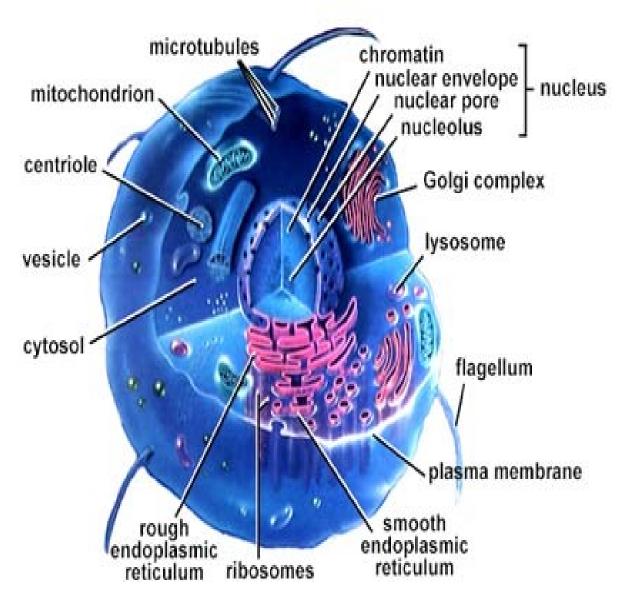
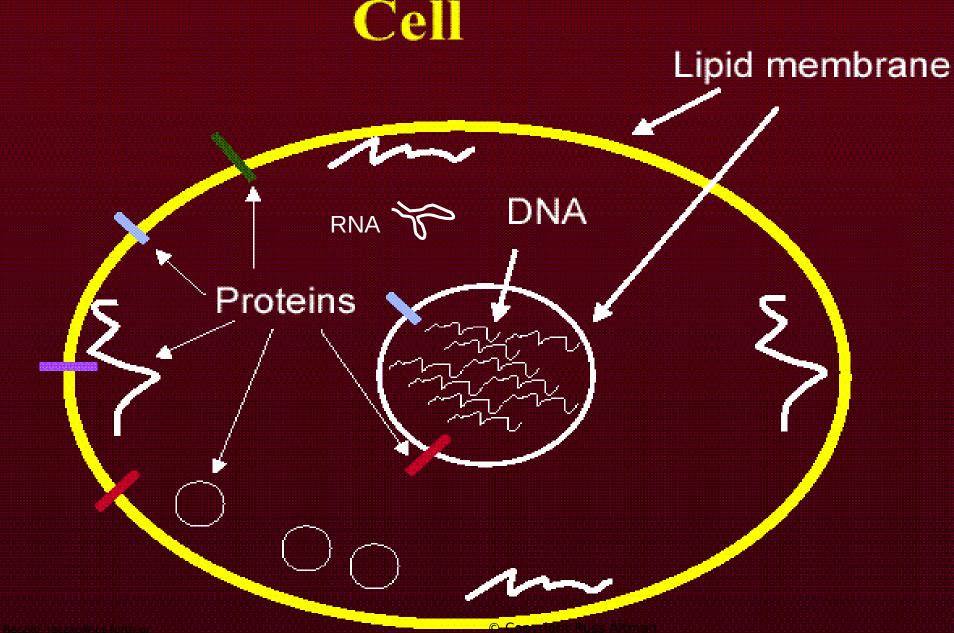
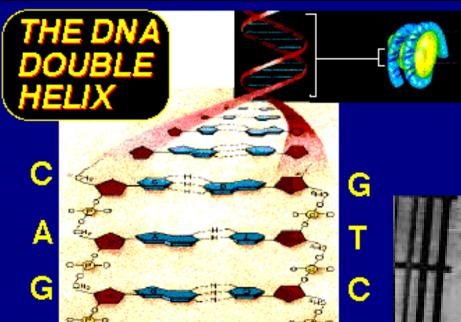


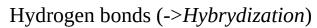
Image from: On-Line Biology Book

Bioinformatics Schematic of a Cell



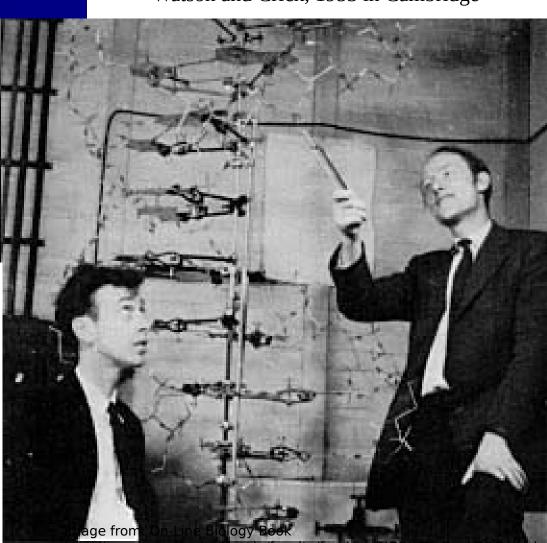


Watson and Crick, 1953 in Cambridge

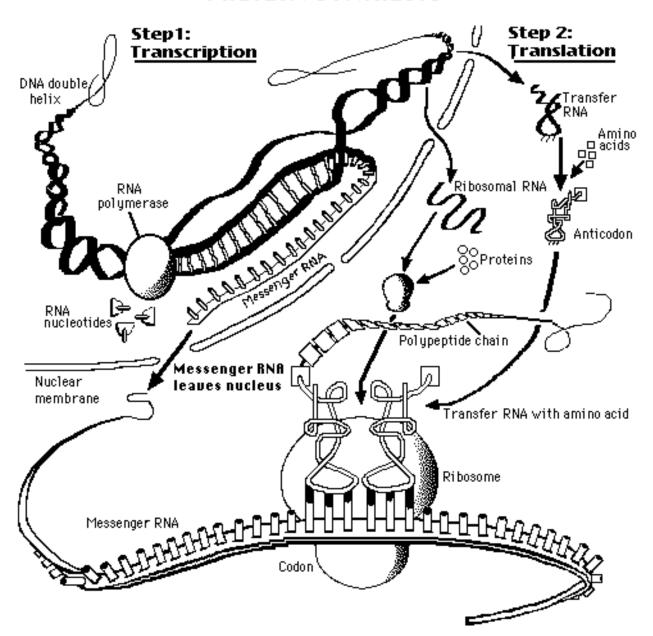




Rosalind Franklin



PROTEIN SYNTHESIS



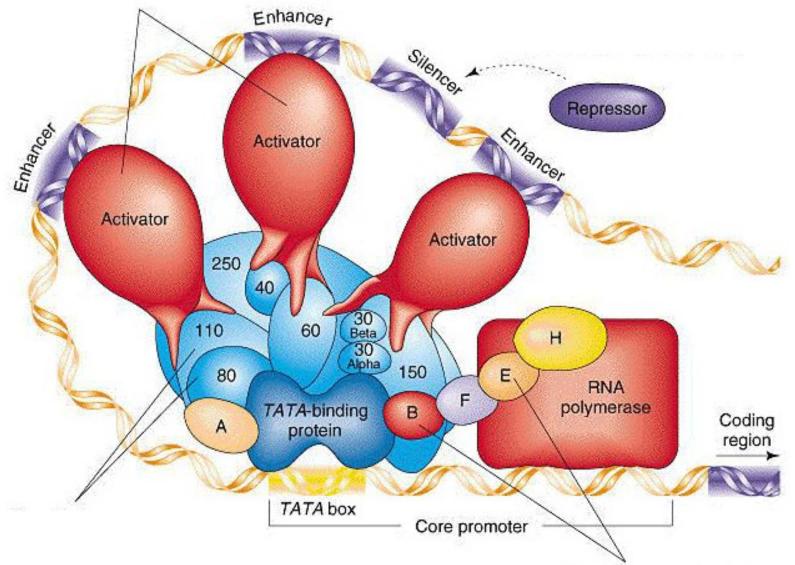
Transcription

transcription is accomplished by RNA polymerase

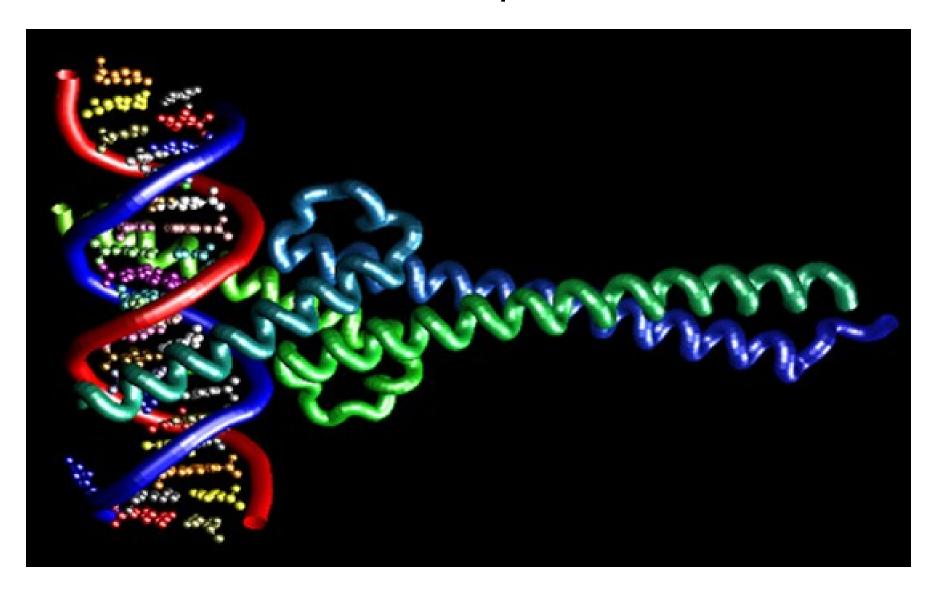
RNA polymerase binds to **promoters** promoters have distinct regions "-35" and "-10" transcription start and stop affected by DNA structure

Additional regulatory sequences can be positive or negative

Complete Assembly of Eukaryotic Gene Regulatory System



Interaction of a transcription factor and DNA



Myc Proto-Oncogene Protein, causing cell division and proliferation

Transcription: DNA -> RNA

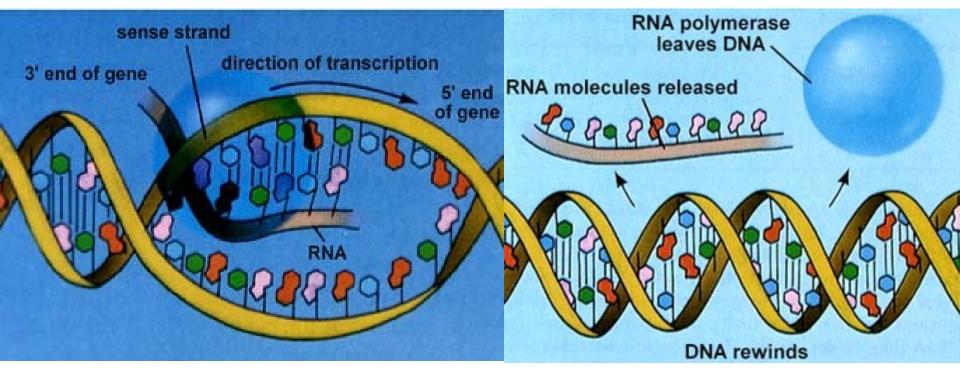


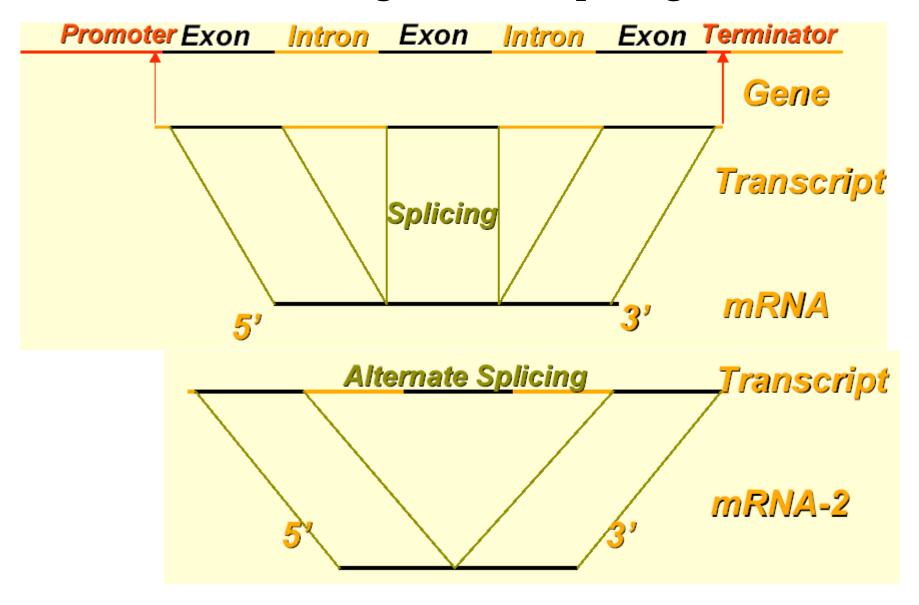
Image from: On-Line Biology Book http://www.emc.maricopa.edu/faculty/farabee/BIOBK/BioBookTOC.html

RNA processing

eukaryotic genes are interrupted by introns

these are "spliced" out to yield final messenger RNA (mRNA) splicing done by spliceosomes splicing sites are quite degenerate but not all are used

Processing of RNA = splicing

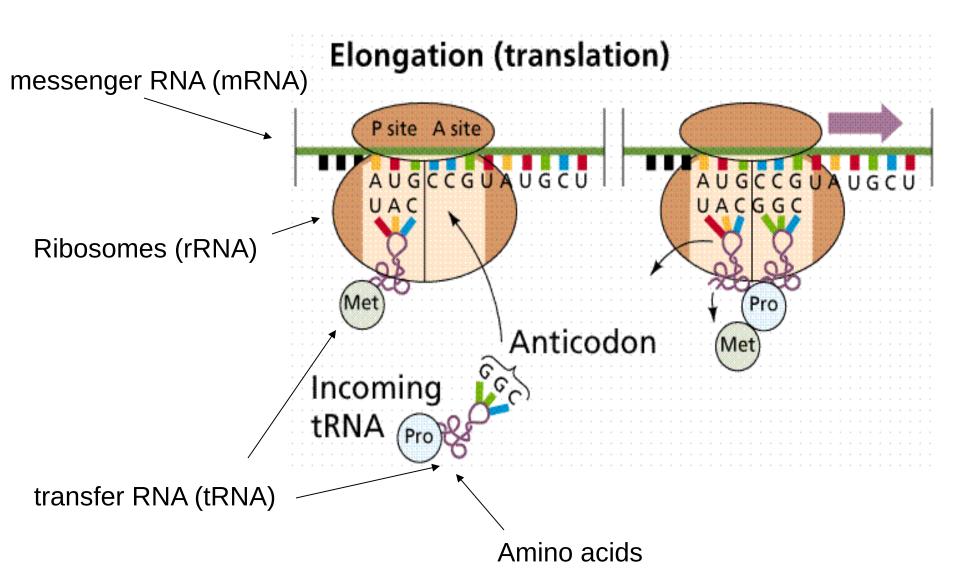


Images from: http://biochem218.stanford.edu (Doug Brutlag)

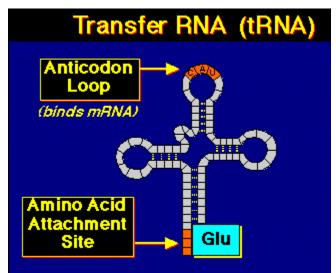
Translation

conversion from RNA to protein is by codon: 3 bases = 1 amino acid translation done by ribosome translation stops after reading the stop codon

Building proteins:



The 'universal' genetic code:



64 different transfer RNA molecules



Second letter

		U		С		Α		G		
	U	UUU	Phenyl- alanine	UCU UCC UCA UCG	Serine	UAU UAC	Tyrosine	UGU UGC	Cysteine	O
		UUA UUG	Leucine			UAA UAG	Stop codon Stop codon	UGA	Stop codon Tryptophan	A G
e	c	CUU	CONTRACTOR OF THE PROPERTY OF	CCU CCC CCA CCG	Proline	CAU	Histidine	CGU CGC	Arginine	O C
lett		And the second second second second				CAA CAG	Glutamine	CGA	CGA	A G
FIrst	А	AUU	Isoleucine	ACU ACC ACA ACG	Threonine	AAU AAC	Asparagine	AGU AGC	Serine	0 0
Ξ		AUA	Methionine; initiation codon			AAA AAG	Lysine	AGA AGG	Arginine	A G
	G	GUU GUC GUA GUG	GCU GCC GCA GCG	Alanine	GAU GAC	Aspartic acid	GGU GGC	Glycine	U C	
					GAA GAG	Glutamic acid	GGA GGG	Joycine	A G	

Pictures taken from On-Line Biology Book

The 20 amino acids, building blocks for proteins

Tyrosin (Tyr)

Valin (Val)

Tryptophan (Trp)

Threonin (Thr)

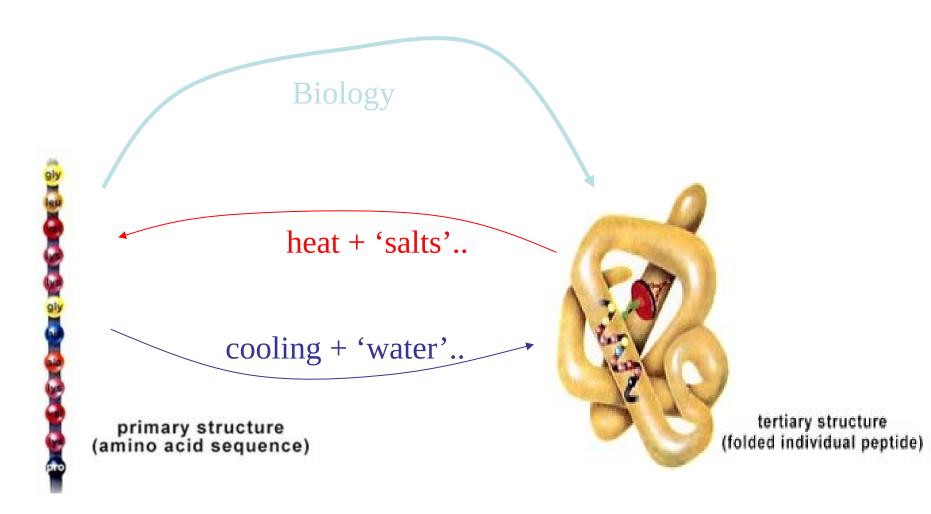
Building proteins (chemistry):

Amino acids are linked together by joining the amino end of one molecule to the carboxyl end of another. Removal of water allows formation of a type of covalent bond known as a <u>peptide bond</u>.

The above image is from

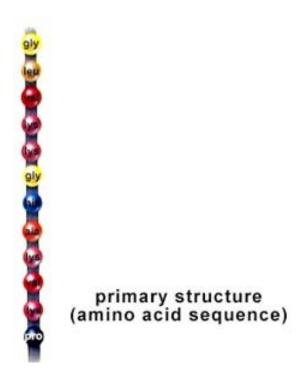
http://zebu.uoregon.edu/internet/images/peptide.gif.

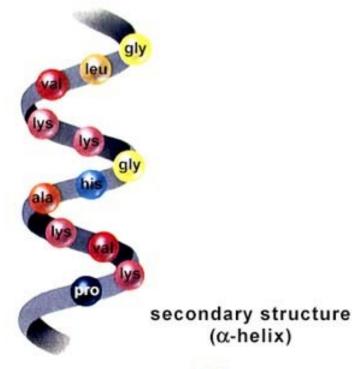
Protein folding: Sequence determines structure

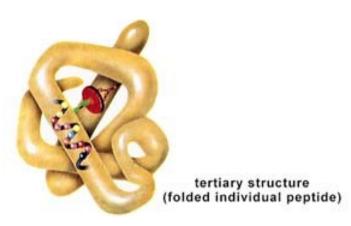


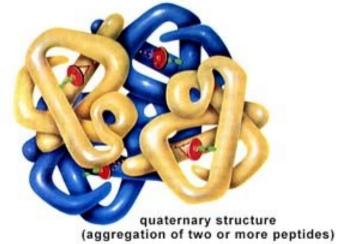
C. Anfinsen, 1973

Levels of structural description









The above images are from

http://www.biosci.uga.edu/almanac/bio 103/notes/may 14.html.

Protein localization

leader sequences can specify cellular location (e.g., insert across membranes) leader sequences usually removed by cleavage

Like an address sticker

Protein localization

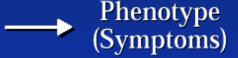
protein traffic compartments CYTOSOL NUCLEUS PEROXISOME endosome cytosol MITOCHONDRIA **PLASTIDS** lysosome Golgi apparatus ENDOPLASMIC RETICULUM peroxisome mitochondrion endoplasmic reticulum GOLGI with membrane-bound polyribosomes LYSOSOME SECRETORY free nucleus polyribosomes VESICLES ENDOSOME plasma membrane - 15 µm ----CELL SURFACE = gated transport UNFOLDED PROTEIN FOLDED PROTEIN = transmembrane transport = vesicular transport NH_2 signal СООН peptide (A)

Central Paradigm of Bioinformatics

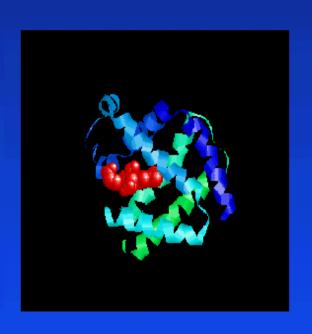


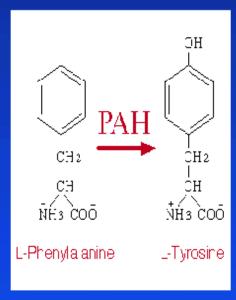


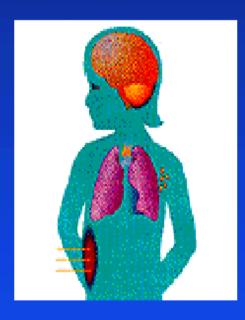




TGCTTTAGCTTT
AAACTACAGGCC
TCACTGGAGCTA
GAGACAAGAAGG
TAAAAAAACGGCT
GACAAAAGAAGT
CCTGGTATCCTC
TATGATGGGAGA
AGGAAACTAGCT
AAAGGGAAGAAT
AAATTAGAGAAA
AACTGGAATGAC
GCTTATACCTGG



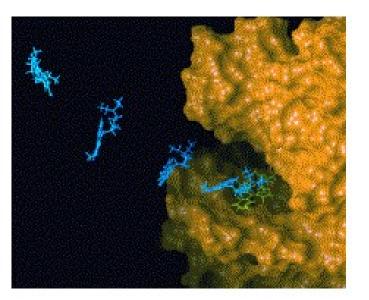


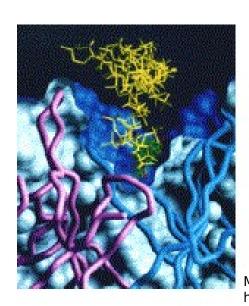


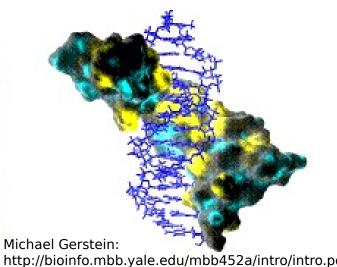
Protein/Ligand interactions:

- Understanding How Structures Bind Other Molecules (Function)
- Designing Inhibitors
- Docking, Structure Modeling

(From left to right, figures adapted from Olsen Group Docking Page at Scripps, Dyson NMR Group Web page at Scripps, and from Computational Chemistry Page at Comell Theory Center).







Information flow

A major task in computational molecular biology is to "decipher" information contained in biological sequences Since the nucleotide sequence of a genome contains all information necessary to produce a functional organism, we should in theory be able to duplicate this decoding using computers

Data growth of EMBL-EBI services volume of data (megabytes) 1.000.000.000.000 10.000.000.000 100,000,000 1,000,000 —Genomic sequences —Mass spectrometry Functional genomic —Metabolomics experiments —Human clinical, genomic and phenome Macromolecular structures

—Imaging data

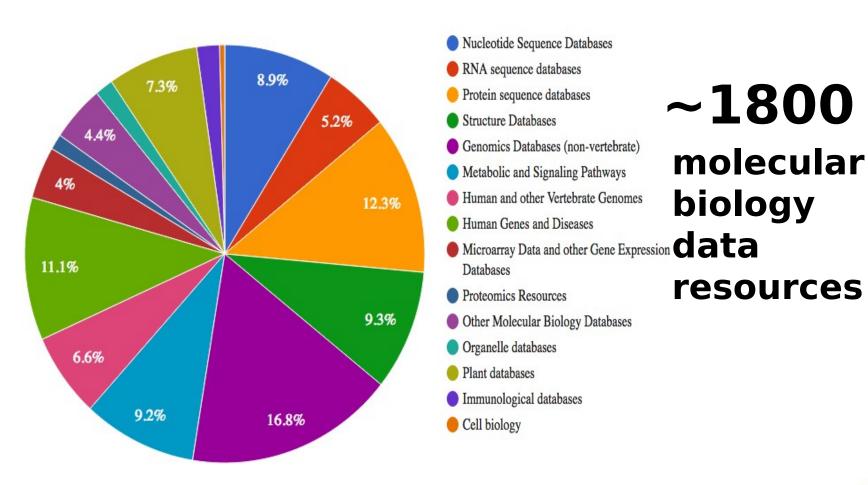
Data growth in the life sciences

- Computer speed and storage capacity is doubling every 18 months and this rate is steady (Moore's law)
- The amount of life science data doubles every 12 months and the growth rate is predicted to continue

Cantelli et al. The European Bioinformatics Institute (EMBL-EBI) in 2021, Nucleic Acids Research, Volume 50, Issue D1, 7 January 2022, Pages D11–D19



Data resources in life sciences







Incoming data size classes:

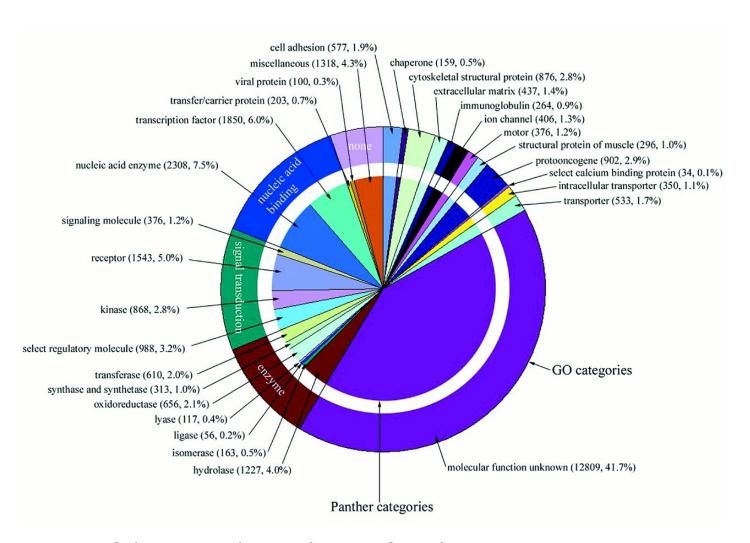
Organism	Number of chromosomes	Genome size in base pairs
<u>Bacteria</u>	1	~400,000 - ~10,000,000
<u>Yeast</u>	12	14,000,000
Worm	6	100,000,000
Fly	4	300,000,000
Weed	5	125,000,000
<u>Human</u>	23	3,000,000,000

Only the surface is scratched:

Organism	The number of predicted genes	Part of the genome that encodes proteins (exons)
E.Coli (bacteria)	5000	90%
Yeast	6000	70%
Worm	18,000	27%
Fly	14,000	20%
Weed	25,500	20%
<u>Human</u>	30,000	< 5%

*A. Brazma et. al.: http://www.ebi.ac.uk/microarray/biology_intro.htmlml *Alien finds a broken hard-disk*, situation

The function of human genes



42 % of the genes has unknown function, even having accurate predicted protein structures (AlphaFold2)

From Genomics to Drugs

Thomas I engauer (Fd)

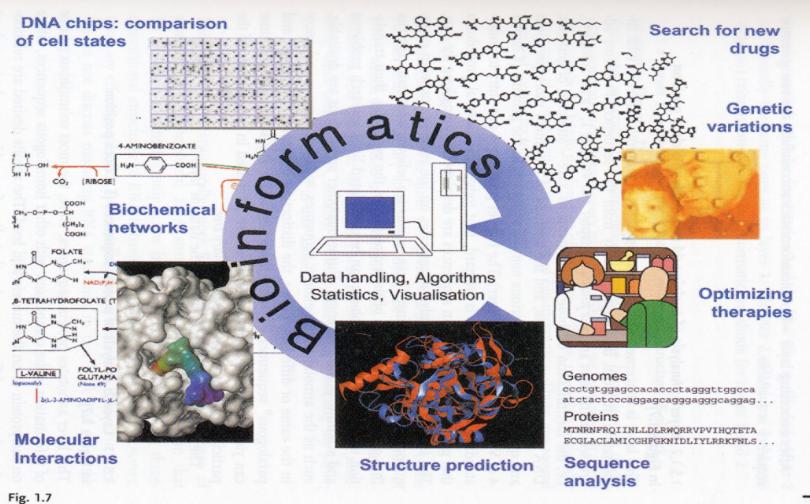
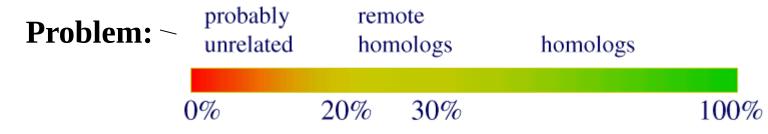


Fig. 1.7
A schematic overview of bioinformatics

A bioinformatics landscape

Homology Modeling

- observation: proteins with similar sequences tend to fold into similar structures
- given: a query sequence Q, database of protein structures
- do:
 - find protein P such that
 - structure of P is known
 - P has high sequence similarity to Q
 - return P's structure as an approximation to Q's structure



pairwise sequence identity

Basic biological sequence analysis:

Exact string matching:

- -Boyer Moore string search algorithm (UNIX: grep)
- suffix trees

Inexact string matching:

- Complete sequence (global) or parts (local)
- Similarity measures

Pairwise vs. multiple comparisons

Aligning Text Strings

```
Raw Data ???
 TCATG
   C A
           T G
                        4 matches, 1 insertion
2 matches, 0 gaps
    C A
                        4 matches, 1 insertion
3 matches (2 end gaps)
            G
                      Ambiguity:
```





Definitions

Global alignment

INPUT: Two sequences S and T of roughly the same length.

QUESTION: What is the maximum similarity between them? Find a best alignment.

Local alignment

INPUT: Two sequences S and T.

QUESTION: What is the maximum similarity between a subsequence of S and a subse-

quence of T? Find most similar subsequences.

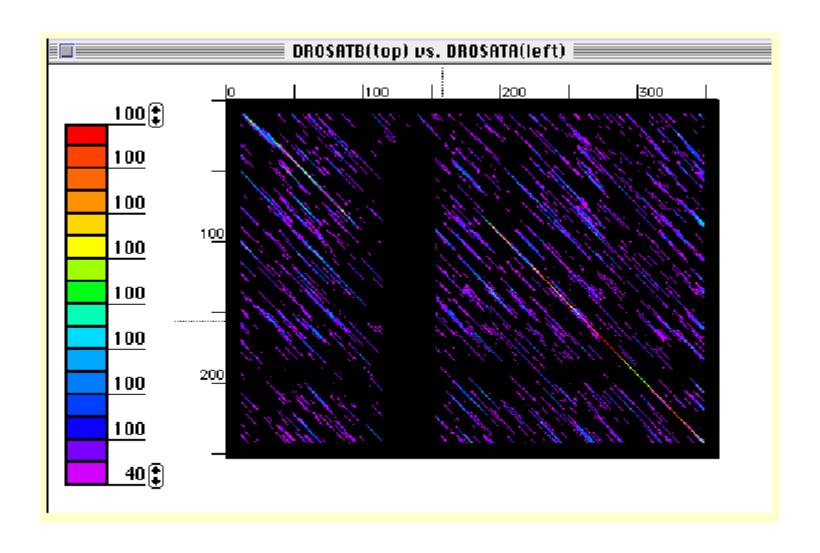
Definition A gap is the maximal contiguous run of spaces in a single sequence within a given alignment. The length of a gap is the number of indel operations on it. A gap penalty function is a function that measures the cost of a gap as a (nonlinear) function of its length.

Gapped alignment

INPUT: Two sequences S and T (possibly of different length).

QUESTION: Find a best alignment between the two sequences using the gap penalty function.

Graphical solution: dot-plot

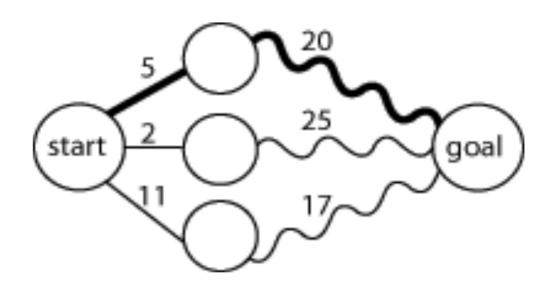


Dynamic programming algorithms for sequence comparison

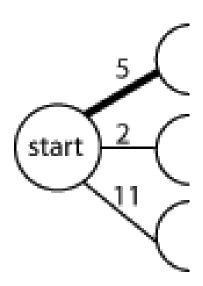
Introduced for biological sequences by

S. B. Needleman & C. D. Wunsch. A general method applicable to the search for similarities in the amino acid sequence of two proteins. *J. Mol. Biol.* 48:443-453 (1970)

Dynamic programming reminder: Shortest path



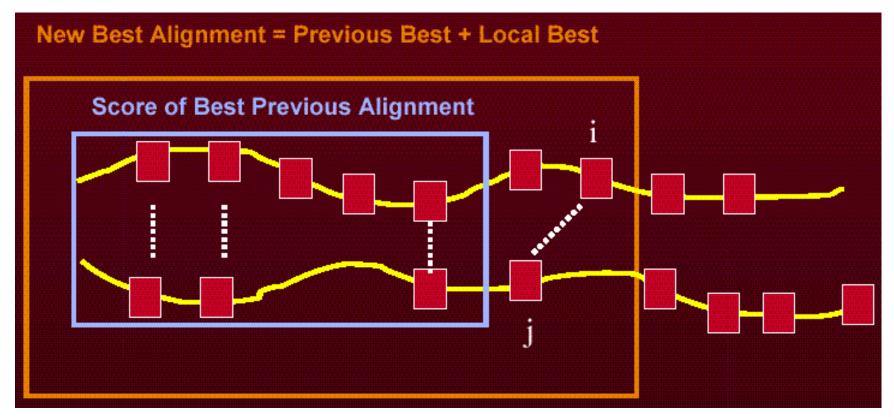
Dynamic programming reminder: Shortest path



Best solutions up to n

One node added: n updates to find new best

Dynamic Programming Idea:



© Copyright Russ Altman 2001,http://smi-web.stanford.edu/projects/helix/bmi214/4-4-02clr.pdf

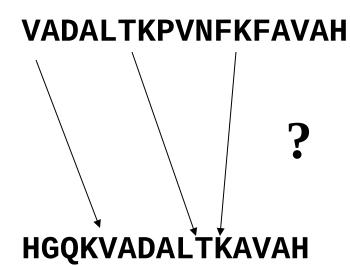
Key Idea in Dynamic Programming

- The best alignment that ends at a given pair of positions (i and j) in the 2 sequences is the score of the best alignment previous to this position PLUS the score for aligning those two positions.
- ♦ An Example Below
 - Aligning R to K does not affect alignment of previous N-terminal residues. Once this is done it is fixed. Then go on to align D to E.
 - o How could this be violated?
 Aligning R to K changes best alignment in box.

ACSQRP--LRV-SH -R SENCV A-SNKPQLVKLMTH VK **D**FCV

Optimal alignment between sequences

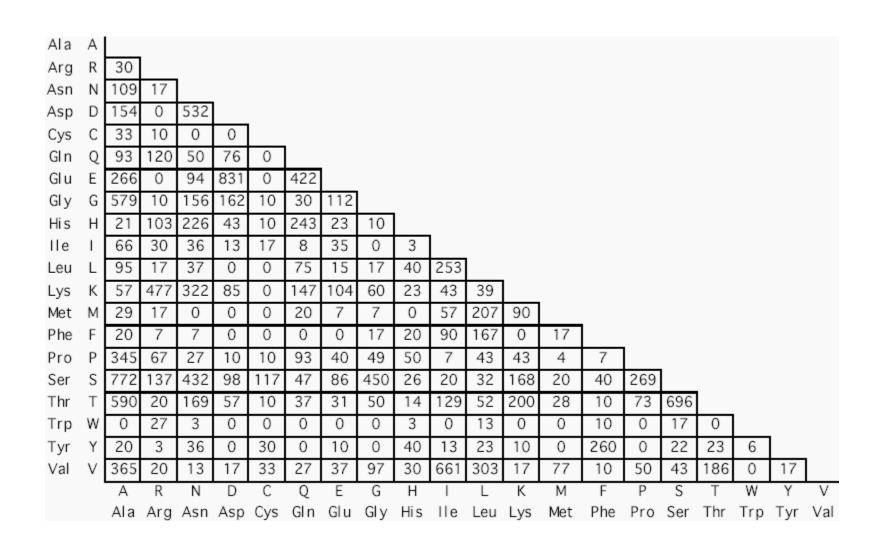
Problem:



similarity score contains:

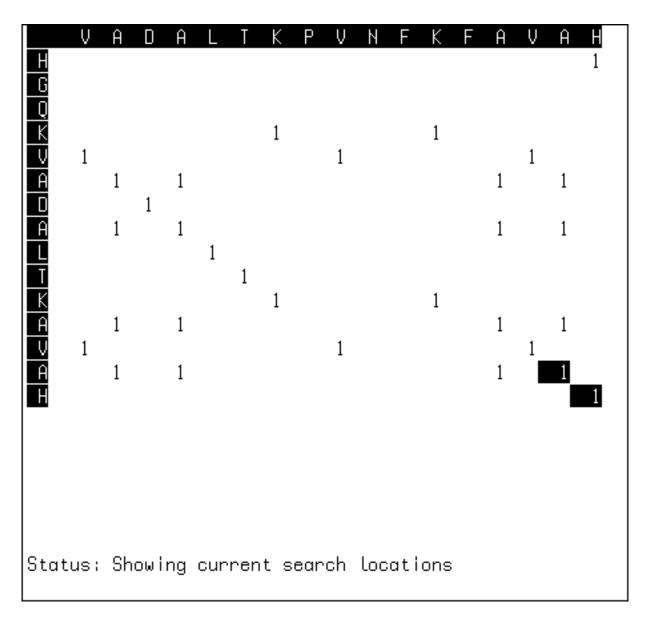
- -variable score for match
- variable cost for gaps
- variable cost for mismatches

Protein amino acid similarity score: Dayhoff's Acceptable Point Mutations (PAMs)

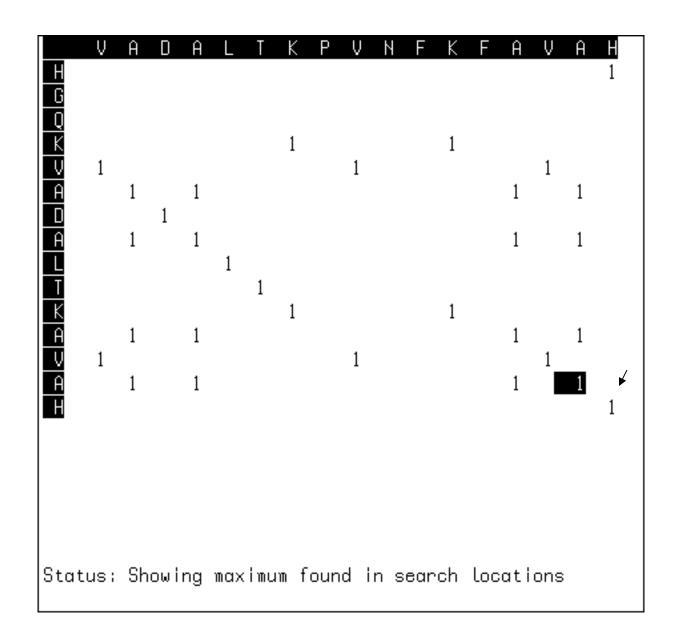


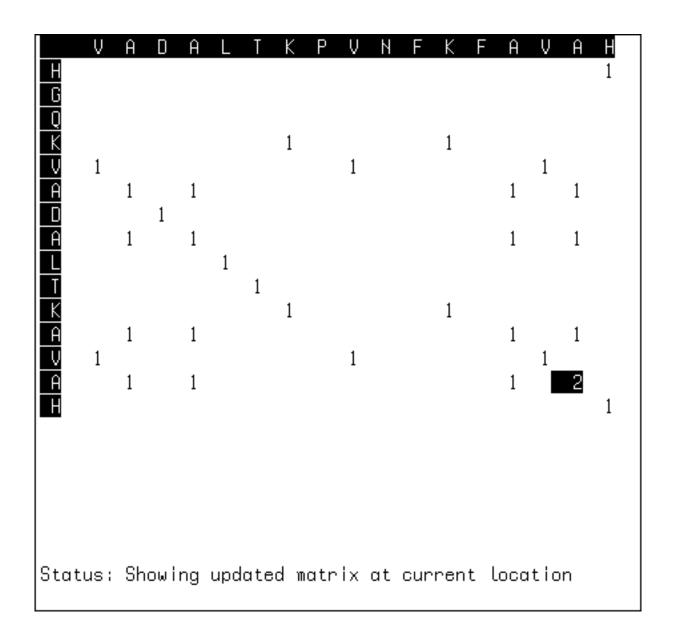
Steps of basic dynamic programming method

- 1. Initialize matrix to match scores (for simplicity: 0 or 1)
- 2. Do summation operation
 - Finds the maximum number of matches that can be obtained starting at any position and proceeding "forward"
- 3. Traceback to find maximum match alignment

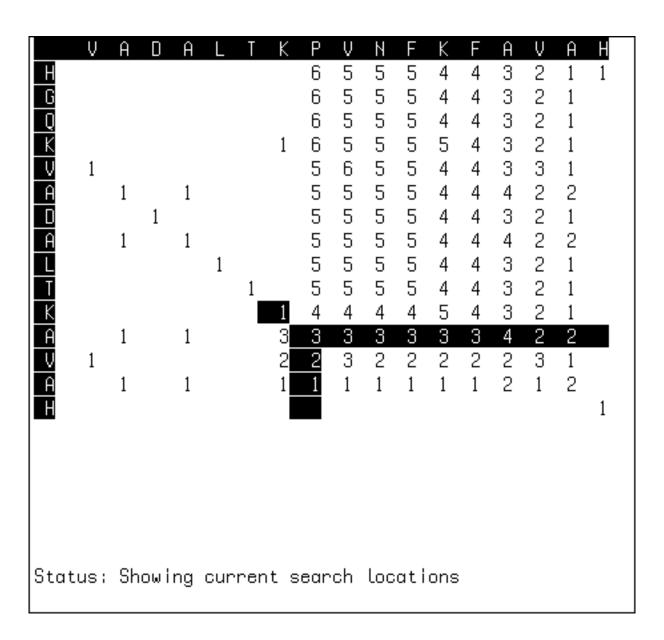


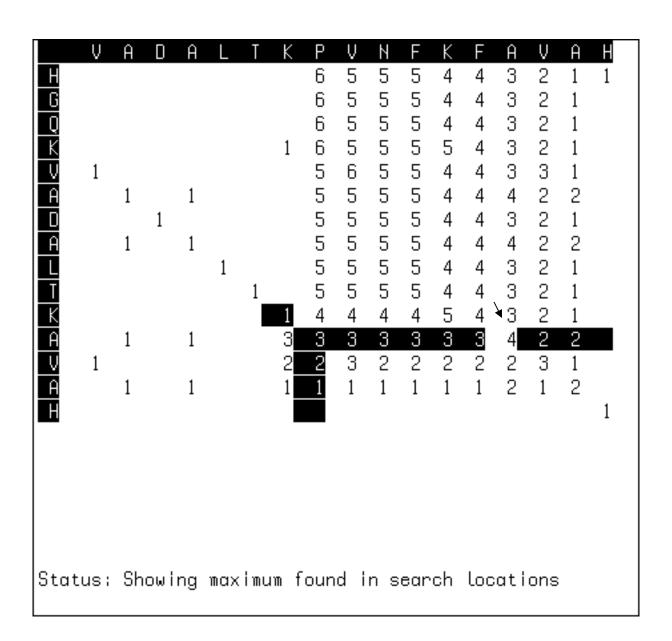
Robert F. Murphy: http://www.cmu.edu/bio/education/courses/03310/LectureNotes/ LecturesPart07.pptf

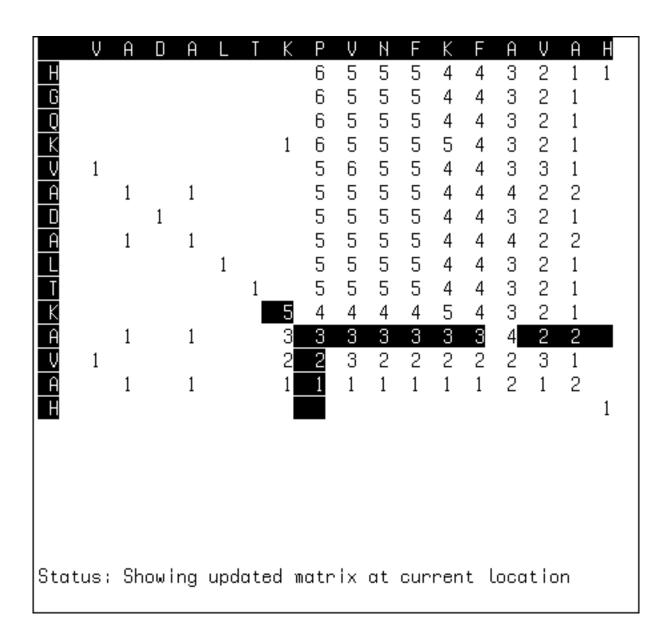


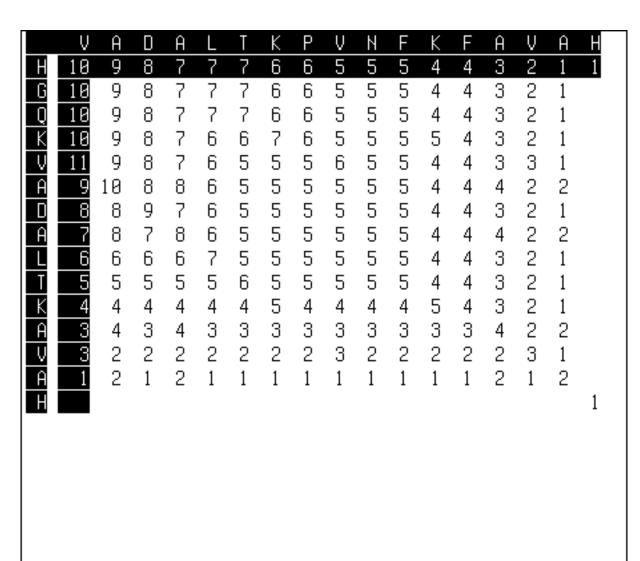


Robert F. Murphy:

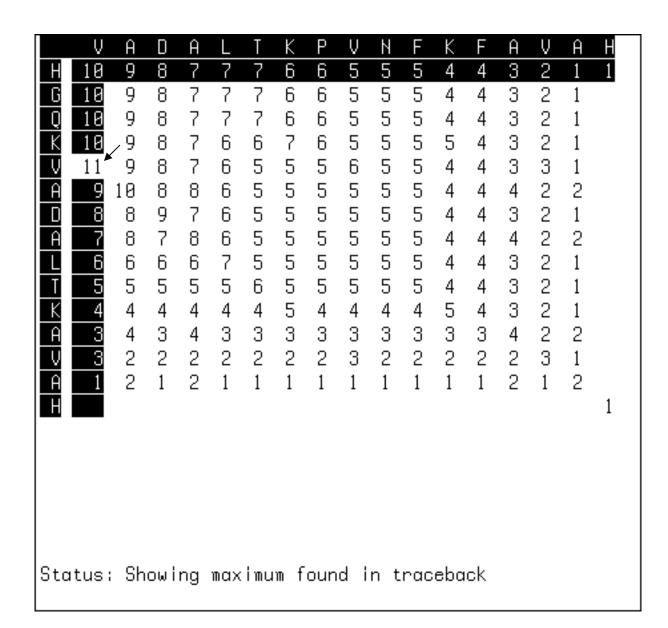


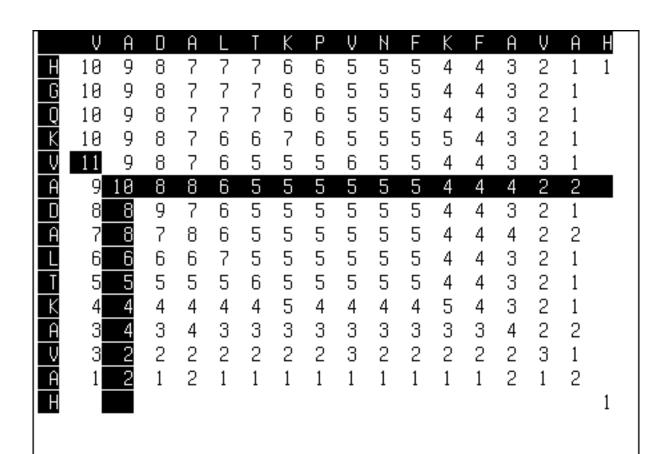






Status: Showing current traceback search locations

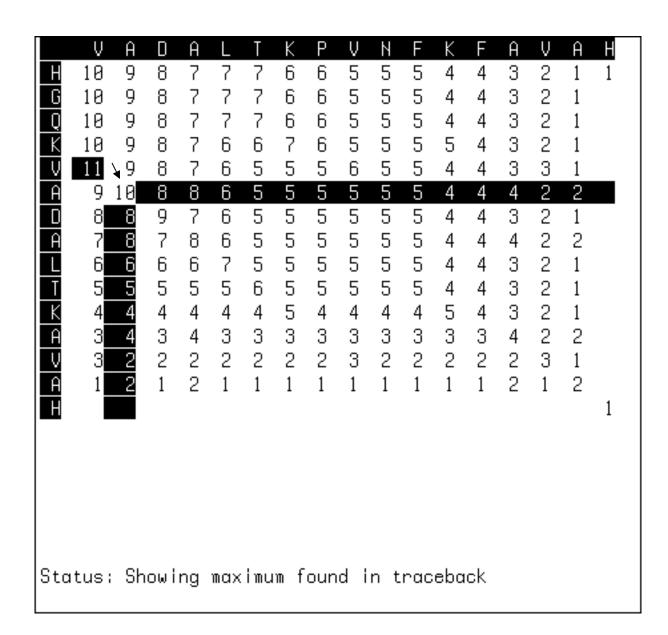


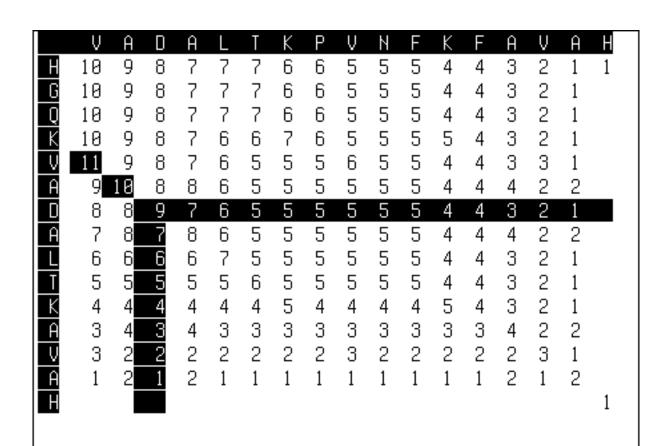


Status: Showing current traceback search locations

- - - - V

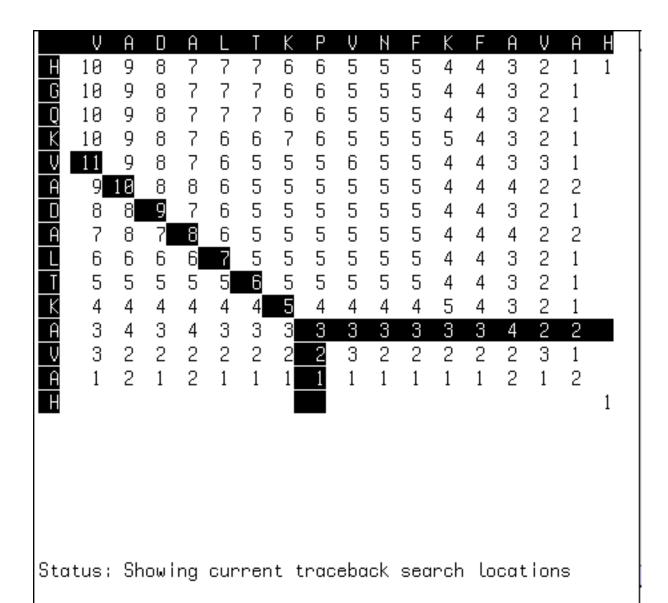
HGQKV





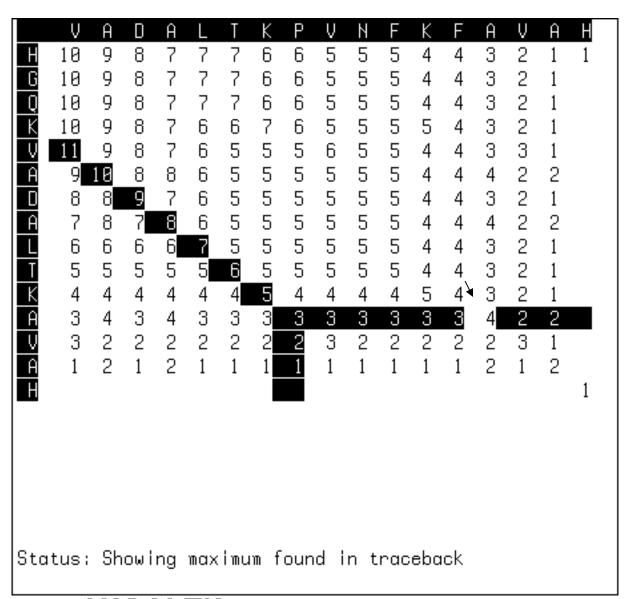
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---VA



----VADALTK

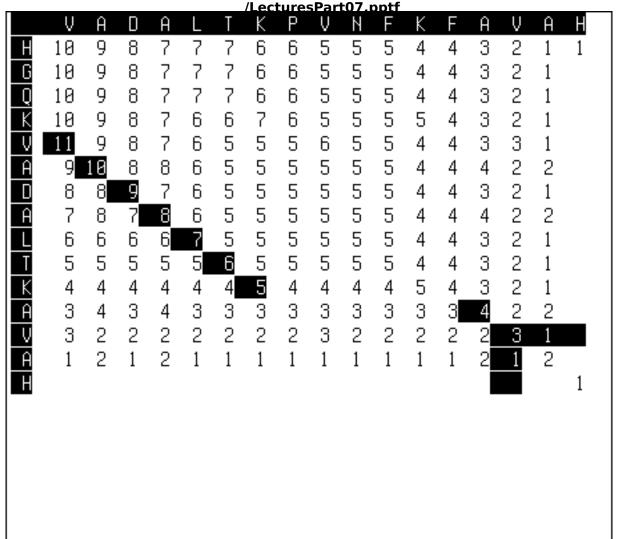
HGQKVADALTK



---VADALTK

HGQKVADALTK

Robert F. Murphy: http://www.cmu.edu/bio/education/courses/03310/LectureNotes



Status: Showing current traceback search locations

---VADALTKPVNFKFA

HGQKVADALTK----A

Robert F. Murphy: http://www.cmu.edu/bio/education/courses/03310/LectureNotes

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H	10	9	8	7	7	7	6	6	5	5	5	4	4	3	2	1	1
G	10	9	8	7	7	7	6	6	5	5	5	4	4	3	2	1	
Q	10	9	8	7	7	7	6	6	5	5	5	4	4	3	2	1	
K	10	9	8	7	6	6	7	6	5	5	5	5	4	3	2	1	
V	11	9	8	7	6	5	5	5	6	5	5	4	4	3	3	1	
А	9	10	8	8	6	5	5	5	5	5	5	4	4	4	2	2	
D	8	8	9	7	6	5	5	5	5	5	5	4	4	3	2	1	
А	7	8	7	8	6	5	5	5	5	5	5	4	4	4	2	2	
	6	6	6	6	7	5	5	5	5	5	5	4	4	3	2	1	
Ī	5	5	5	5	5	6	5	5	5	5	5	4	4	3	2	1	
K	4	4	4	4	4	4	5	4	4	4	4	5	4	3	2	1	
А	3	4	3	4	3	3	3	3	3	3	3	3	3	4	2	2	
V	3	2	2	2	2	2	2	2	3	2	2	2	2	2	3	1	
A	1	2	1	2	1	1	1	1	1	1	1	1	1	2	1	2	
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----VADALTKPVNFKFAVAH

HGQKVADALTK----AVAH

Summation operation

- 1. Start in lower right corner
- 2. Move up one position and left one position
- 3. Find largest value in either (a) row segment starting one below current position and extending to the right or (b) column segment starting one to the right of current position and extending down

Summation operation (cont.)

- 4. Add this value to the value in the current cell
- 5. Repeat steps 3 and 4 for all cells to the left in current row and all cells above in current column
- 6. If we are not in the top left corner, go to step 2

Multiple sequence alignment

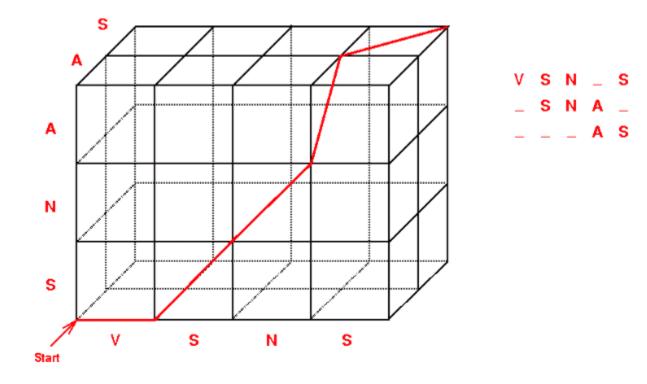
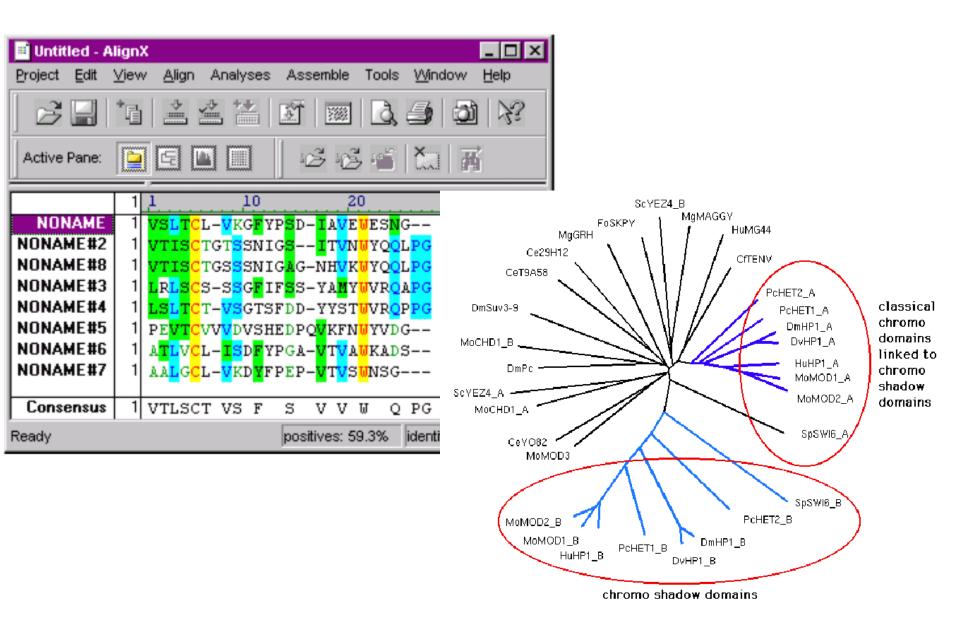


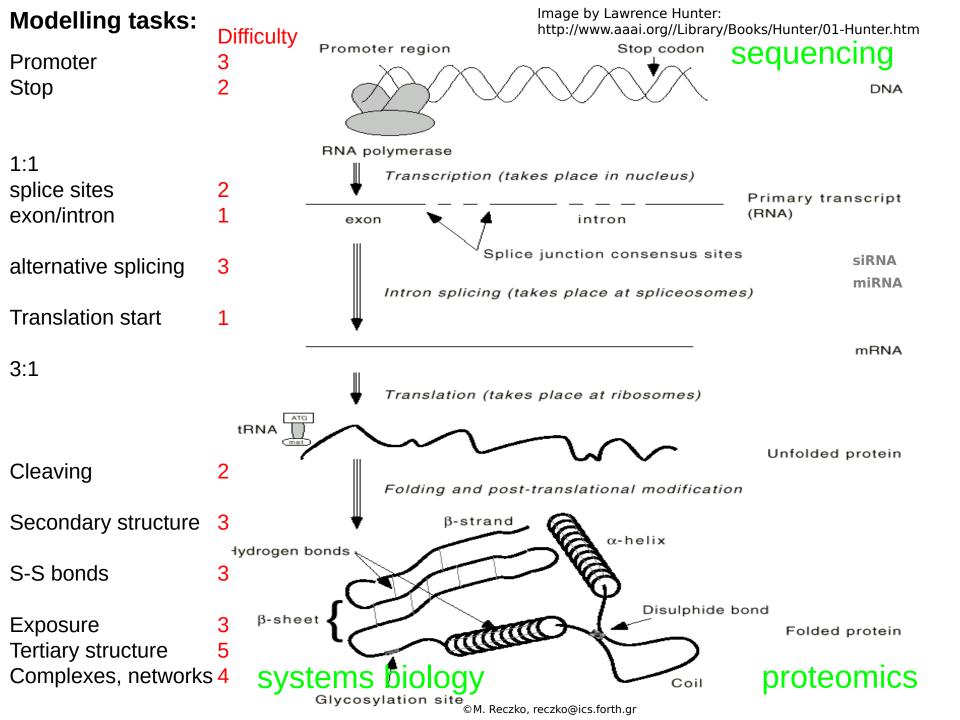
Figure source: http://www.techfak.unibielefeld.de/bcd/Curric/MulAli/node2.html#SECTION0002000000000000000

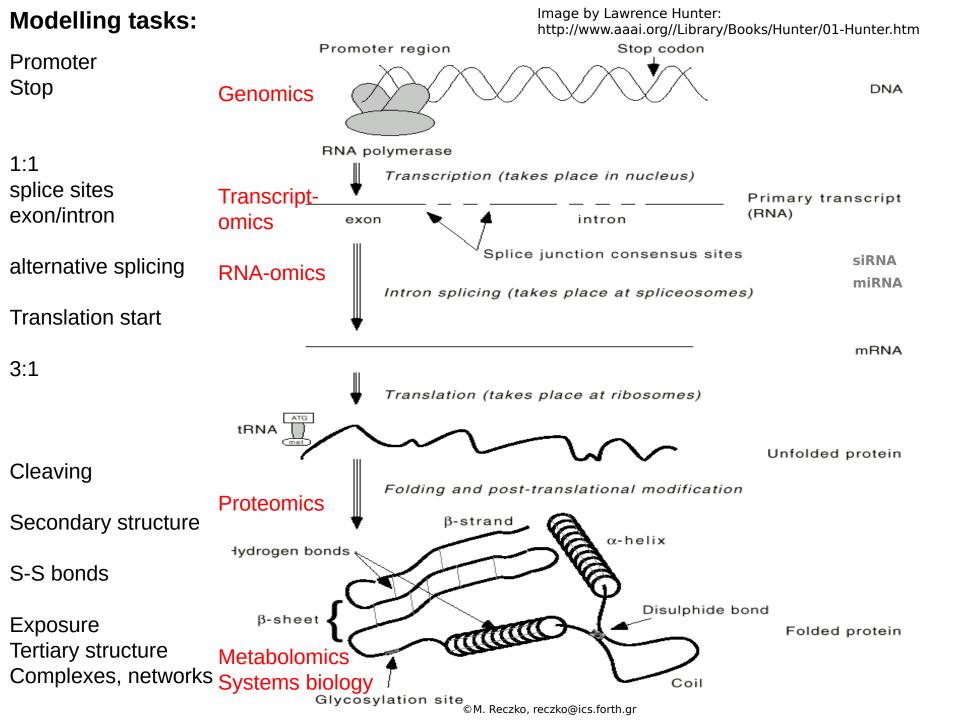
Calc. of optimal solution infeasible for >5 sequences

- **⇒** Heuristic solutions
- \Rightarrow e.g. progressive alignment (CLUSTALW)

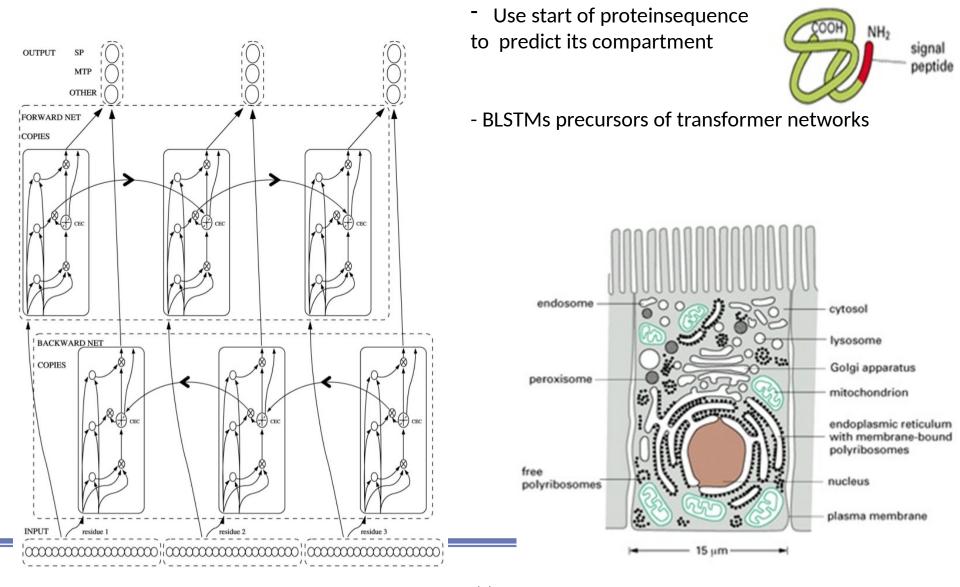
Multiple sequence alignment for phylogenetic trees



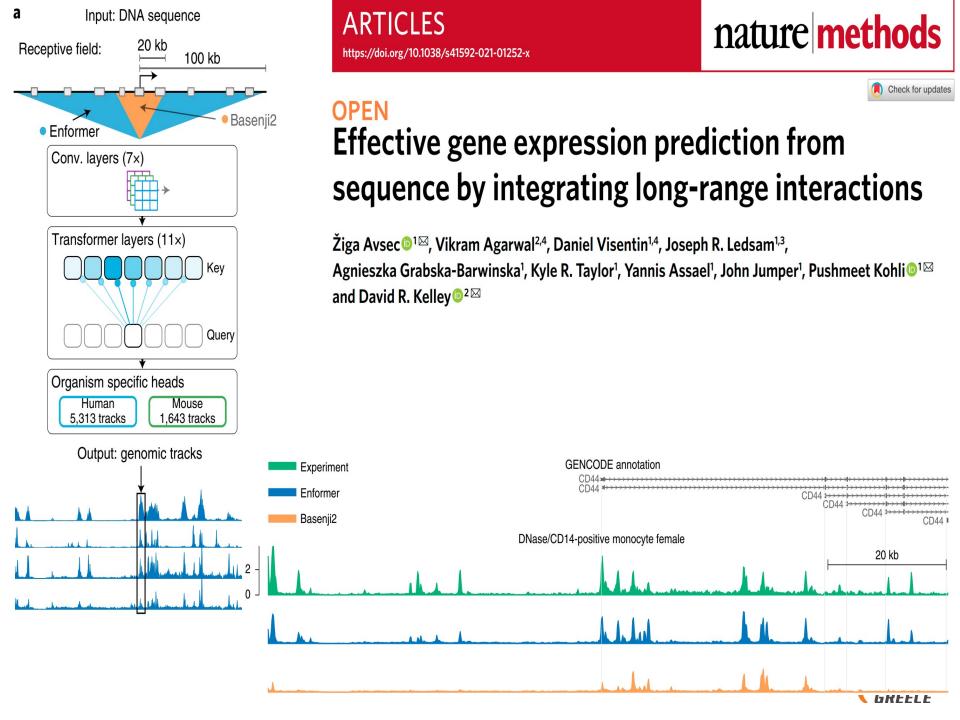


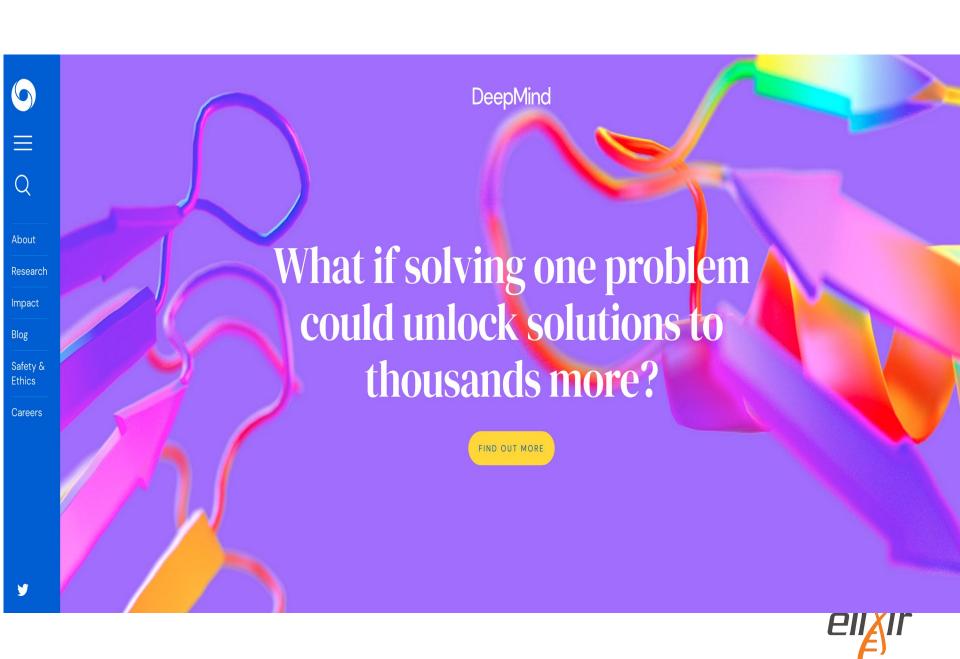


Introduction novel sequence learning algorithm (BLSTM)

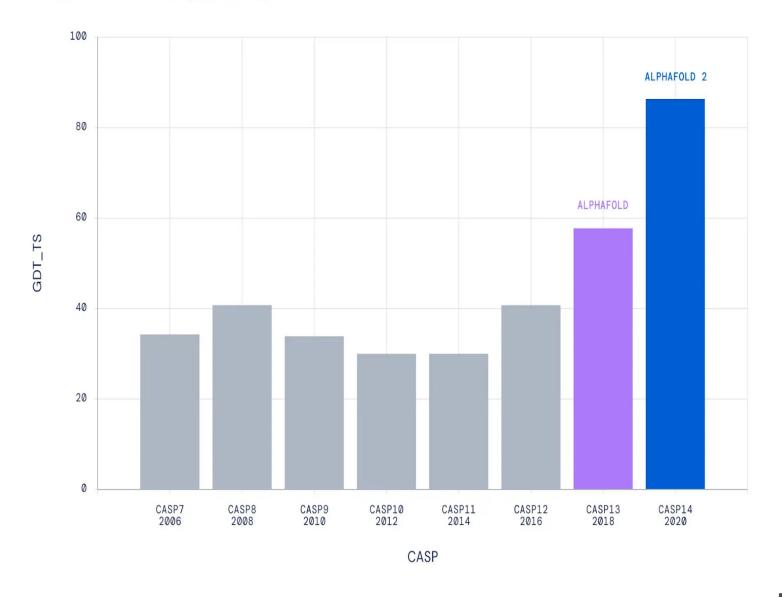


IEEE Transactions on Computational Biology and Bioinformatics 2006, 4(3), pp.441-6.



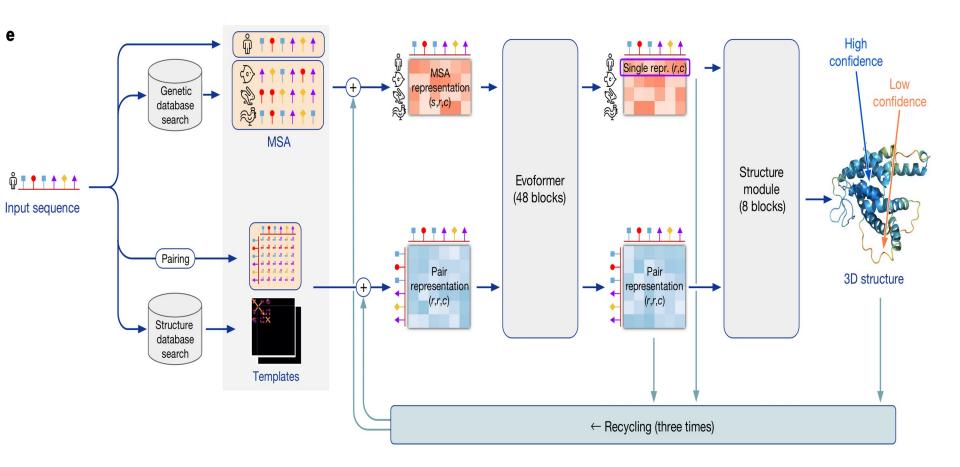


Median Free-Modelling Accuracy



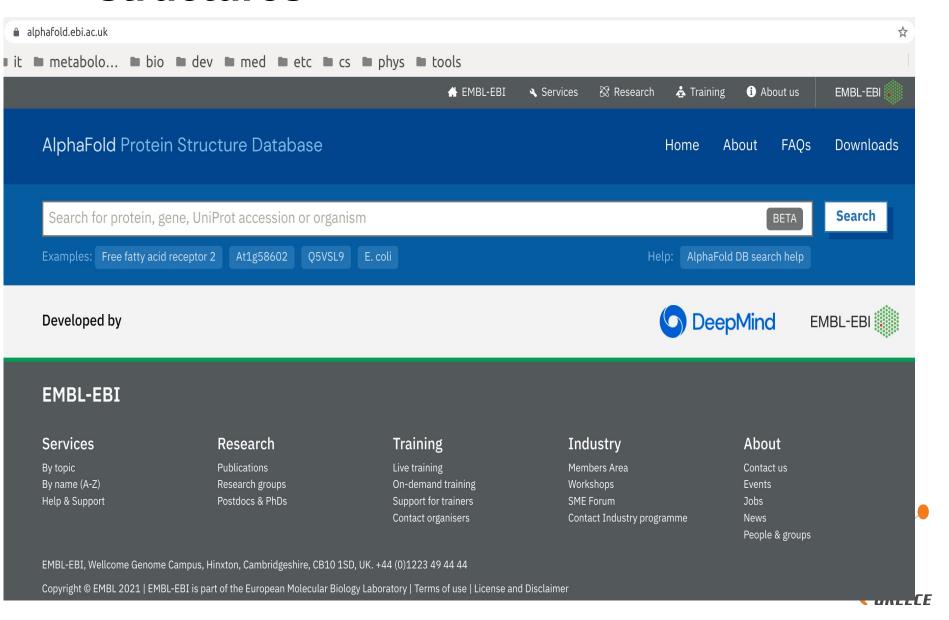
Improvements in the median accuracy of predictions in the free modelling category for the best team in each CASP, measured as best-of-5 GDT.

AlphaFold2 architecture





AlphaFold2 database of predicted structures



ELIXIR

ELIXIR is an intergovernmental organisation that brings together life science resources such as databases, software tools, training materials, standards and compute resources, from across Europe.

The goal of ELIXIR is to coordinate life science resources from across Europe so they form a single infrastructure. This makes it easier for scientists to:

Find and share data
Exchange expertise
Agree on best practices in scientific research

Check: https://elixir-europe.org

https://elixir-greece.org



https://www.covid19dataportal.org & https://covid19dataportal.gr



About



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Expression

Proteins

Biochemistry

Literature

Accelerating research through data sharing



Viral sequences



Raw and assembled sequence and analysis of SARS-CoV-2 and other coronaviruses.

111,900 records >

Host sequences 🕣



Raw and assembled sequence and analysis of human and other hosts.

973 records >

About this portal

The COVID-19 Data Portal was launched in April 2020 to bring together relevant datasets for sharing and analysis in an effort to accelerate coronavirus research. It enables researchers to upload, access and analyse COVID-19 related reference data and specialist datasets as part of the wider European COVID-19 Data Platform.

To enquire on how to collaborate on the European COVID-19 platform: ecovid19@ebi.ac.uk.

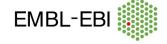
To share your data on COVID-19 Data Portal: virus-dataflow@ebi.ac.uk.

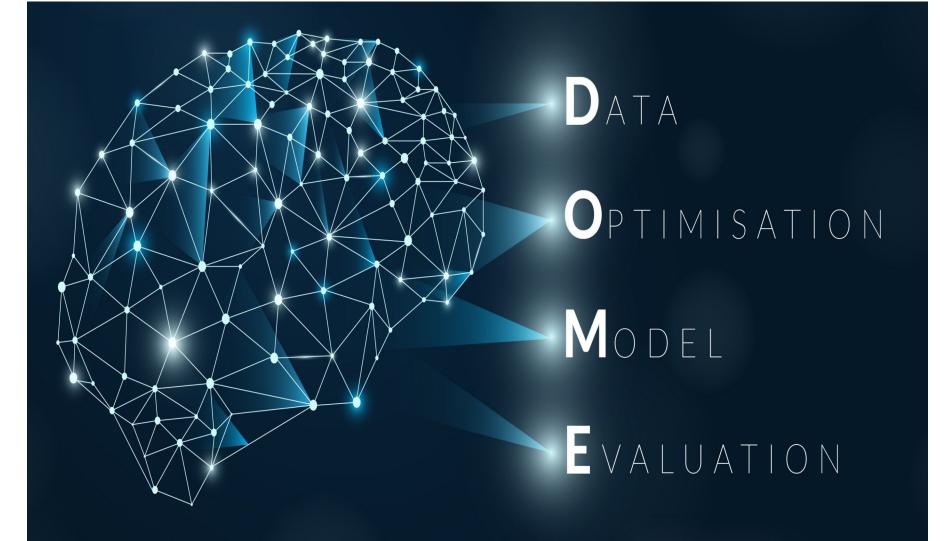
Viral sequences Host sequences Expression **Proteins**

Biochemistry Literature Related Resources

About the Portal SARS-CoV-2 Data Hubs **Our Partners** Submit Data







MACHINE LEARNING FOCUS GROUP



Website: https://dome-ml.org/

Data Optimisation Model Evaluation

Provenance Data splits Redundancy Availability

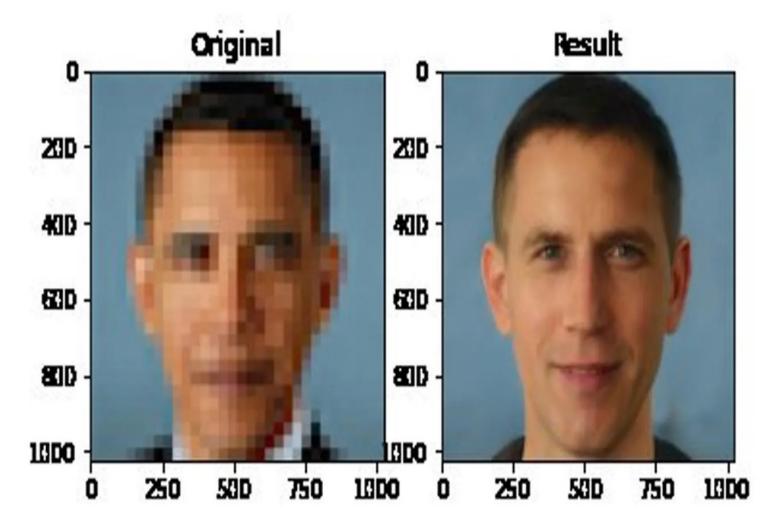
Algorithm Data encoding Parameters Features Fitting Availability

Interpretability Meta-predictions Execution time Availability of software

Evaluation Performance Comparison Confidence Availability

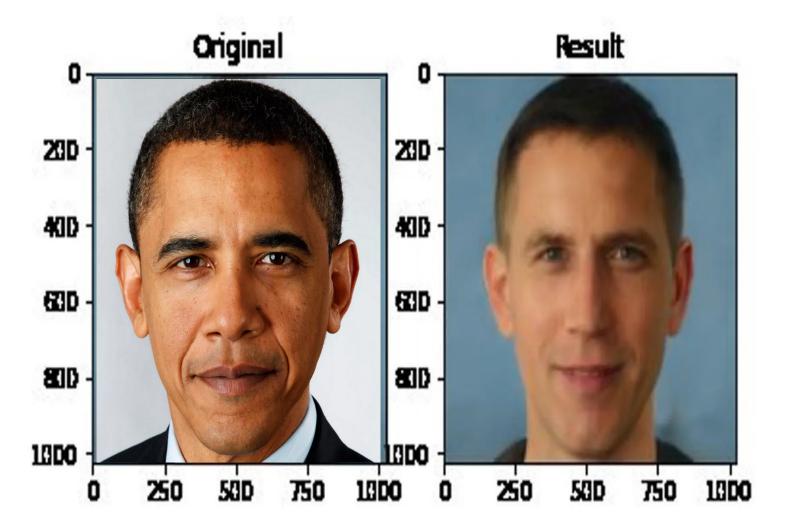


Dangers of deep/machine learning



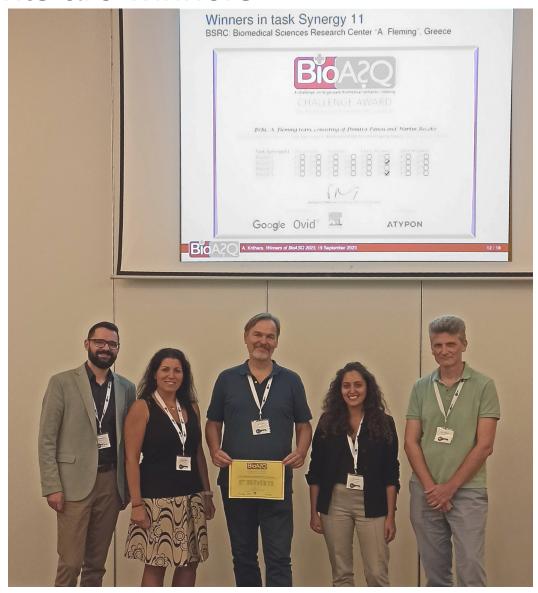
https://www.theverge.com/21298762/face-depixelizer-ai-machine-learning-tool-pulse-stylegan-obama-bias

Dangers of deep/machine learning: Bias

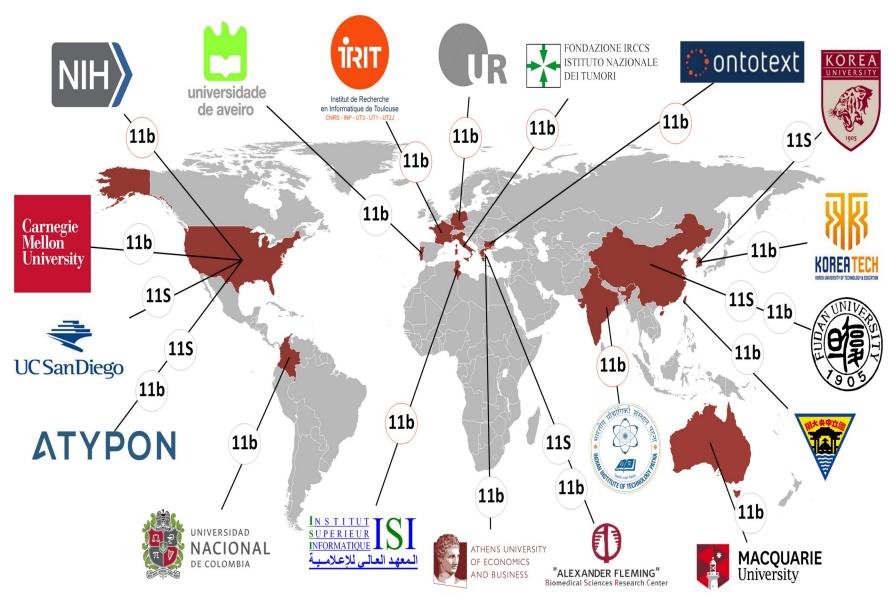


https://www.theverge.com/21298762/face-depixelizer-ai-machine-learning-tool-pulse-stylegan-obama-bias

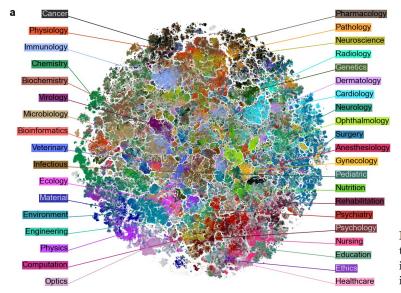
ITBI students are winners



BioASQ: Int. competition for biomedical QA



Transformers help clustering all scientifc papers



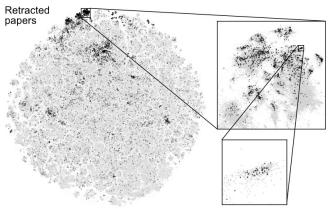


Figure 6: Retracted papers group together. All retracted papers with intact abstracts (11,756) are highlighted in black, plotted on top of the non-retracted papers. First inset corresponds to one of the regions with higher density of retracted papers (3.8%), covering research on cancer-related drugs, marker genes, and microRNA. Second inset corresponds to a subregion with a particularly high fraction of retracted papers (10.8%), the one we used for manual inspection.

https://www.biorxiv.org/content/10.1101/2023.04.10.536208v2



HEALTH & WELLNESS

A boy saw 17 doctors over 3 years for chronic pain. ChatGPT found the diagnosis

Alex experienced pain that stopped him from playing with other children but doctors had no answers to why. His frustrated mom asked ChatGPT for help.







Sept. 11, 2023, 5:42 PM EEST / Updated Sept. 12, 2023, 5:31 PM EEST / Source: TODAY

By Meghan Holohan

During the COVID-19 lockdown, Courtney bought a bounce house for her two young children. Soon after, her son, Alex, then 4, began experiencing pain.

"(Our nanny) started telling me, 'I have to give him Motrin every day, or he has these gigantic meltdowns," Courtney, who asked not to use her last name to protect her family's privacy, tells TODAY.com. "If he had Motrin, he was totally fine."

Then Alex began chewing things, so Courtney took him to the dentist. What followed was a three-year search for the cause of Alex's increasing pain and eventually other symptoms.

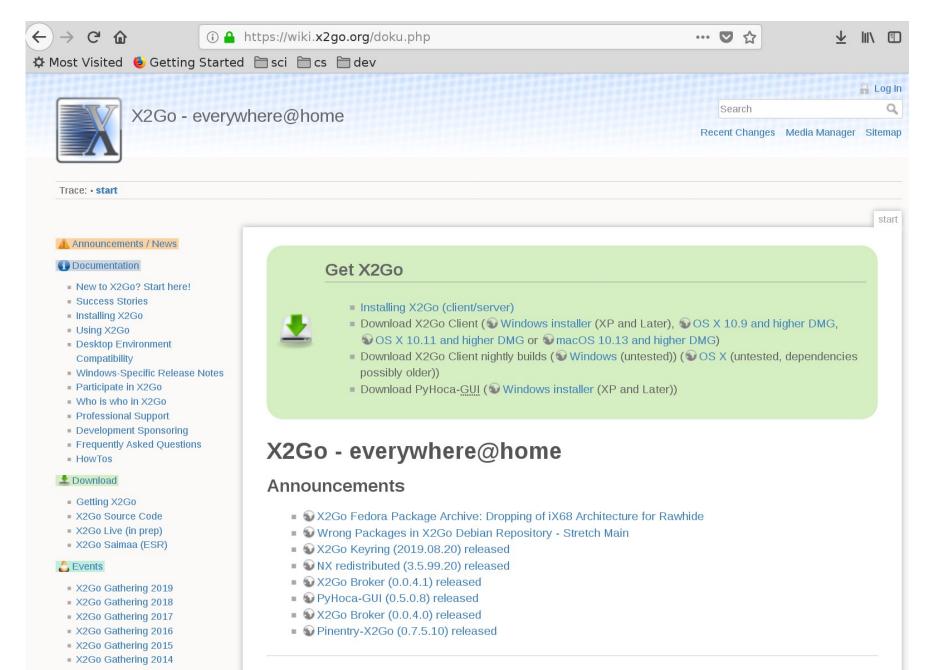


Alex saw 17 doctors over three years for his chronic pain, but none were able to find a diagnosis that explained all of his symptoms, his mom says. Courtesy Courtney

Get your account on the Virtual Machine for the exercises in hands-on during the lectures and at home

- use 28 CPUs, 248GB RAM for all
- 20GB disk-space for each + 400GB shared

Install x2go to access graphical user interface



Access to virtual machine

Install x2go from: https://wiki.x2go.org/doku.php/download:start

	<< change icon
	Path: /
X2go:	Server
	Host: snf-;
	Login: ubuntu
	SSH port: 22
	Use RSA/DSA key for ssh connection:
	Try auto login (via SSH Agent or default SSH key)
	Kerberos 5 (GSSAPI) authentication
	Delegation of GSSAPI credentials to the server
	Use Proxy server for SSH connection
	Session type
	Custom desktop Command: /usr/bin/lxsession -:

- (Virtualbox: http://genomics-lab.fleming.gr/fleming/uoa/vm/TrinityVM U16.ova

Introduction to Bioinformatics 2022-2023

Exercise 1 (M. Reczko):

(Adapted from:

https://web.archive.org/web/20150425010121/http://www.ableweb.org/volumes/vol-28/v28repri nt.php?ch=8

In a hypothetical scenario many people in a city suddenly come down with a serious illness. All the victims have in common is that they were all in a downtown pedestrian mall at a certain time five days before. Could terrorists have released a cloud of viruses or bacteria from a vehicle downwind of the mall? You work for the Centers for Disease Control and Prevention, and you have to find out.

A sample of non-human DNA (bacterial or viral) has been isolated from the victims. Identify the DNA sample as well as you can. Some of the DNA molecules are very short, and have been partially degraded. You will notice that the sequence is sprinkled with Ns, "N" stands for "nucleotide" and means that the nucleotide at that position could not be determined. Some judgment is called for as you interpret your results. First, everyone has bacteria and viruses in his or her body, and sometimes they can cause disease. However, we are looking for exotic pathogens with bioterrorism potential (e.g., anthrax or smallpox rather than the common cold). Even AIDS, although it is deadly, would not work as a bioterror weapon because the disease develops too slowly and the virus is too hard to disseminate. For the purposes of this exercise, we will not consider a pathogen a bioterror agent unless it is listed as a potential agent on the Centers for Disease Control and

Prevention Web site at https://emergency.cdc.gov/agent/agentlist.asp..

Second, organisms that are evolutionarily related have similar DNA, which might lead you to sound a false alarm. For example, say you find the following when you do a BLAST search on a certain DNA sample:

Sequences producing significant alignments:	Score (Bits)	E Value
gi 40012 emb X02369.1 BSORIC Bacillus subtilis oriC region gi 32468687 emb Z99104.2 BSUB0001 Bacillus subtilis complete gi 467326 dbj D26185.1 BAC180K B. subtilis DNA, 180 kilobase reg gi 39877 emb X12778.1 BSDNAA Bacillus subtilis dnaA gene 5'-regi gi 56160984 gb CP000002.2 Bacillus licheniformis ATCC 14580, co gi 52346357 gb AE017333.1 Bacillus licheniformis DSM 13, comple gi 39878 emb X12779.1 BSDNAAN Bacillus subtilis genes for dnaA (5967 5967 5967 846 690 690 587	0.0 0.0 0.0 0.0 0.0 0.0 8e-164
gi 39893 emb X17013.1 BSDPD Bacillus subtilis lys gene for di gi 51973633 gb CP000001.1 Bacillus cereus E33L, complete genome gi 49328240 gb AE017355.1 Bacillus thuringiensis serovar kon gi 50082967 gb AE017334.2 Bacillus anthracis str. 'Ames Ancesto gi 49176966 gb AE017225.1 Bacillus anthracis str. Sterne, compl	525 337 329 329 329	2e-145 1e-88 3e-86 3e-86 3e-86

Bacillus subtilis is a harmless and very common soil bacterium. It is closely related to Bacillus anthracis. Bacillus anthracis causes anthrax, and is a dangerous bioterror weapon. Note from the similarity score (second column from the right) that Bacillus subtilis DNA is far more similar to the sample than Bacillus anthracis DNA is. Unless one of your samples gives a stronger indication of Bacillus anthracis than this, the mention of B. anthracis in the output is probably just due to genetic similarities between it and B. subtilis.

1. Analyze the samples

>outbreak14

GCCGAGTTAGTCTTGTGCTNACGGAACTTATTGTATGAGTANTGATTTGAAAGAGCTANANT TAAAAAATCACTAATNAATNTAAGAGCGGACTTAACNAGCGTAAAACTGTCTTACTAATTAAT TGTCAGTTAGCTCGTTCAGGTAATGGTTCCTANCGGNCAATGCAGGAAGAGTTCTACCTGG AACTGANAGACCGCTGGCGGTGACAACACACACTACGTCAAAATAAGA

>outbreak15

TAGTCTTGTGCTNACGGAACTTATTTATGAGGTACCCACCGANTCTGAAAAACCGCTAATANA GCACTTTAAAAAATAAGAGCAGAATGGGATTTAAGGATAG

separately using both megablast and blastn at

https://blast.ncbi.nlm.nih.gov/Blast.cgi?PROGRAM=blastn&PAGE_TYPE=BlastSearch&BLAST_SPEC=&LINK_LOC=blasttab

- and to determine if there is any evidence of bioterror agents. Use the general nucleotide collection (nr/nt). Report any differences between the 2 algorithms.
- 2. Check the CDC Web site at https://emergency.cdc.gov/agent/agentlist.asp. to see if the CDC considers any found organism to be a potential weapon. If you've found a bioterror agent, research it on the CDC site so you can describe its effects on humans.
- 3. The health effects of many pathogenic bacteria are briefly described on the NCBI Genomes Web site at < http://www.ncbi.nlm.nih.gov/genomes/lproks.cgi. Click on a species name to see its information. It also might be helpful to do a general web search.

SEND SOLUTIONS (for M.Reczko exercise) ONLY TO: mareczko@di.uoa.gr