

Lecture 1: Introduction to bioinformatics

Torgeir R. Hvidsten

Professor Norwegian University of Life Sciences

Guest lecturer Umeå Plant Science Centre Computational Life Science Centre (CLiC)

Course information (I)

- 10 Lectures
- 7 computer labs
- Book: Introduction to Bioinformatics. Arthur M. Lesk, Oxford University Press.
- Credit points: 4 ECTS
- To pass:
 - attend lectures and labs
 - send lab to: david.sundell@plantphys.umu.se





Course information (II)

- Course webpage:
 - http://www.trhvidsten.com/complife/2012
- Here you can find the
 - course program
 - online resources
- and download
 - lecture slides
 - labs and suggestions for solutions to labs
 - additional material/examples research articles

🕕 Computational Life Science - Opera									
<u>File Edit View Bookmarks Feeds</u> <u>T</u> ools <u>H</u> elp									
😫 Downloads	🛛 🗙 🔂 Google Re	ader (75) × 🕨 🛍 Computational Life S	Sci × 🕂			•			
← → ⊃ •- (🚯 Web www.trh	vidsten.com/complife/2012/index.html		🚖 🛃 👻 Search with Google					
() Aftenposten db Dag	gbladet YRYr W W	/ikipedia 🔣 Birro 🌒 Football-italia 👫 Mil	ano Siamo Noi 🛯 🕕 TRH 🛛 🗷 CiteULike 🔀 IntraUMB 🔼 SpruceWiki	UPSC NetAlign					
20 Th	D12.09.20 hursday	Lecture 1: 9-11 KBF30	Introduction to bioinformatics, computational biology and systems biology	Slides 1	WEEK 38	*			
		Self-Study	Introduction, Scientific publication and archives, Archives and information retrieval	Chapter 1, 3 and 4					
20 Fr	2012.09.21 Lecture 2: 9-11 Friday KB4C10		Programming in Perl: Introduction 1	Slides 2					
		Self-study	Introduction, Scientific publication and archives, Archives and information retrieval	Chapter 1, 3 and 4					
20 Ma	012.09.24 onday	Computer LAB 1: 9-11 MA316	Basic expressions, scalars, arrays, loops, conditions, file handling	Lab 1 (solutions) numbers.txt Perl tutorials	WEEK 39				
		Lecture 3: 13-15 KBF30	Programming in Perl: Introduction 2	Slides 3					
20 Tu	012.09.25 uesday	Computer LAB 2: 9–11 MA316	Hashes, data structures, references, subroutines, modules	Lab 2 (solutions) microarray.txt network.txt					
		Self-study	LAB 2 cont.						
We	ednesday	KBF30	Programming in Peri: Introduction 3	511065 4					
		Computer LAB 3: 13–15 MA 316	Bioperl	Lab 3 (solutions) hvidsten.com.html genpept.txt					
20 Th	012.09.27 hursday	Computer LAB 3: 9-11 MA316	LAB 3 cont.						
Lecture 5: 13-15 KBF30		Lecture 5: 13-15 KBF30	Algorithm design and time/space complexity analysis	Slides 5					
20 Fr	012.09.28 iday	Computer LAB 4: 13-15 MA316	Pseudo-code, algorithm design and time/space complexity analysis	Lab 4 (solutions) promoters.txt MEME results (pdf) MEME results shuffled (pdf)					
	Self-study		LAB 3/Lab 4						
20 M	2012.10.01 Lecture 6: 9-11 Monday KBF30		Sequence alignment	Slides 6 Chapter 5	WEEK 40				
Computer LAB 5: 13–15 MA316		Computer LAB 5: 13-15 MA 316	Sequence alignment	Lab 5 (solutions)					
20 Tu	2012.10.02 Tuesday	Lecture 7: 9-11 KBF30	Evolutionary analysis, phylogenetic analysis	Slides 7 Chapter 2 and 5					
		Computer LAB 6: 13-15 MA316	Phylogenetic analysis	Lab 6 get_sequences.pl change_names.pl hiv_ids.txt					
20 We	012.10.03 ednesday	Lecture 8: 9-11 KBF30	Protein structure analysis	Slides 8 Chapter 6					
		Self-study	Read the book	Chapter 2, 5 and 6					
20 Th	012.10.04 hursday	Lecture 9: 9-11 KBF30	Machine learning	Slides 9					
		Computer LAB 7: 13–15 MA316	Machine learning	Lab 7 adenoca_markers.xIs Dennis-ClinCancerRes2005.pdf					
20 Fr	012.10.05 riday	Lecture 10: 9-11 KB4C10	Systems biology	Slides 10 Course evaluation: eval If not present at the last lecture, mail the		-			
🗈 📥 🕜 http://v	www.trhvidsten.com/	/complife/2012/index.html		andoration to one Theodori	0	_			

Course goals

At the end of this course you will :

- have basic knowledge of online bioinformatics resources (databases, ontologies, etc)
- know how to write and debug basic Perl programs and use online Perl libraries
- recognize the different algorithm design techniques and be able to do basic time/space complexity analysis
- know how to apply, and interpret results from, classical bioinformatics approaches such as pairwise and multiple alignment, phylogenetic analysis, and machine learning
- have a basic understanding of approaches applied in structural bioinformatics and systems biology

This lecture

- Introduction to bioinformatics
- Introduction to the topics of this course
- Introduction to programming:
 - programming languages
 - pseudo-code



DNA sequence comparison: First success story of bioinformatics

- In 1984 Russell Doolittle and colleagues found similarities between a cancer-causing gene and a normal growth factor (PDGF) gene using a database search
- Finding sequence similarities with genes of known function is a common approach to predict the function of a newly sequenced gene
- "Bioinformatics will disappear and become an integrated part of biology" (Doolittle, 2002)

Bioinformatics



Definitions

- *Bioinformatics*: Research, development, or application of computational tools and approaches for expanding the use of biological data, including those to acquire, store, organize, archive, analyze, or visualize such data.
- *Computational biology*: The development and application of dataanalytical and theoretical methods, mathematical modeling and computational simulation techniques to the study of biological systems.
- *System biology*: the systematic study of complex interactions in biological systems (integration/holism instead of reduction)

The book

- Introduction to Bioinformatics. Arthur M. Lesk, Oxford University Press (2nd edition is online)
- Covers mostly basic bioinformatics (databases, online resources)
- Some of the lectures will be more advanced and towards computational biology/systems biology

Programming

- Lectures 2-4, Labs 1-3
- Do bioinformaticians need to know how to program?
- We will use the programming language Perl



Algorithm designs

- Lecture 5, Lab 4
- Types of search algorithms
- Time/space complexity



Sequence alignment as a search problem



Deletion Matches Insertion

-TGCAT-A-C AT-C-TGATC

Algorithm design (I)

- Exhaustive algorithms (brute force): examine every possible alterative to find the solution
- Branch-and-bound algorithms: omit searching through a large number of alternatives by branch-and-bound or pruning
- Greedy algorithms: find the solution by always choosing the currently "best" alternative
- Dynamic programming: use the solution of the subproblems of the original problem to construct the solution

Algorithm design (II)

- Divide-and-conquer algorithms: splits the problem into subproblems and solve the problems independently
- Randomized algorithms: finds the solution based on randomized choices
- Machine learning: induce models based on previously labeled observations (examples)

Time complexity

- Genome assembly: pice together a genome from short reads (~200bp)
 - Aspen: 300M reads
 - Spruce: 3000M reads
- Pair-wise all-against-all alignment for Aspen takes 3 weeks on 16 processors
- What about spruce?



Tractable versus intractable problems

- Some problems requires polynomial time
 - e.g. sorting a list of integers
 - called tractable problems
- Some problems require exponential time
 - e.g. listing every subset in a list
 - called intractable problems
- Some problems lie in between
 - e.g. the traveling salesman problem
 - called NP-complete problems
 - <u>nobody have proved whether a polynomial time algorithm</u> <u>exists for these problems</u>

Traveling salesman problem



Sequence alignment

- Lecture 6, Labs 3 and 5
- Pair-wise alignment (BLAST)

• Multiple alignment (PSI-BLAST, HMMs, pFAM, etc.)

VTISCTGSSSNIGAG-NHVKWYQQLPG VTISCTGTSSNIGS--ITVNWYQQLPG LRLSCSSSGFIFSS--YAMYWVRQA--LSLTCTVSG-SFDD--YYSTWVRQP--PEVTCVVVD-SHEDPQVKFNWYVDG--ATLVCLISDFYPGA--VTVAWKADS--AALGCLVKD-FPEP--VTVSWNSG---VSLTCLVKGFYPSD--IAVEWESNG--



Evolutionary analysis, phylogenetic analysis

• Lecture 7, Lab 6



Phylogeny-aware gap placement

Conclusion:

"The resulting alignments may be fragmented by many gaps and may not be as visually beautiful as the traditional alignments, but if they represent correct homology, we have to get used to them."

A. Löytynoja and N. Goldman. Phylogeny-Aware Gap Placement Prevents Errors in Sequence Alignment and Evolutionary Analysis. Science 320: 1632-35, 2008.



Phylogeny-aware gap placement



Protein structure analysis

- Lecture 8
- The sequence, structure, function relationship
- The protein folding problem



Hemoglobin

Hammer

The protein folding problem

Anfinsen's thermodynamic hypothesis (1973): Protein folding is a strictly physical process, which solely depends on the protein sequence



The folding problem:

discover nature's algorithm for specifying 3D structure of proteins from their amino acid sequences



Computational folding methods

- No effective folding machine exists that is based on physical principles and energy minimization alone
- Current computational methods rely on known protein structures – machine learning approach:
 - Template-based modeling
 - Template-free modeling



Ab initio prediction



Score and select model

Machine learning

- Lecture 9, Lab 7
- Induction of general models from data
 - Bayes decision rule
 - Decision trees
 - Nearest neigbour approach
 - Artificial neural networks (linear versus non-linear methods)
 - Genetic algorithms and genetic programming
 - Model evaluation

What is AI?

"...making a machine behave in ways that would be called intelligent if a human were so behaving"

- John McCarthy, August 31, 1955

"The subfield of computer science concerned with the concepts and methods of symbolic inference by computer and symbolic knowledge representation for use in making inferences."

- The Free On-line Dictionary of Computing (September 27, 2003)

Acting humanly: Turing test

- Turing proposed that a computer program show intelligent behavior if is able to fool a human interrogator:
- The Turing test: the computer is interrogated by a human via a teletype, and passes the test if the interrogator cannot tell if there is a computer or a human at the other end
 - natural language processing
 - knowledge representation
 - automated reasoning
 - machine learning



1912-1954

Definitions of Al

Four categories of definitions:

Empirical science

Engineering

Human-centered

Rationality-centered

Reasoning	Systems that think like humans	Systems that think rationally				
Behavior	Systems that act like humans	Systems that act rationally				

Al techniques

- Logics
- Knowledge representation
- Search
- <u>Machine learning</u>
- Pattern recognition
- Automatic theorem proving
- Planning
- Machine vision
- Natural language processing

Example: Decision tree learning

Country	Communists	Socialists	Greens	Social Democrats	Liberals	A	grarians	Subnational, regional and ethnic parties		hristia mocra	n its	Conservatives	Extreme Right
Norway	0	7	0	38	4		8	0		9		24	6
Sweden	6	0	2	43	10		17	0		2		18	1
Denmark	4	9	0	33	13		14	0		3		15	9
Finland	15	0	2	24	3		25	5		3		21	0
lceland	0	18	3	16	4		22	0		0		36	0
UK	0	0	9	39	15		0	4	0			42	0
Netherlands	2	5	0	30	23		0	0		37		0	0
Belgium	2	0	4	27	19		0	14		31		0	2
Luxembourg	6	1	3	31	21		0	0		34		0	1
Switzerland	2	2	7	22	23		11	0		22		3	5
Austria	1	0	2	48	0		U	0		41		0	8
Germany	1	0	3	40	9		0	0		46		0	1
France	15	2	2	28	20		0	0		a		25	5
Italy	29	0	3	15	4		0	3		35		2	6
Greece	10	0	0	39	6		0	0		U		44	0
Spain	8	0	0	39	16		0	10		0		21	0
Portugal	15	0	1	31	38		0	0		1		11	0

Class knowledge:

Group 1: Nordic countries

Group 2: UK, France, Greece, Spain,

Portugal

Group 3: Benelux countries,

Switzerland, Austria, Italy, Germany





Systems biology

- Lecture 10
- The systematic study of complex interactions in biological systems (integration and holism instead of reduction)



Systems biology





Interacting genes/protein/metabolites

Emergent properties: differential expression



Emergent properties: AND logics in regulation



Lorrelation between the gene and	
regulator 1:	0.55 (P < 0.06)
regulator 2:	0.65 (P < 0.02)

Correlation between the gene and	
linear model:	0.77 (P < 0.003)
non-linear model:	0.91 (P < 1.2E-05)



= degree exponent

Systems biology

Scale free networks

Laws of genome evolution

- Log-normal distribution of the А. evolutionary rates between orthologous genes
- Negative correlation between B. gene sequence evolution rate and expression level (or protein abundance)
- Power law-like distributions of C membership in paralogous gene families and node degree in biological networks
- D. Distinct scaling of functional classes of genes with genome size

Review Are There Laws of Genome Evolution?

Eugene V. Koonin

OPEN & ACCESS Freely available online

National Library of Medicine, National Institutes of Health, Bethesda, Maryland, United States of America



0. No dependence: translation

- 1. Linear dependence: enzymes
- 2. Quadratic dependence:
 - regulation/signaling

Koonin. PLoS Computational Biology 7:e1002173, 2011.

PLOS COMPUTATIONAL BIOLOGY

Computer programming

Algorithm

- Algorithm: a sequence of instructions that one must perform in order to solve a well-formulated problem
- Correct algorithm: translate every input instance into the correct output
- Incorrect algorithm: there is at least one input instance for which the algorithm does not produce the correct output
- Many successful algorithms in bioinformatics are incorrect algorithms

Programs

- Algorithms are implemented in a programming language to form programs
- Programs consists of:
 - Variables: names with values (float, integer, string) or arrays/tables/hashes of values
 - Conditional statements: IF-THEN-ELSE
 - Loops: while, for, until, etc.
 - Modularity: procedures/functions/subroutines/objects/methods
- Pseudo-code: programming language-independent, often used to sketch a program using pen and paper

Pseudo-code

Sorting problem: Sort a list of *n* integers: $a = (a_1, a_2, ..., a_n)$ e.g. a = (7,92,87,1,4,3,2,6)

SelectionSort(*a*,*n*)

- 1 for $i \leftarrow 1$ to n-1
- 2 $j \leftarrow$ Index of the smallest element

among $a_i, a_{i+1}, \ldots, a_n$

- 3 Swap elements a_i and a_j
- 4 return *a*

Example run

i = 1: (7,92,87,1,4,3,2,6) i = 2: (1,92,87,7,4,3,2,6) (1,2,87,7,4,3,92,6) i = 3: i = 4: (1,2,3,7,4,87,92,6) i = 5: (1,2,3,4,7,87,92,6) i = 6: (1,2,3,4,6,87,92,7) i = 7: (1,2,3,4,6,7,92,87) (1,2,3,4,6,7,87,92)

Syntax versus semantics

- Syntax: the rules for constructing valid statements in a programming language
- Semantics: the meaning of a program
- A specific algorithm implemented in different programming languages would use different syntax, but have the same semantics
- Syntax is easy and can be checked before execution (the interpreter will tell you when you make syntax mistakes)
- Semantics is hard and "bugs" typically only reveal themselves at execution time

Programming languages

- Imperative programming: describes computation as statements that change a program state (e.g. Perl, Fortran, C, and Java)
- Functional programming: treats computation as the evaluation of (mathematical) functions, and often avoids state (e.g. LISP)
- Declarative programming: while imperative programs explicitly specify an algorithm to achieve a goal, declarative programs explicitly specify the goal and leave the implementation of the algorithm to the support software (e.g. PROLOG)

Sorting: imperative

Sorting problem: Sort a list of *n* integers: $a = (a_1, a_2, ..., a_n)$

SelectionSort(*a*,*n*)

- 1 for $i \leftarrow 1$ to n-1
- 2 $j \leftarrow$ Index of the smallest element

among $a_i, a_{i+1}, \ldots, a_n$

- 3 Swap elements a_i and a_j
- 4 return *a*

Pseudo-code hides ugly details such as

"Swap elements a_i and a_j "

$$\begin{array}{cccc}
1 & tmp \leftarrow a_j \\
2 & a_j \leftarrow a_i \\
3 & a_i \leftarrow tmp
\end{array}$$

or

" $j \leftarrow$ Index of the smallest element among a_i, a_{i+1}, \dots, a_n "

IndexOfMin(array, first, last) $index \leftarrow first$ $for k \leftarrow first + 1 to last$ $if array_k < array_{index}$ $index \leftarrow k$ return index

Remember, though, that the devil is in the details!

Recursion

RecursiveSelectionSort(*a*,*first*,*last*)

- 1 **if** (first \leq last)
- 2 *index* \leftarrow Index of the smallest element among a_{first} , $a_{first+1}$, ..., a_{last}
- 3 Swap elements a_{first} and a_{index}
- 4 $a \leftarrow \text{RecursiveSelectionSort}(a, first+1, last)$
- 5 return a

Example I

Write pseudo-code for a program that solves a quadratic equation $ax^2 + bx + c = 0$:

QuadraticEquationSolver (a, b, c)

Remember that:
$$x = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$$

QuadraticEquationSolver(*a*, *b*, *c*) 1 root \leftarrow b²-4ac; 2 if root < 0 3 return "No solution" 4 x1 $\leftarrow \frac{-b+\sqrt{root}}{}$

$$5 \quad x2 \leftarrow \frac{2a}{-b - \sqrt{root}}$$

6 **if**
$$x1 = x2$$

7 **output** "Solution:
$$x = x1$$
"

8 else

9 **output** "Solutions: x = x1 or x = x2"

Example II

Write pseudo-code for a program that adds a constant to every number in an array $a = (a_1, a_2, ..., a_n)$:

AddConstant (*a*, *n*, *c*)

E.g. a = (1, 2, 4) and c = 3 outputs (4, 5, 7)

AddConstant(**list**, *n*, *c*)

- 1 for $i \leftarrow 1$ to n
- 2 $list_i \leftarrow list_i + c$
- 3 return **list**

Example III

Write pseudo-code for a program that remove duplicates in an array $\mathbf{a} = (a_1, a_2, ..., a_n)$

RemoveDuplicates (*a*, *n*)

E.g. **a** = (1, 2, 2, 4, 4) outputs (1, 2, 4)

RemoveDuplicates(**list**, *n*) 1 newlist \leftarrow () 2 for $i \leftarrow 1$ to n $m \leftarrow$ length of **newlist** 3 foundDuplicate \leftarrow false 4 5 for $j \leftarrow 1$ to m 6 **if** $list_i = newlist_i$ 7 foundDuplicate = true 8 break 9 **if** *foundDuplicate* = *false* add *list*_{*i*} to **newlist** 10 11 return **newlist**

Example IV

Write pseudo-code for a program that counts from 0 to $\mathbf{n} = (n_1, n_2, ..., n_m)$:

12

Count (*n*, *m*)

E.g.
$$\boldsymbol{n} = (1, 2)$$
 outputs: 00
10
11

Count(**n**, *m*) 1 $c \leftarrow (0, 0, ..., 0)$ 2 while forever for $i \leftarrow m$ to 1 3 **if** $c_i = n_i$ 4 $c_i \leftarrow 0$ 5 6 else 7 $c_i \leftarrow c_i + 1$ break 8 9 output c **if** c = (0, 0, ..., 0)10 11 break