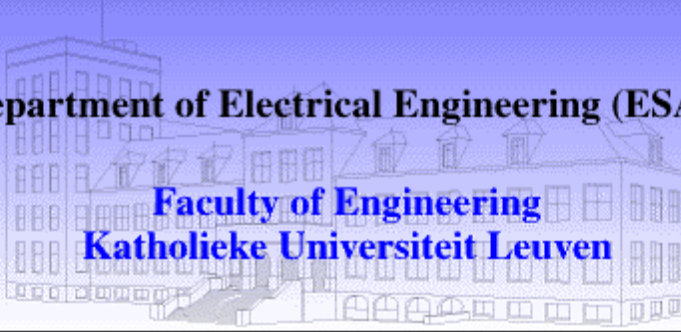




Department of Electrical Engineering (ESAT)

Faculty of Engineering
Katholieke Universiteit Leuven



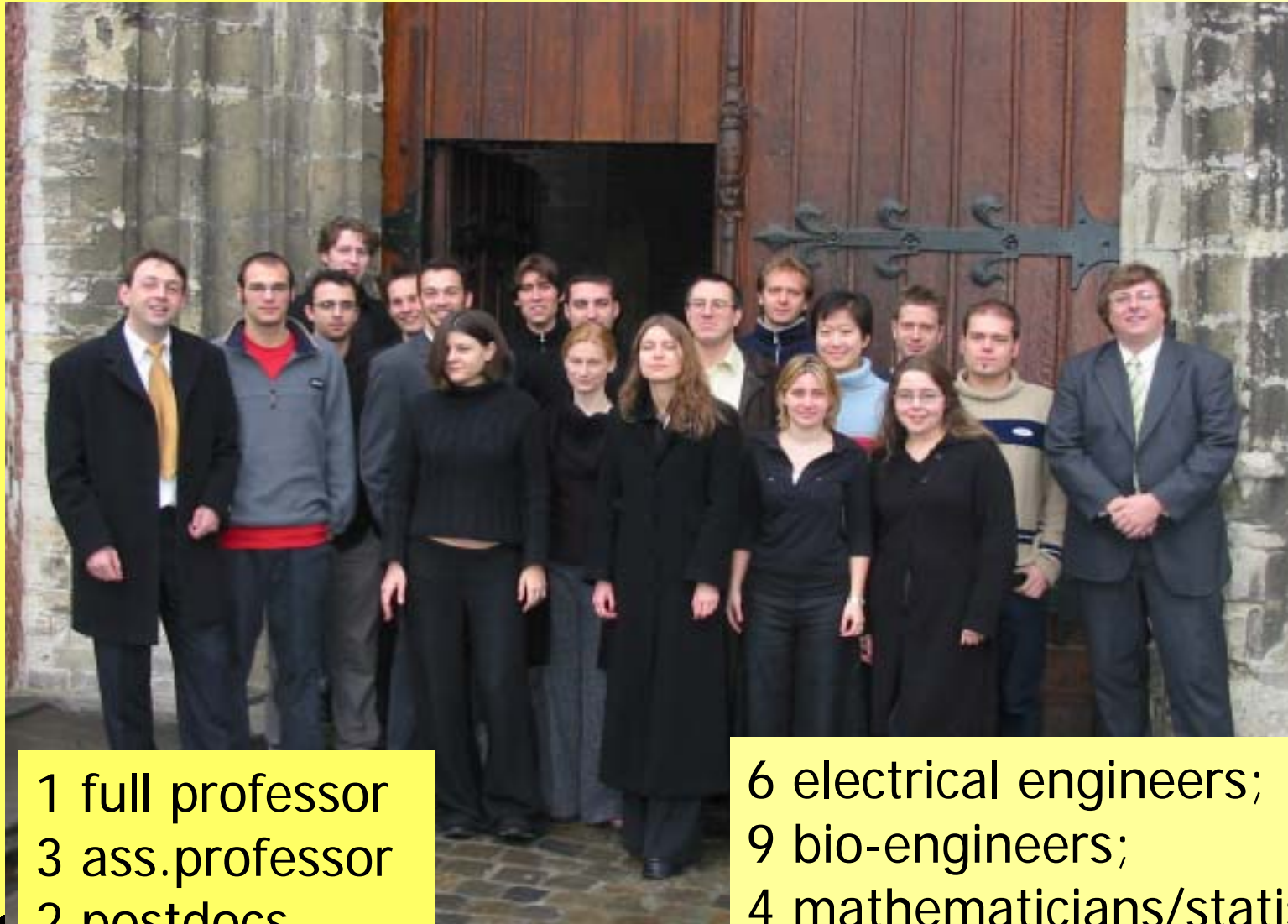
Systems Theory in Systems Biology

Bart De Moor
ESAT-SCD
Katholieke Universiteit Leuven

Bart.demoor@esat.kuleuven.ac.be
www.esat.kuleuven.ac.be/~demoor



Our team



1 full professor
3 ass. professor
2 postdocs
14 phd students

6 electrical engineers;
9 bio-engineers;
4 mathematicians/statisticians
1 medical doctor



Contents

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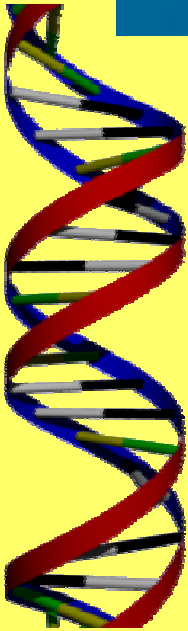
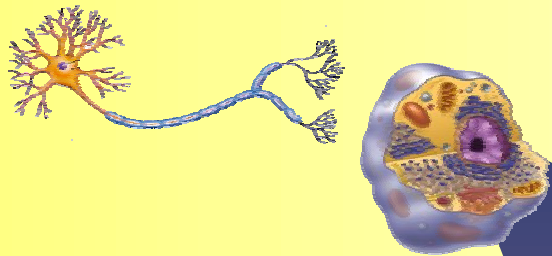
Conclusions



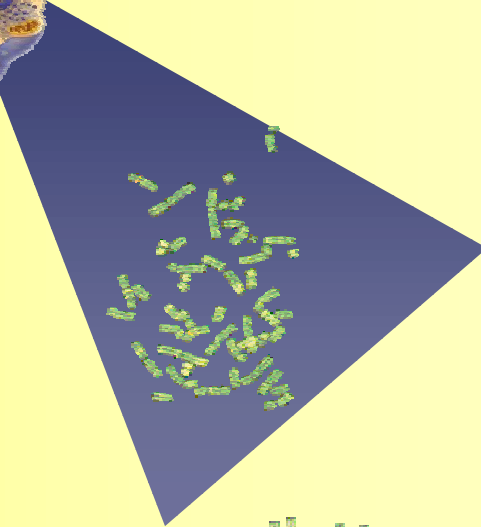
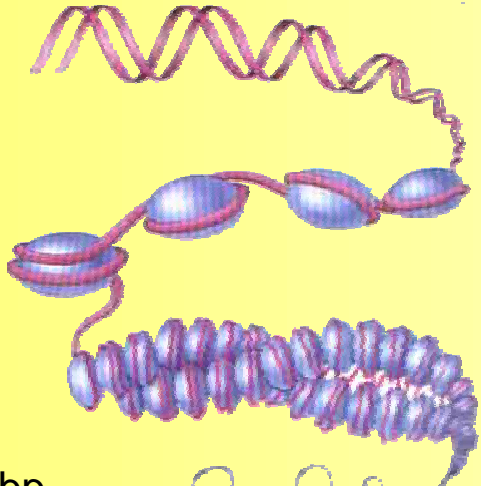
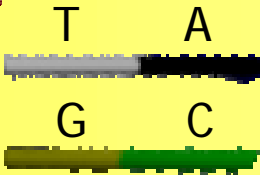
Biology



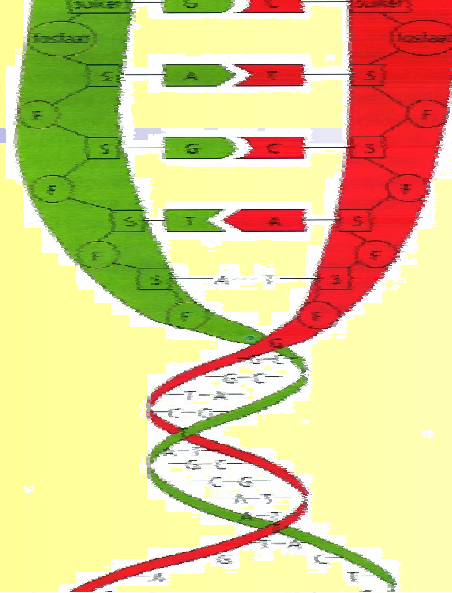
1.000.000 cell types
100.000.000.000.000 cells



3.201.762.515 bp



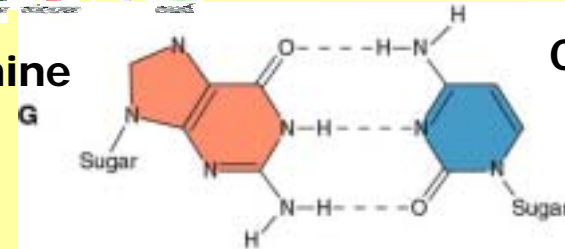
Double helix of DNA



It has not escaped our notice that the specific pairing we have postulated immediately suggests a possible copying mechanism for the genetic material.

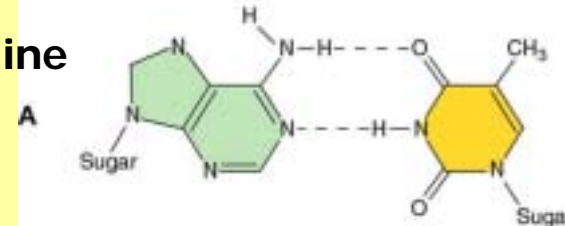


Guanine



Cytosine

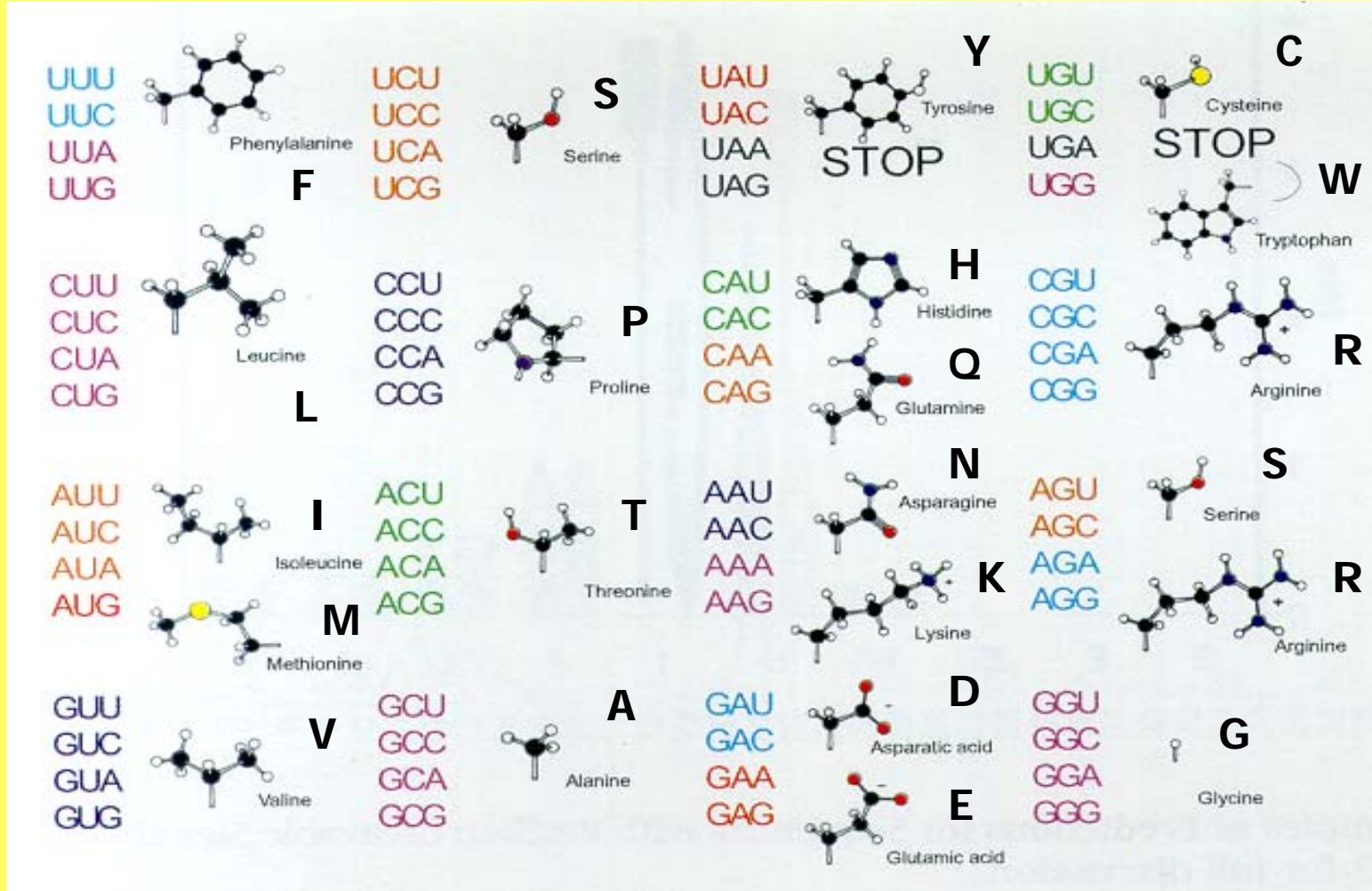
Adenine



Thymidine



Genetic (almost) universal code: codons



T in DNA
U in RNA

-20 amino acids

-64 codons: Redundancy - robustness

Stop=UAA,UAG,UGA

Start=AUG



SNP: Single Nucleotide Polymorphism

11 million SNPs / 3 billion nucleotides

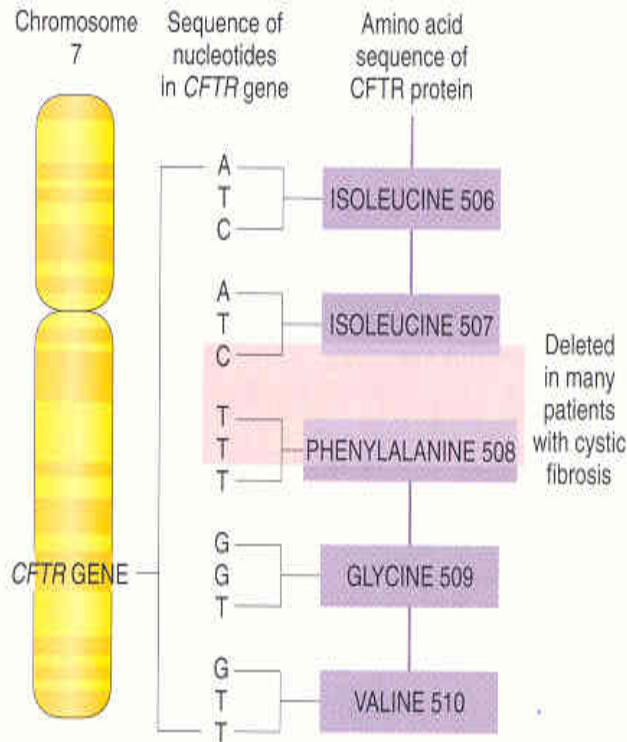
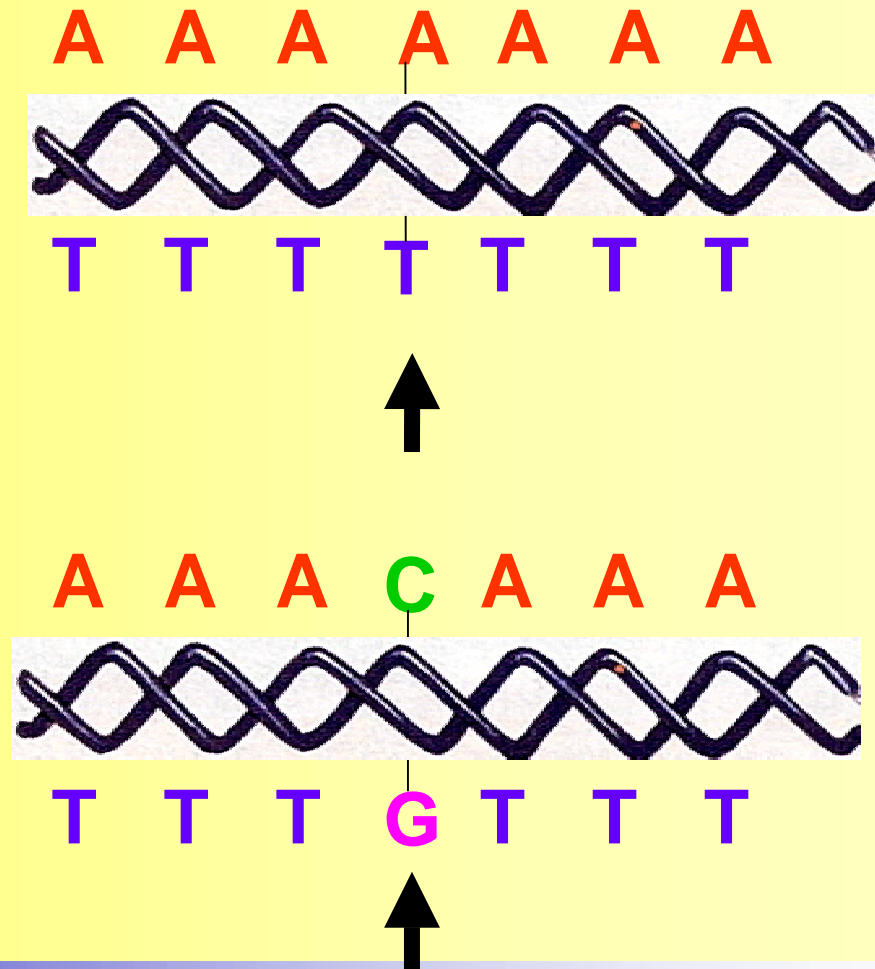


Figure 3-28



Monogenic diseases

- 1983: Huntington
- 1986: Muscular dystrophy
- 1989: Cystic-fibrosis



The genome

.....ACACATTAATCTTATATGCTAAAACCTAGGTCTCGTTTTAGGGATGTTTATAA
CCATCTTTGAGATTATTGATGCATGGTTATTGGTTAGAAAAATATACGCTTGTTTTCTTTCCT
AGGTTGATTGACTCATAACATGTGTTTCATTGAGGAAGGAACCTAACAAAACTGCACTTTTTTCA
ACGTCACAGCTACTTTAAAAGTGATCAAAGTATATCAAGAAAGCTTAATATAAAGACATTTGTT
TCAAGGTTTCGTAAGTGCACAATATCAAGAAGACAAAAATGACTAATTTTGTTTTTCAGGAAGC
ATATATATTACACGAACACAAATCTATTTTTGTAATCAACACCGACCATGGTTCGATTACACAC
ATTAATCTTATATGCTAAAACCTAGGTCTCGTTTTAGGGATGTTTATAACCATCTTTGAGATTAT
TGATGCATGGTTATTGGTTAGAAAAATATACGCTTGTTTTCTTTCCTAGGTTGATTGACTCAT
ACATGTGTTTCATTGAGGAAGGAACCTAACAAAACTGCACTTTTTTCAACGTCACAGCTACTTT
AAAAGTGATCAAAGTATATCAAGAAAGCTTAATATAAAGACATTTGTTTCAAGGTTTCGTAAGT
GCACAATATCAAGAAGACAAAAATGACTAATTTTGTTTTTCAGGAAGCATATATATTACACGAAC
ACAAATCTATTTTTGTAATCAACACCGACCATGGTTCGATTACACACATTAATCTTATATGCT
AAAACCTAGGTCTCGTTTTAGGGATGTTTATAACCATCTTTGAGATTATTGATGCATGGTTATTG
GTTAGAAAAATATACGCTTGTTTTCTTTCCTAGGTTGATTGACTCATAACATGTGTTTCATTGA
GGAAGGAACCTAACAAAACTGCACTTTTTTCAACGTCACAGCTACTTTAAAAGTGATCAAAGT
ATATCAAGAAAGCTTAATATAAAGACATTTGTTTCAAGGTTTCGTAAGTGCACAATATCAAGAA
GACAAAAATGACTAATTTTGTTTTTCAGGAAGCATATATATTACACGAACACAAATCTATTTTTG
TAATCAACACCGACCATGGTTCGATTAACACATTAATCTTATATGCTAAAACCTAGGTCTCGTT
TTAGGGATGTTTATAACCATCTTTGAGATTATTGATGCATGGTTATTGGTTAGAAAAATATAC
GCTTGTTTTCTTTCCTAGGTTGATTGACTCATAACATGTGTTTCATTGAGGAAGGAACCTAACAA
AACTGCACTTTTTTCAACGTCACAGCTACTTTAAAAGTGATCAAAGTATATCAAGAAAGCTTA
ATATAAAGACATTTGTTTCAAGGTTTCGTAAGTGCACAATATCAAGAAG.....



..... Humane Genome

- +/- 30 000 genes of 60 – 120 kB;
- only 3 % DNA = gene (exon: codes for protein);
- rest = intergenic (introns, regulatory elements, see later);
- Each person's genome is 99.8 % identical to everyone else's;



... Some genome numbers

Group	Species	Genes	Genome (Mbase)
Phages	Bacteriophage MS2	4	0.003560
Viruses	HIV Type 2	9	0.009671
Bacteria	Haemophilus influenzae (1995)	1760	1.83
Archaea	Methanococcus jannaschii	1735	1.74
Fungi	Saccharomyces cerevisiae (yeast) (1996)	5800	12.1
Protoctista	Oxytricha similis	12000	600
Arthropoda	Drosophila melanogaster (fruit fly) (2000)	12000	165
Nematoda	Caenorhabdiis elegans (Round worm)(1998)	14000	100
Mollusca	Loligo Pealii	35000	2700
Plantae	Arabidopsis thaliana (Mustard cress)(2000)	25000	70-145
Chordata	Homo Sapiens	30000	3000

Estimated 265-350 genes are required for 'life'.



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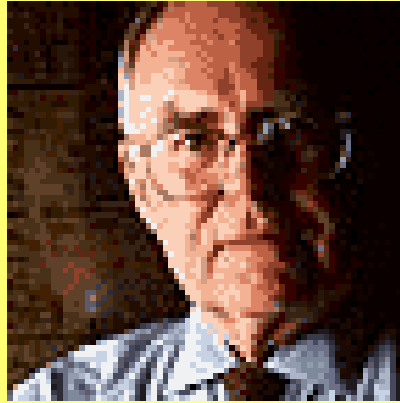
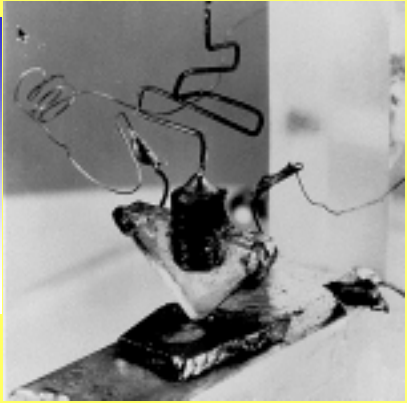
Bioinformatics

Systems biology

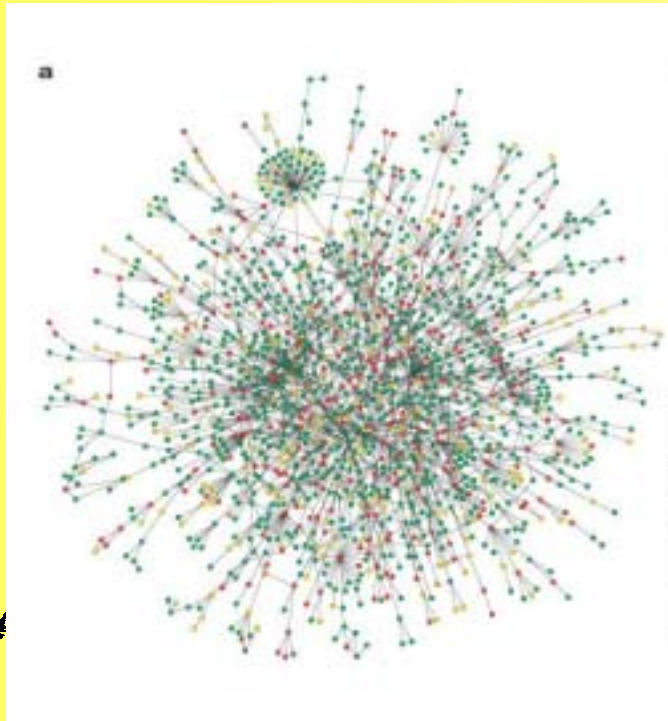
Conclusions



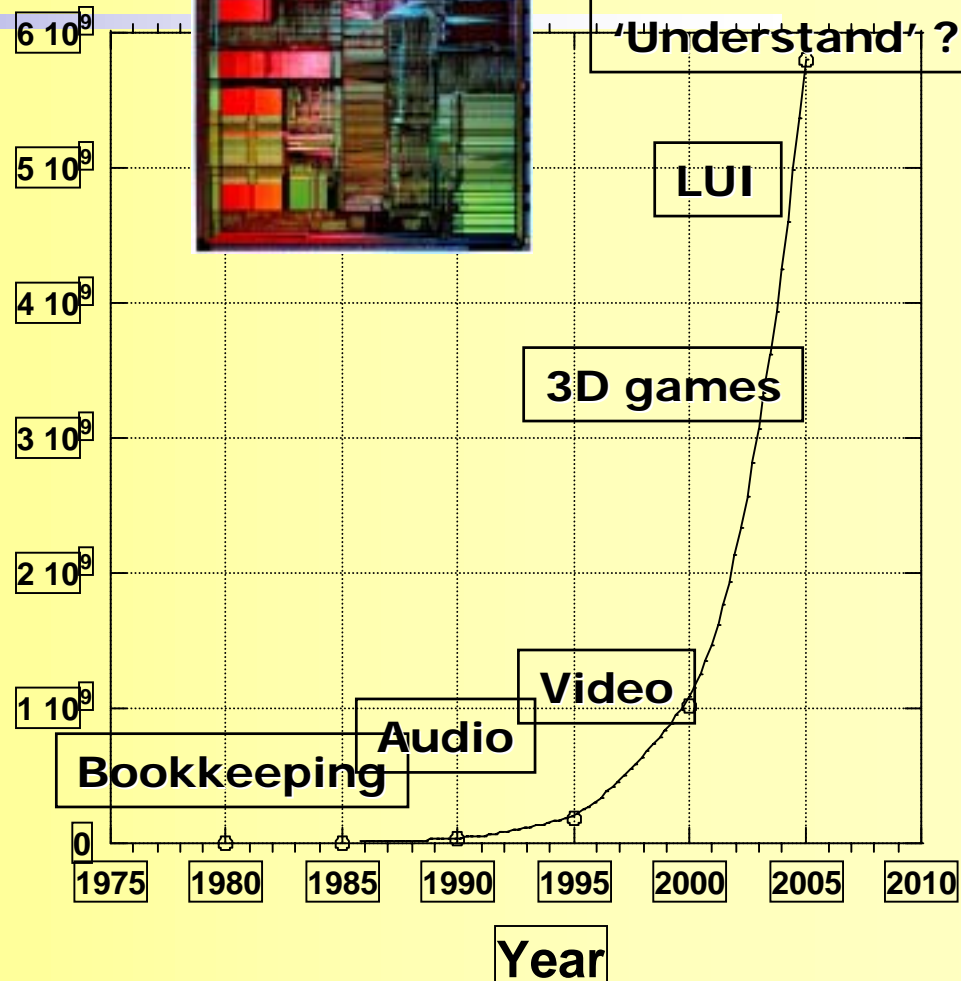
Moore's law



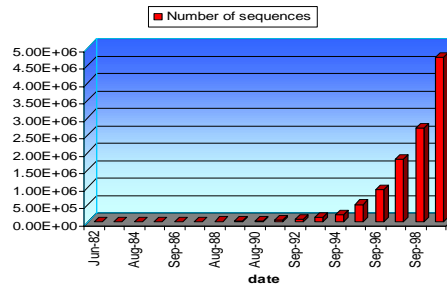
Small World



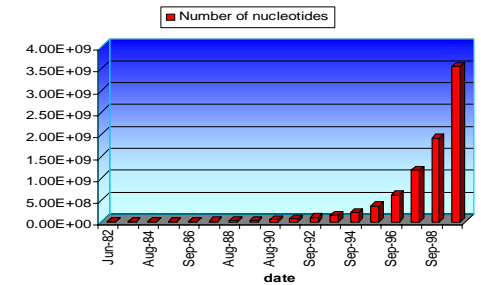
Operations/second



Database growth: Number of sequences



Database growth: Number of nucleotides



Mathematics and biology



1865: Mendel's Laws = statistics



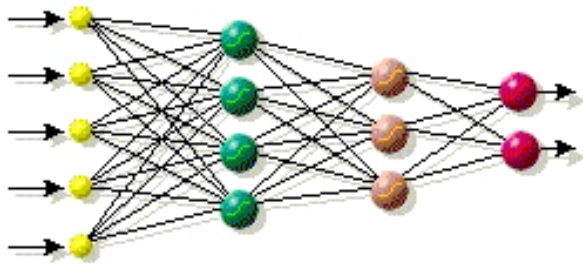
Shannon: 1940 PhD
An algebra for theoretical genetics



1952: Turing
The chemical basis of morphogenesis



1944: Schrödinger: What's life ?



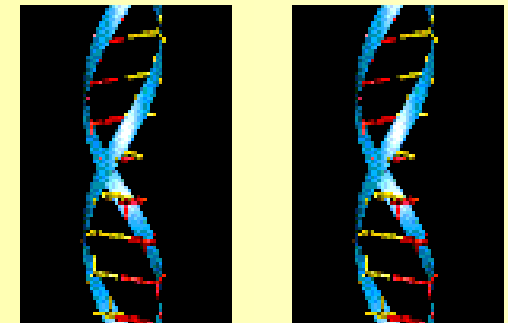
Neural networks !

Genetic algorithms = optimization by 'survival of the fittest'

Parent A

Parent B

Offspring



DNA computers



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Mathematics

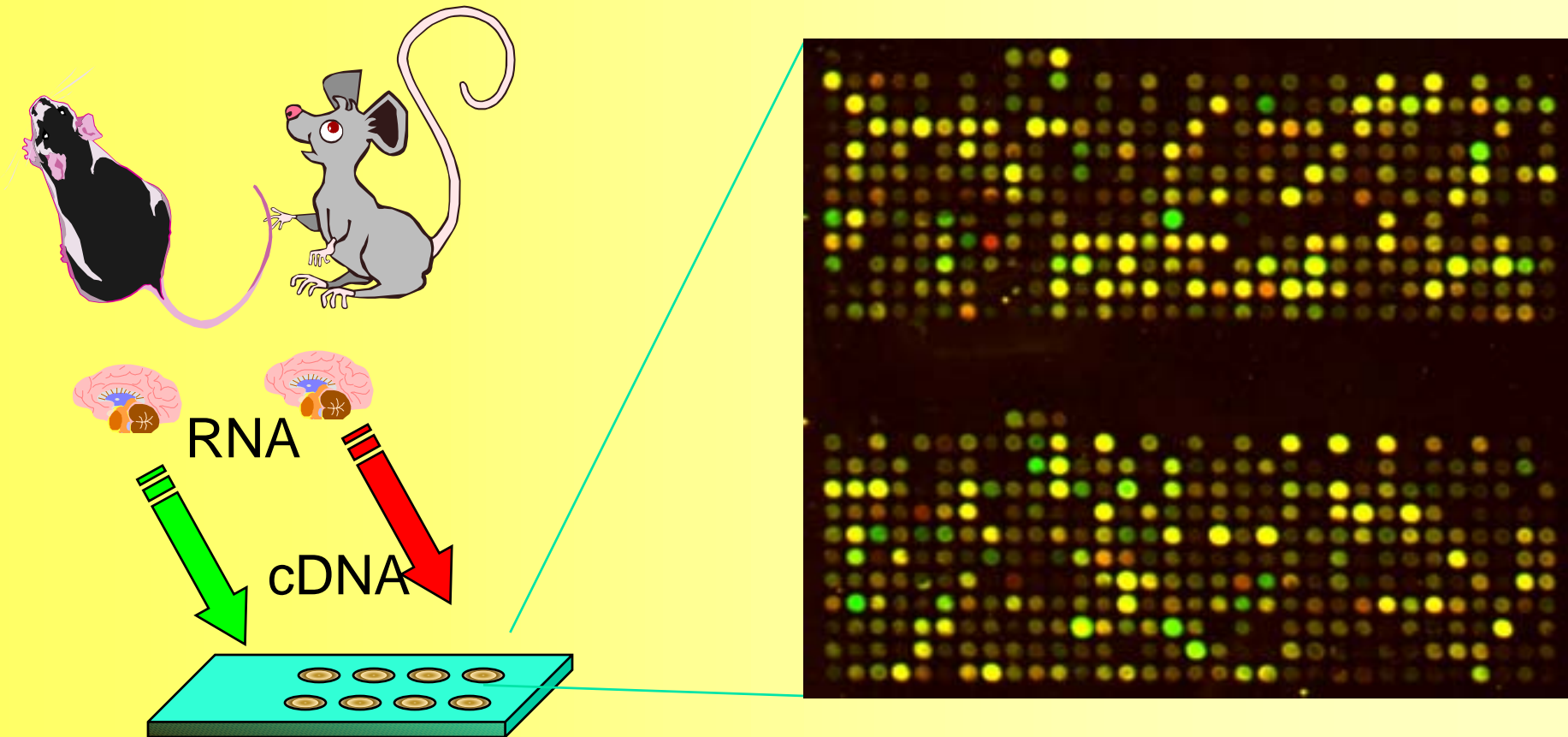
Bioinformatics

Systems biology

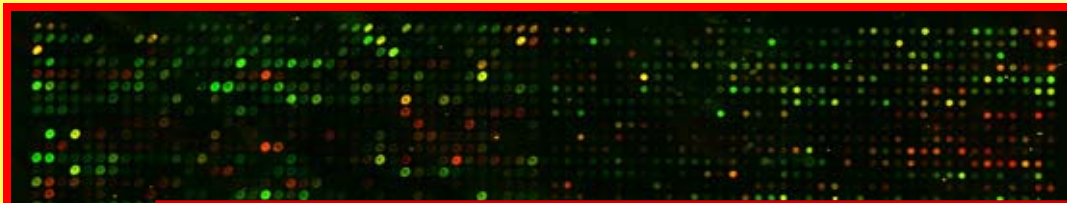
Conclusions



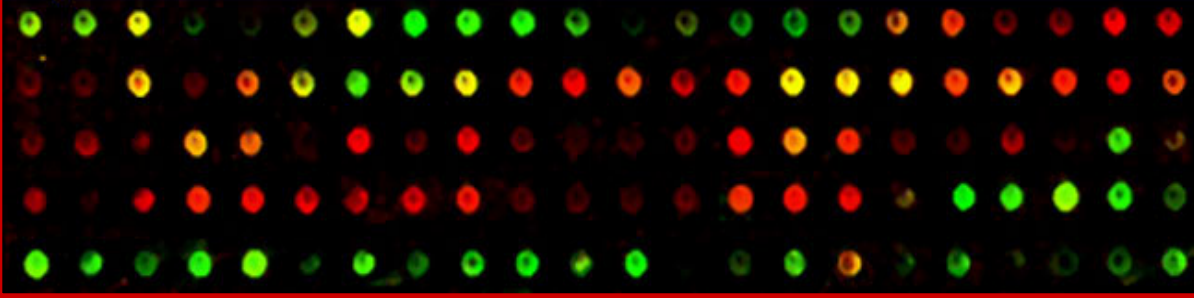
Differentially expressed genes

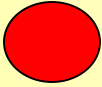
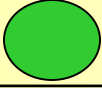
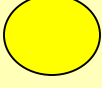



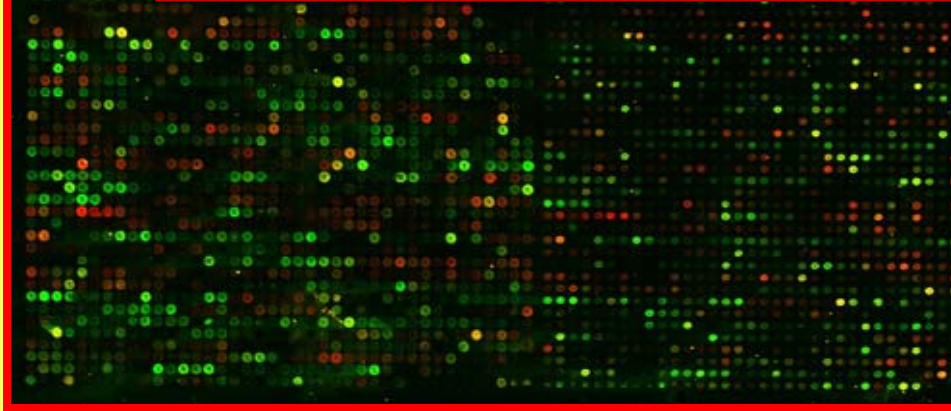
Technology: Microarrays/DNA-chips



Two color hybridization on a yeast array with two differing samples of genomic DNA.



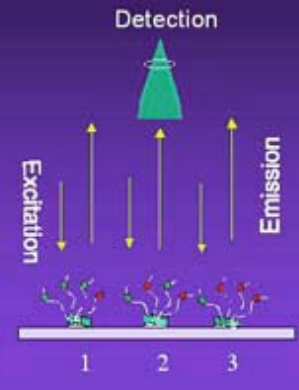
	Test	Ref.
	High	Low
	Low	High
	High	High
	Low	Low



Relative Abundance Detection



Mix and hybridize



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Bio-informatics

- High-throughput technology → lots of 'wet lab' data
- Computers → computing power
- Internet → Publicly accessible databases
- Applied mathematics, statistics, numerical algorithms, machine learning, data mining

Some cases / examples:

- Clinical bio-i: Classification of leukemia
- Gene regulation bio-i: Finding motifs in DNA sequences



Example: Classification of leukemia

12 600 genes

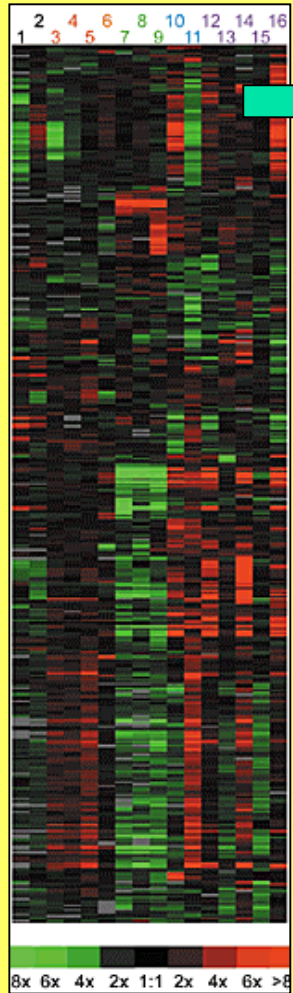
72 patients:

- 28 Acute Lymphoblastic Leukemia (ALL)
- 24 Acute Myeloid Leukemia (AML)
- 20 Mixed Linkage Leukemia (MLL)

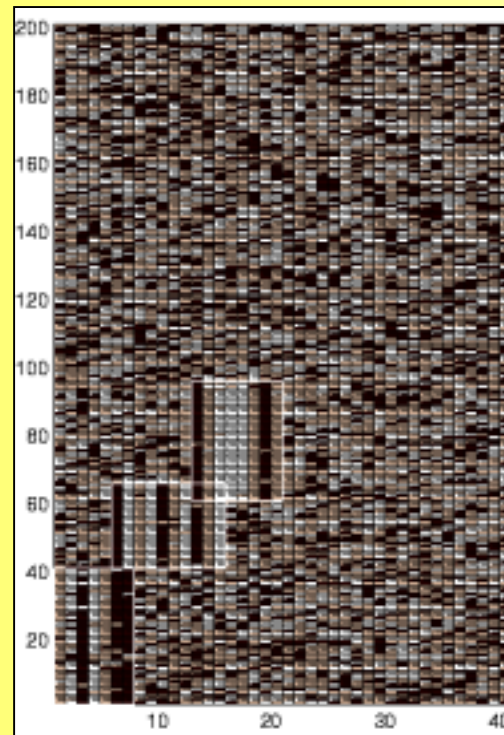


Pattern recognition algorithms

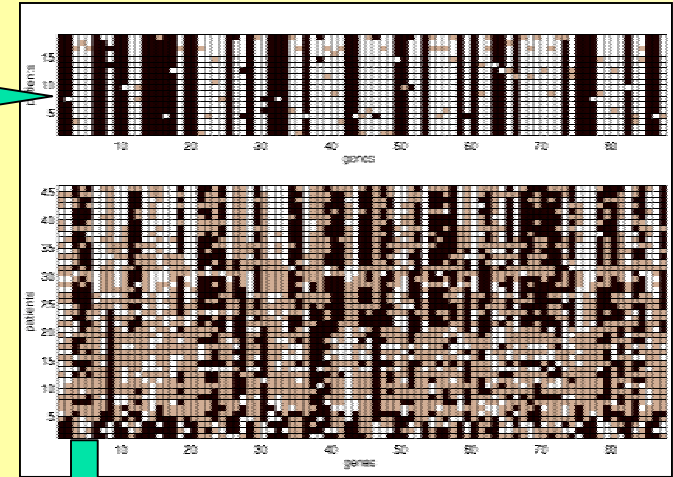
Data matrix



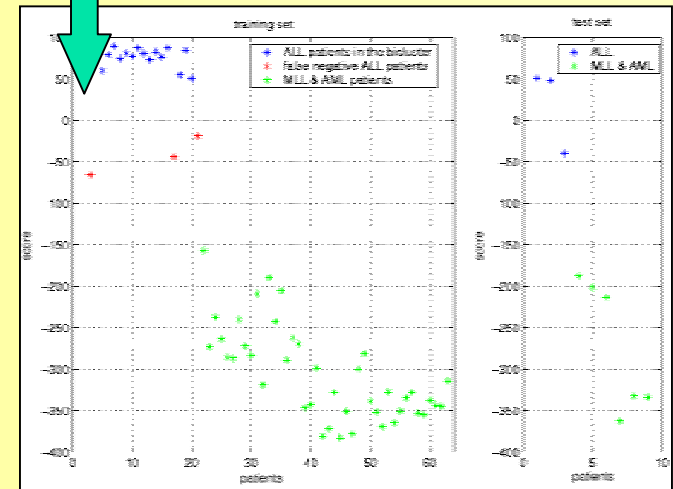
Hidden pattern



Find the pattern

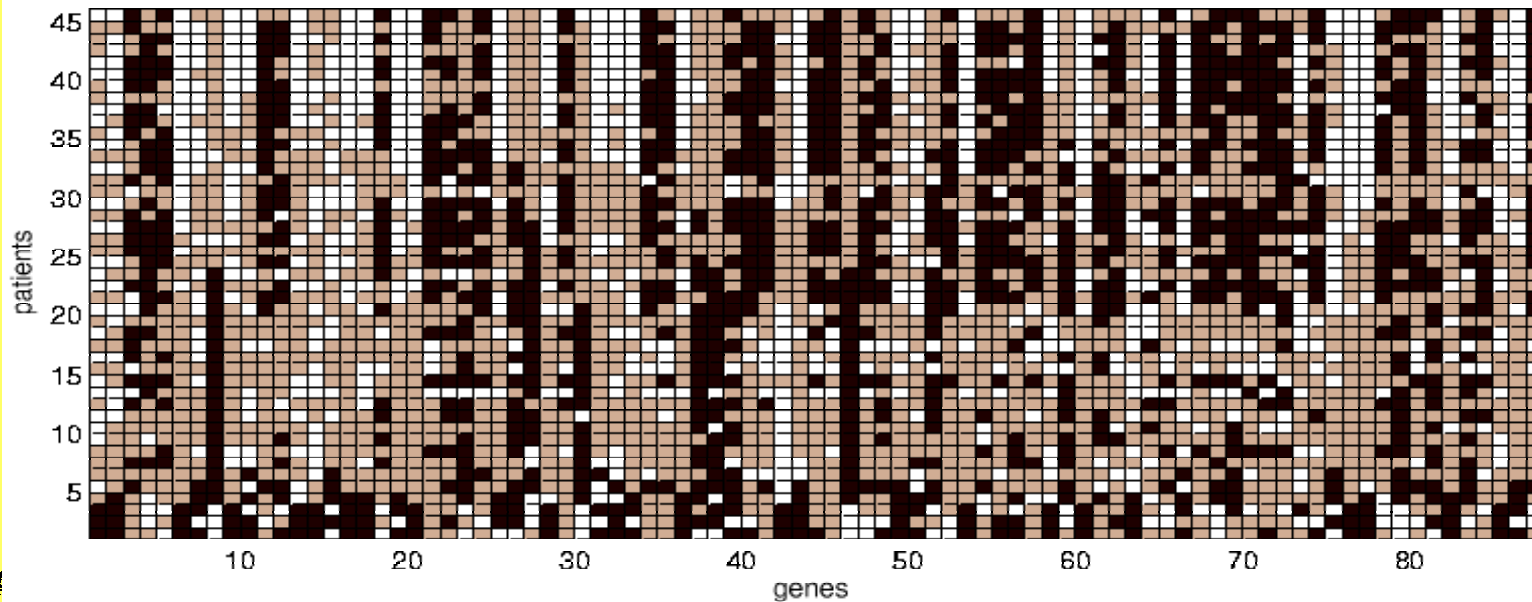
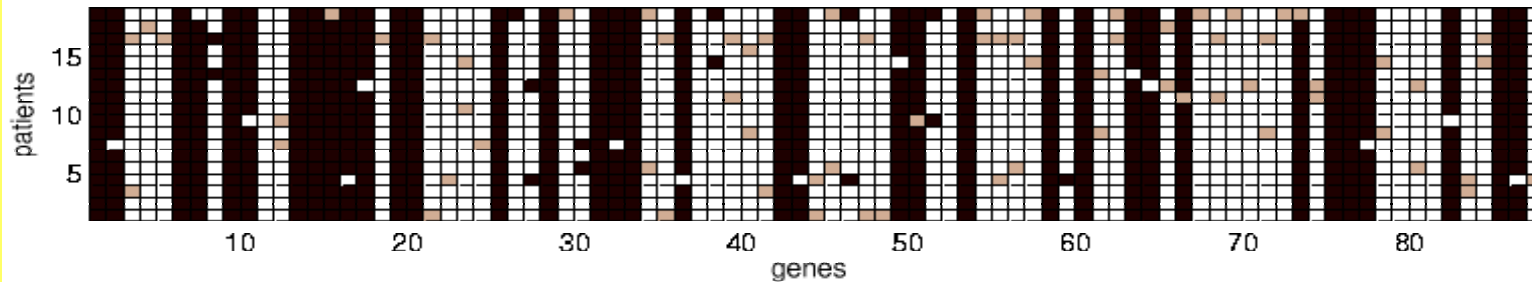


Pattern validation



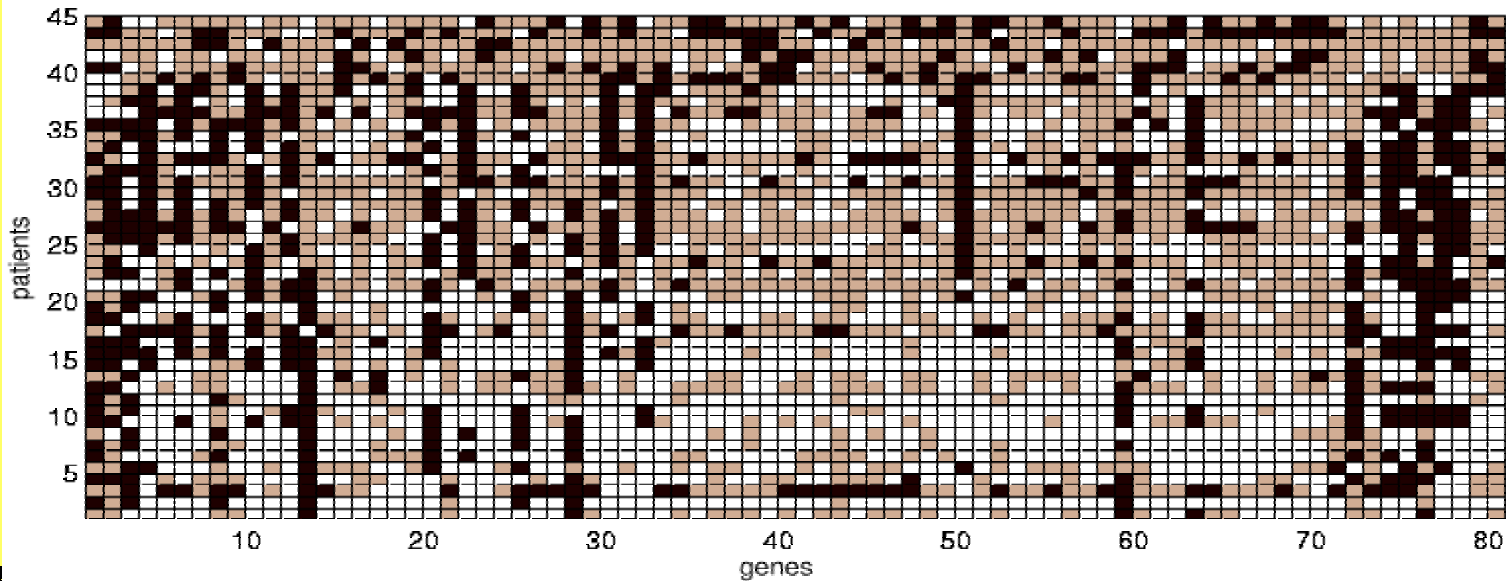
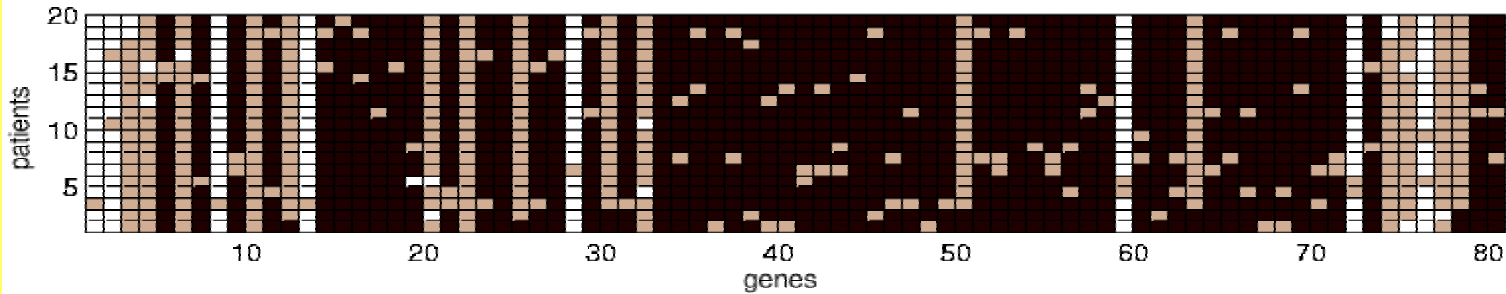
AML Pattern (=fingerprint)

18 AML patients (of 21) with 87 genes



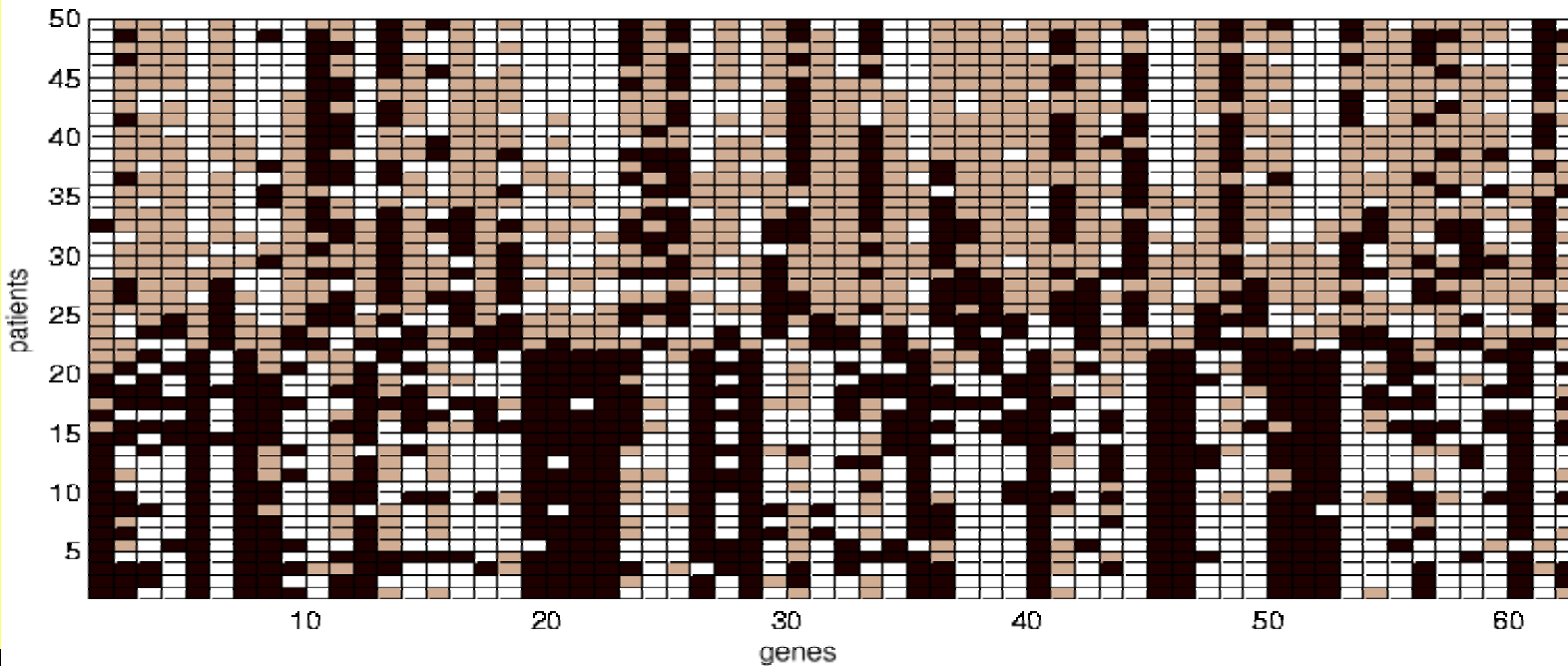
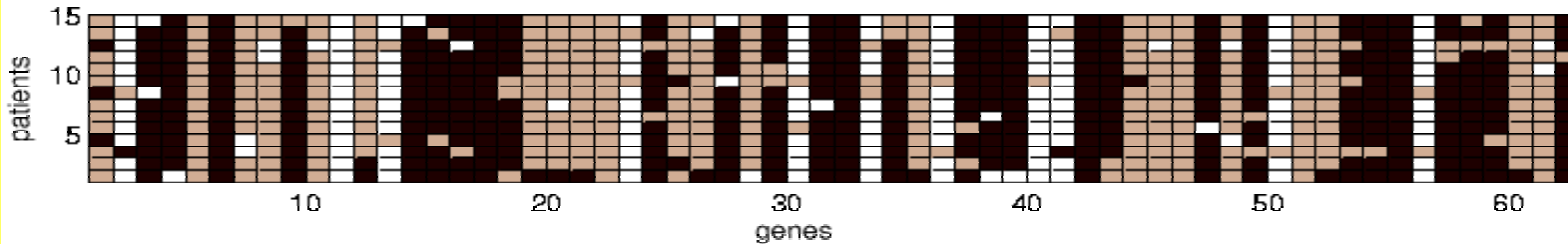
ALL pattern (=fingerprint)

19 ALL patienten (of 25) with 80 genes

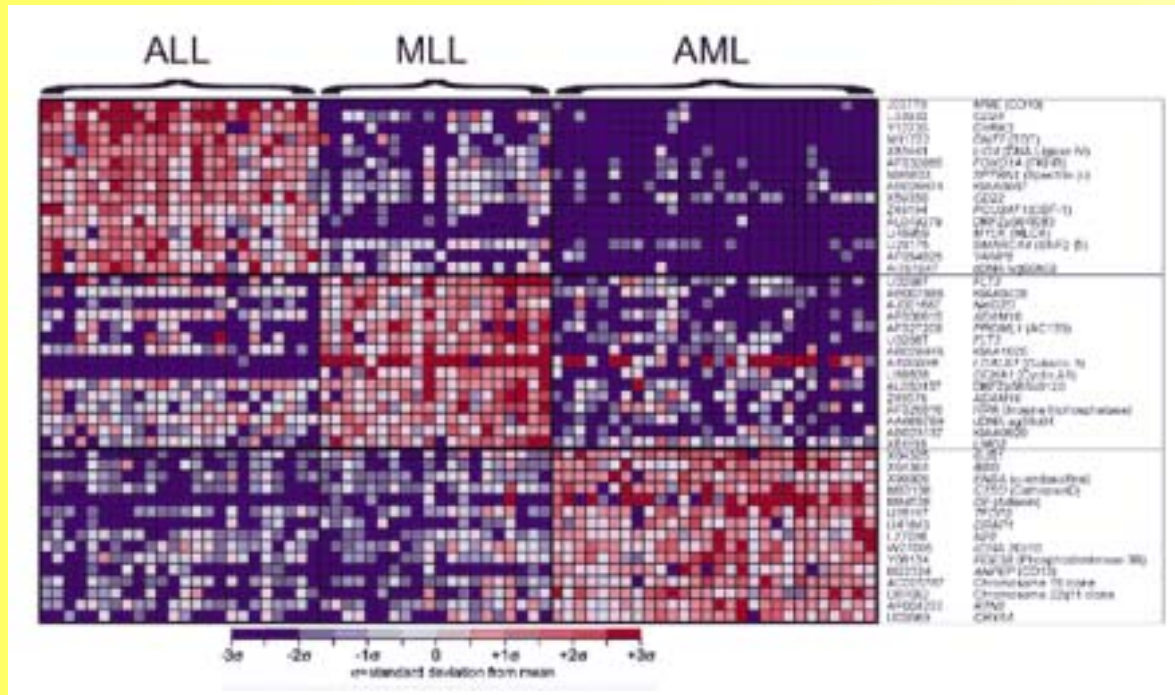


MLL pattern (=fingerprint)

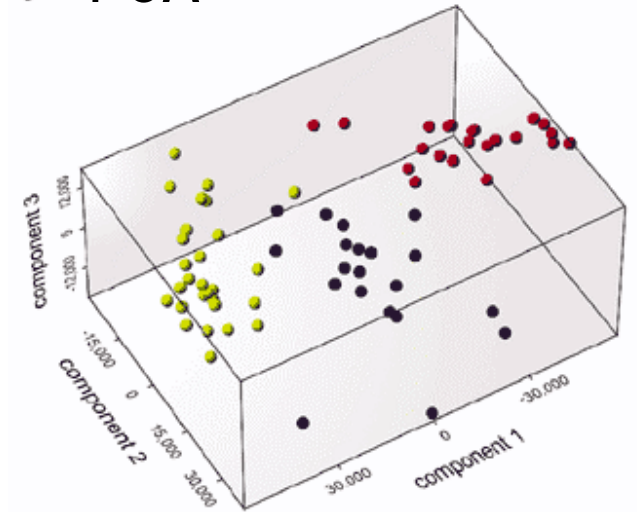
14 MLL patienten (of 17) with 62 genes



ALL/AML/MLL dataset



b PCA



© Armstrong SA et al. Nat Genet. 2002 Jan;30(1):41-7.

12 600 genes

72 patients:

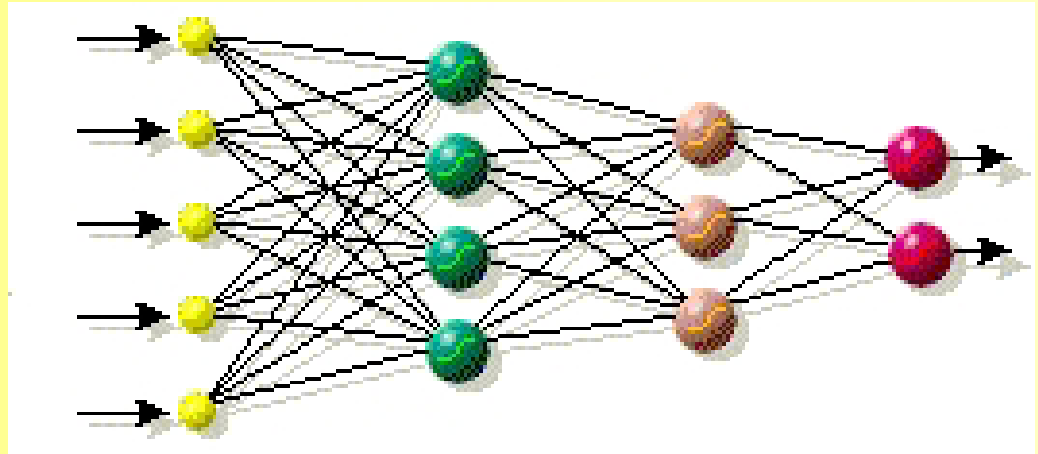
- 28 Acute Lymphoblastic Leukemia (ALL)
- 24 Acute Myeloid Leukemia (AML)
- 20 Mixed Linkage Leukemia (MLL)

3 patients for each class used as test set



How many genes needed for diagnosis ?

Neural net



number of genes	% area ROC training	% area ROC prospective
20	1	1
15	1	1
10	1	99.29
5	1	98.57
4	1	98.57
3	1	97.50
2	98.32	98.21
1	93.60	71.07

**Relevance:
Diagnostic kit**



Bio-informatics

- High-throughput technology → lots of 'wet lab' data
- Computers → computing power
- Internet → Publicly accessible databases
- Applied mathematics, statistics, numerical algorithms, machine learning, data mining

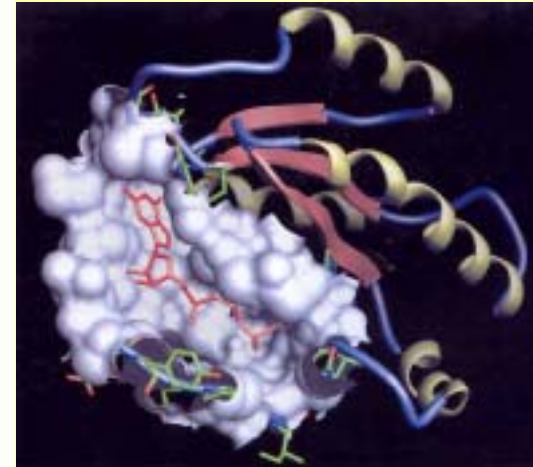
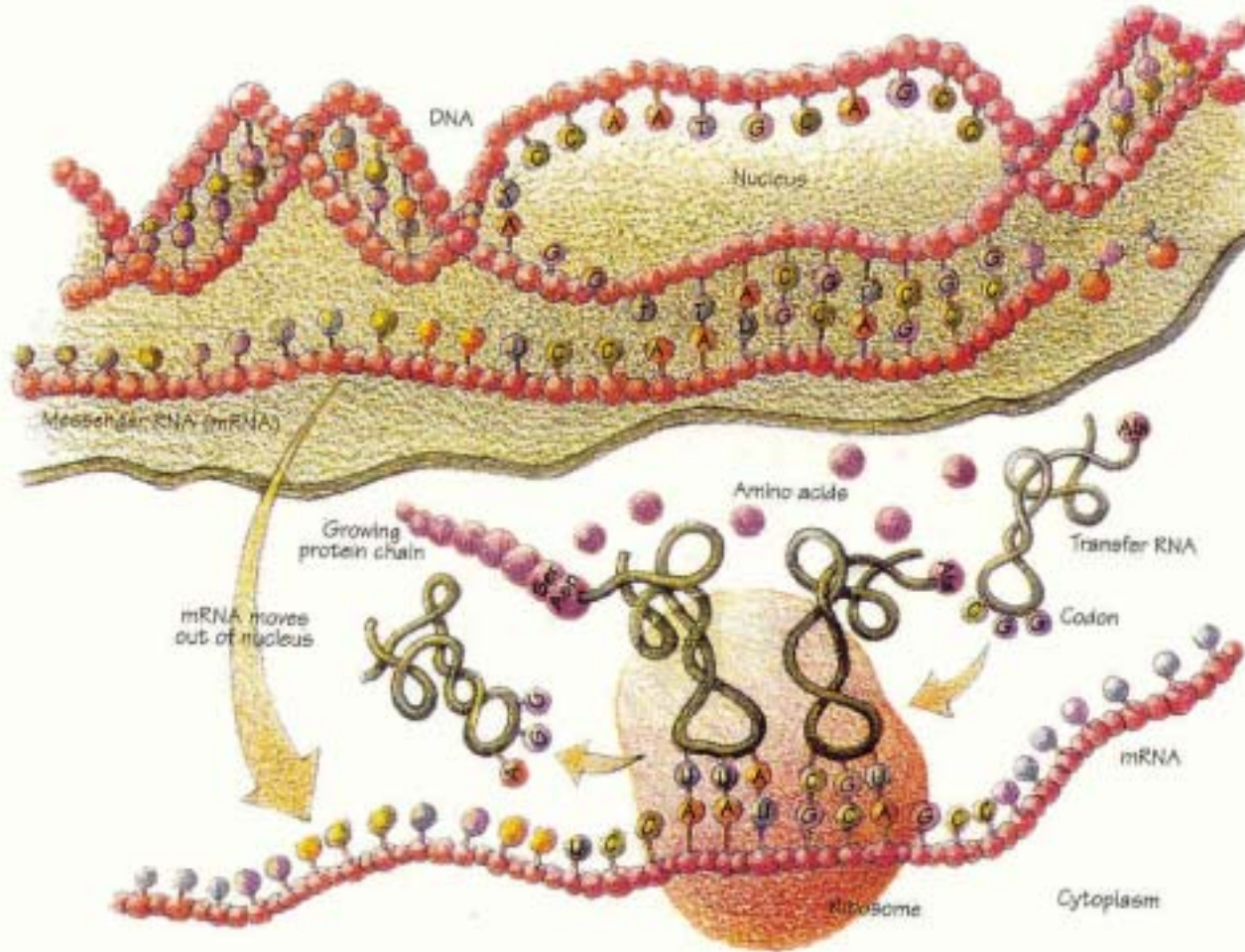
Some cases / examples:

- Clinical bio-i: Classification of leukemia
- Gene regulation bio-i: Finding motifs in DNA sequences



Central dogma (Crick, 1958)

DNA – mRNA – codon – amino-acid - protein



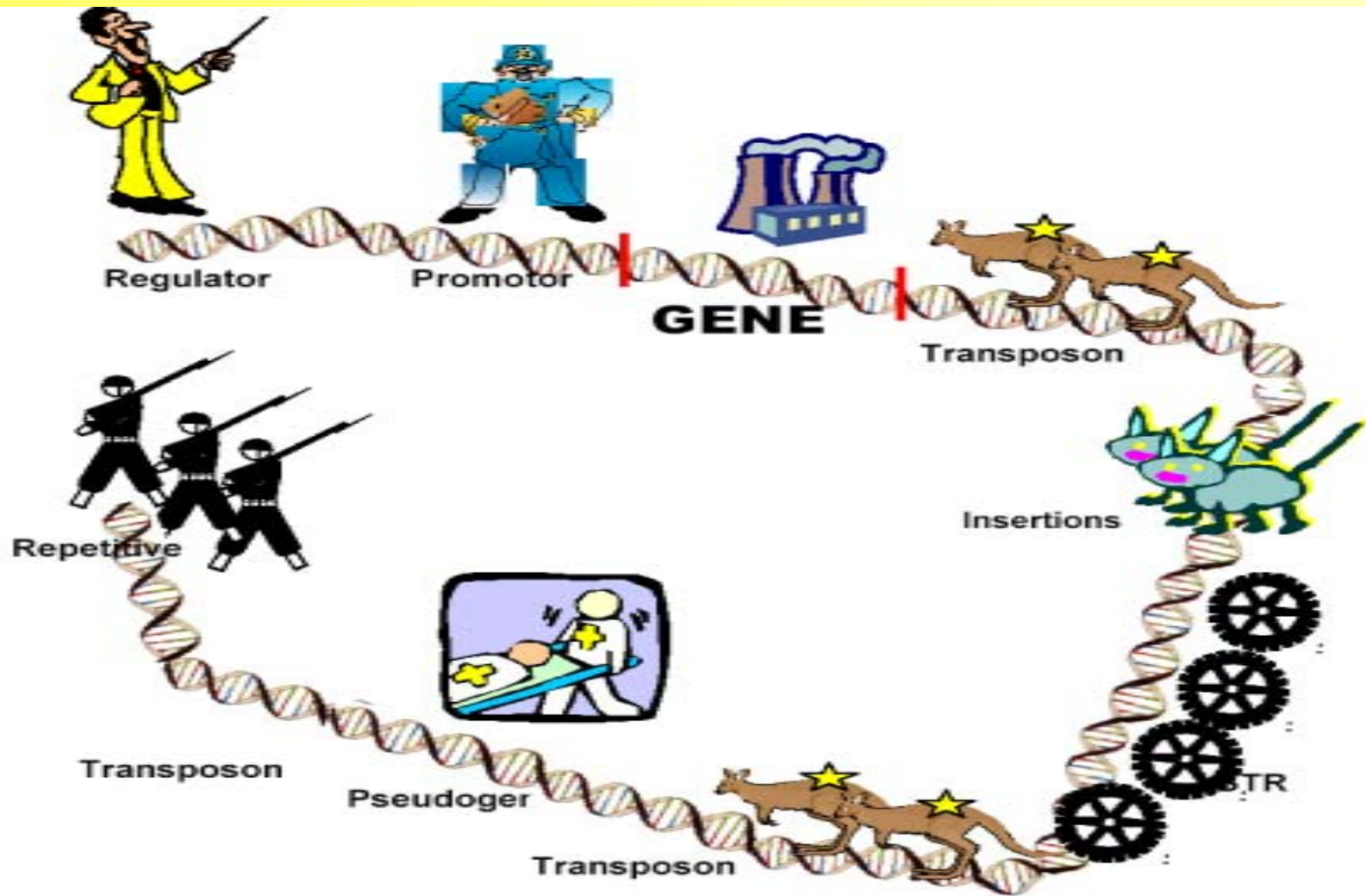
Protein:

- linear polymer
- 100000s
- 3D-Folding / docking
- 'workhorse'

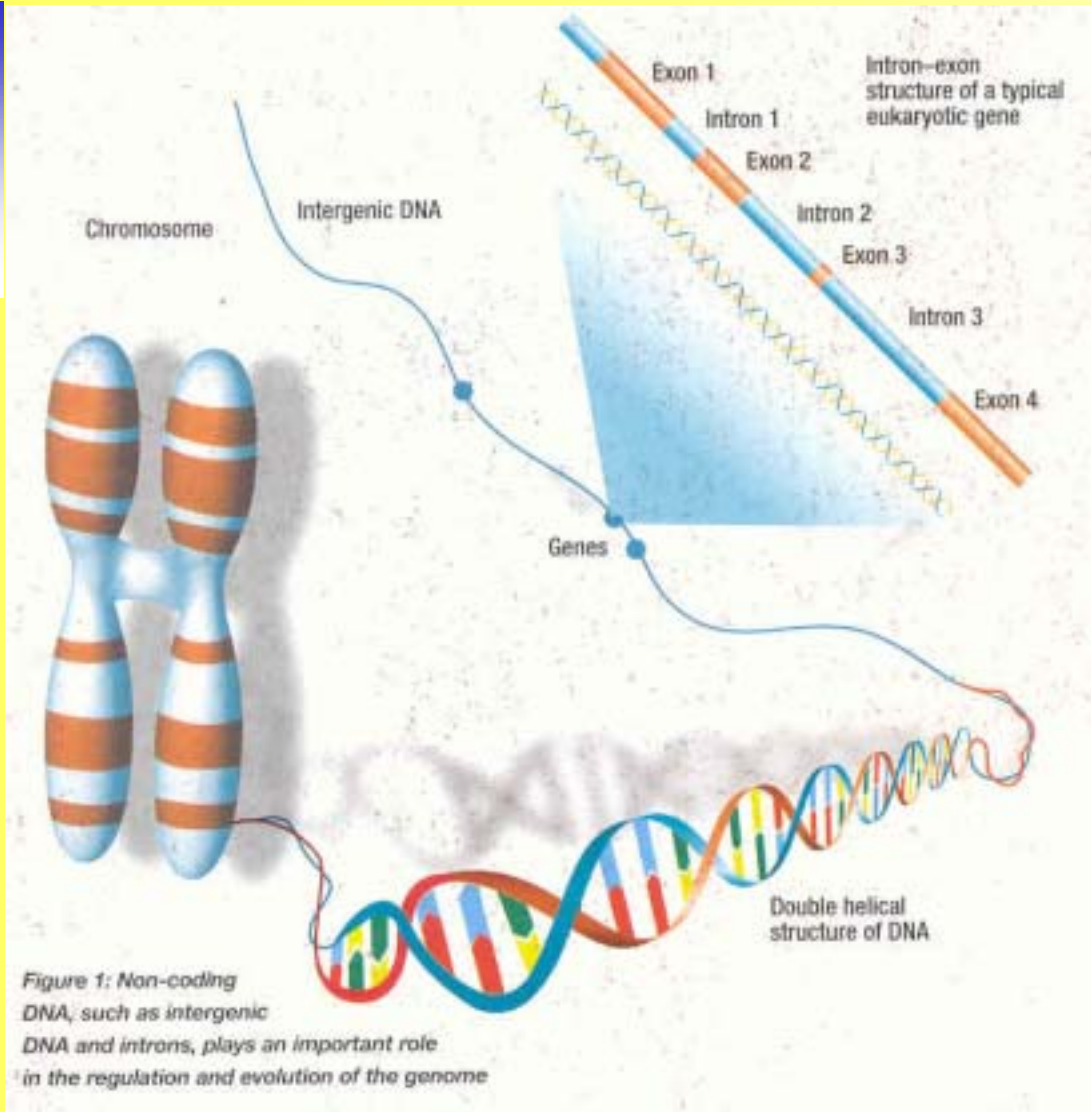
Exceptions exist: e.g. retrovirus (HIV)



Detecting regulatory elements



Junk DNA ?



**3 % of human genome: genes
97 % non-coding**

Introns contain

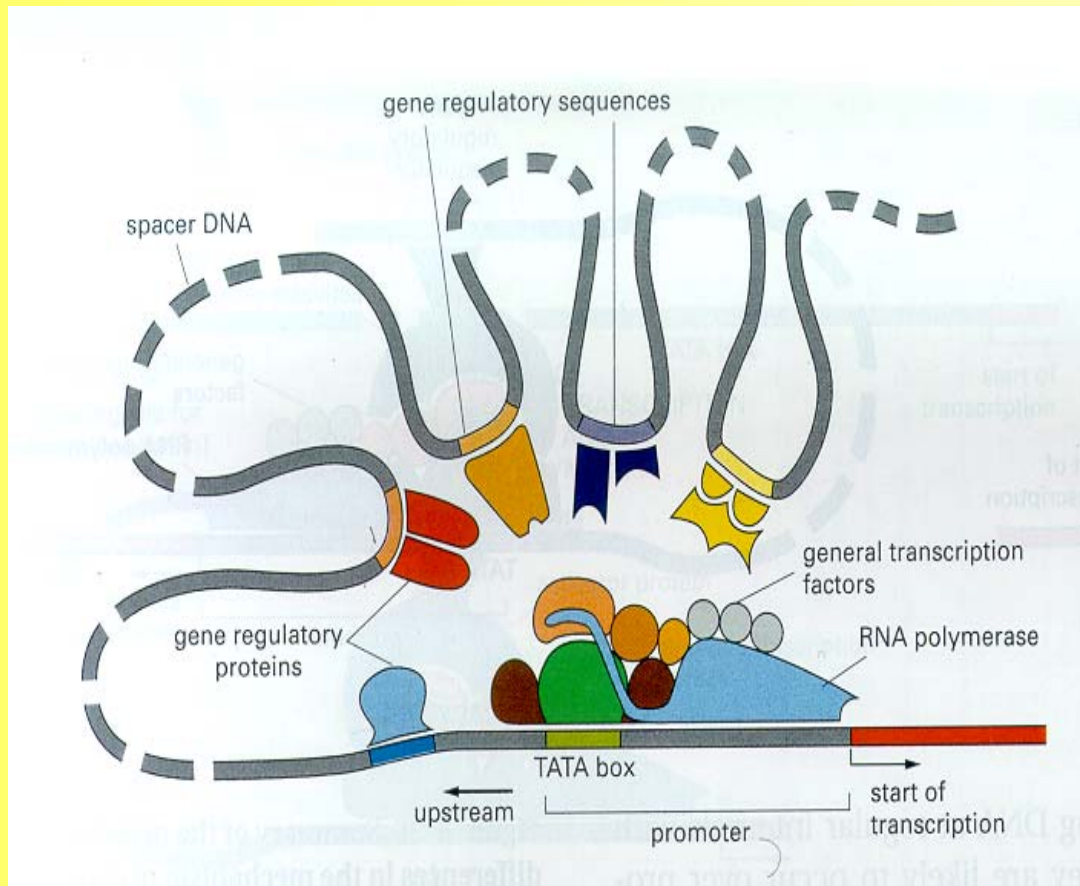
- Lots of DNA function unknown
- Centromeres
- Telomeres
- Regulators
 - Promoters, enhancers
 - Suppressors

**During transcription, introns
are removed (splicing)**



Regulatory elements

- Many intermediate signals co-determine gene activity
- Regulatory elements determine when and how much a gene is active

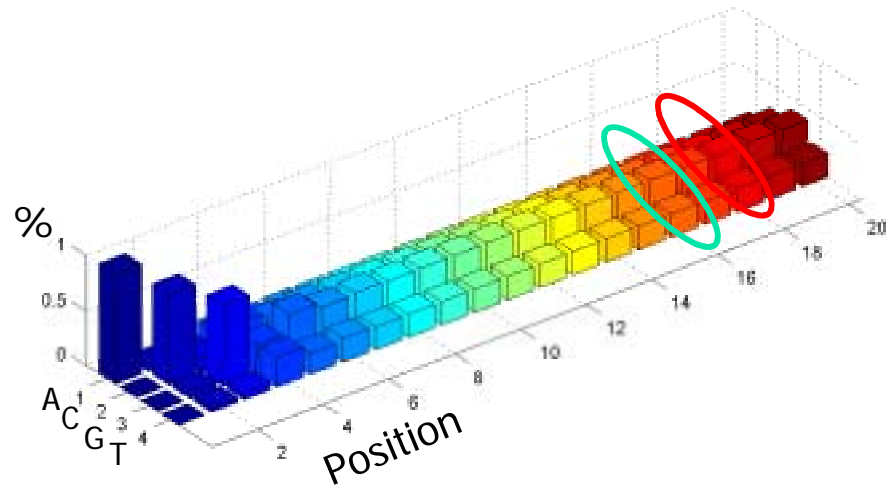


DNA Markov model

	A	C	G	T
A	0.0643	0.8268	0.0659	0.0430
C	0.0598	0.0484	0.8515	0.0403
G	0.1602	0.3407	0.1736	0.3255
T	0.1507	0.1608	0.3654	0.3231

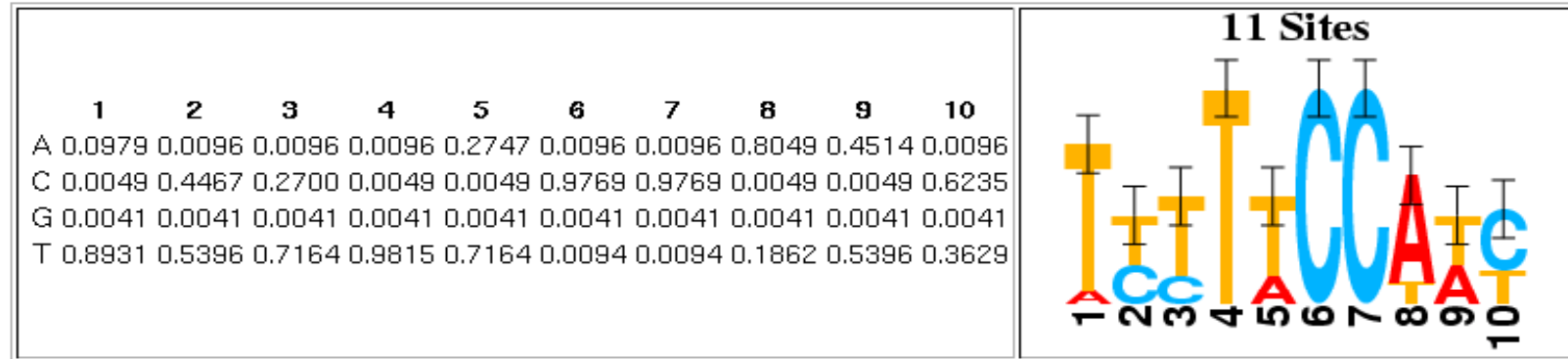
$$\begin{pmatrix} 0.1188 \\ 0.2788 \\ 0.3905 \\ 0.2119 \end{pmatrix}^T \cdot \begin{pmatrix} 0.0643 & 0.8268 & 0.0659 & 0.0430 \\ 0.0598 & 0.0484 & 0.8515 & 0.0403 \\ 0.1602 & 0.3407 & 0.1736 & 0.3255 \\ 0.1507 & 0.1608 & 0.3654 & 0.3231 \end{pmatrix} = \begin{pmatrix} 0.1188 \\ 0.2788 \\ 0.3905 \\ 0.2119 \end{pmatrix}$$

ACGCGGTGTGCGTTTGACGA
 ACGGTTACGCGACGTTTGGT
 ACGTGCGGTGTACGTTGTACG
 ACGGAGTTTGCGGCACCGGT
 ACGCGCGTGACGTACGCGTG
 AGACGCGTGCGCGCGGACGC
 ACGGGCGTGCGCGCGTGGCG
 AACGCGTTTGTGTTTCGCTGC
 ACCGCGTTTGTGACGTTCGCTTC
 ACGTGACGCGTAGTTTCGACG
 ACGTGACACGGACGTACGCG
 ACCGTACTCGCGTTGACACG
 ATACGGCGCGGCGGGCCCGG
 ACGTACGCGTACACGCGGGA
 ACGCGCGTGTTTACGACGTG
 ACGTTCGCACGCGTTCGCTGTG
 ACGGCGGTTCGGTACACGTCG
 ACGTTGCGACGTTCGCTTCG
 ACGGAACGACGACCGGACGC
 ACGGCGTGTTTCGCGTTCGG



Statistical model of a motif

Motif 2: TyTTTCCA_wC



Scores

[Information Content](#) 1.1892
[LogLikelihood](#) 65.1569
[Consensus Score](#) 1.3043

Alignment:

Name	Position	Site	Prob.
Seq 1	268	TTTTTCCAAT	0.8517
Seq 2	283	TTTTACCAAC	0.9887
	635	TTCTTCCAAT	0.7876
Seq 4	512	TTTTTCCATT	0.9053
Seq 5	442	TTTTTCCAAC	0.9980
	200	TCTTTCCTTC	0.9183
Seq 6	349	TCTTACCATC	0.9994
	513	TCCTACCATC	0.9465
Seq 7	286	TCTTTCCTTC	0.9509
Seq 8	346	ATTTTCCATT	0.5764
Seq 9	229	TCCTTCCAAC	0.9806

How to find motifs ?

W.r.t. DNA background,
look for 'overrepresented' patterns

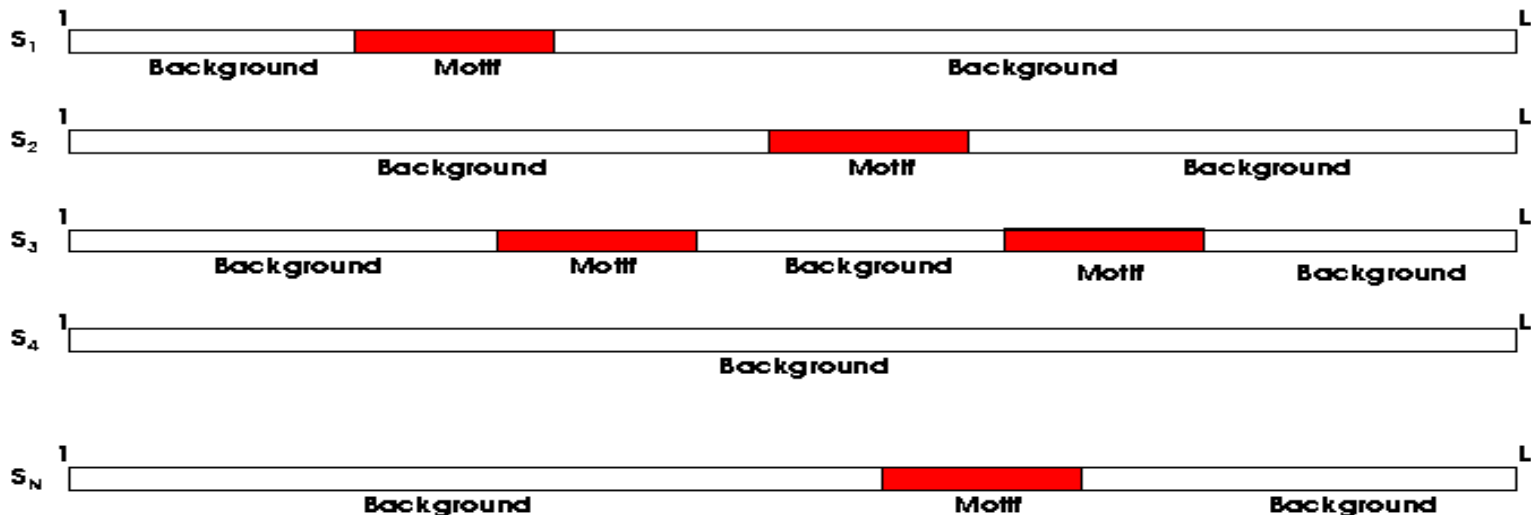
-by analysing 'similarity' in DNA
conserved regions between species;

-'upstream' of co-expressed genes
in one species;

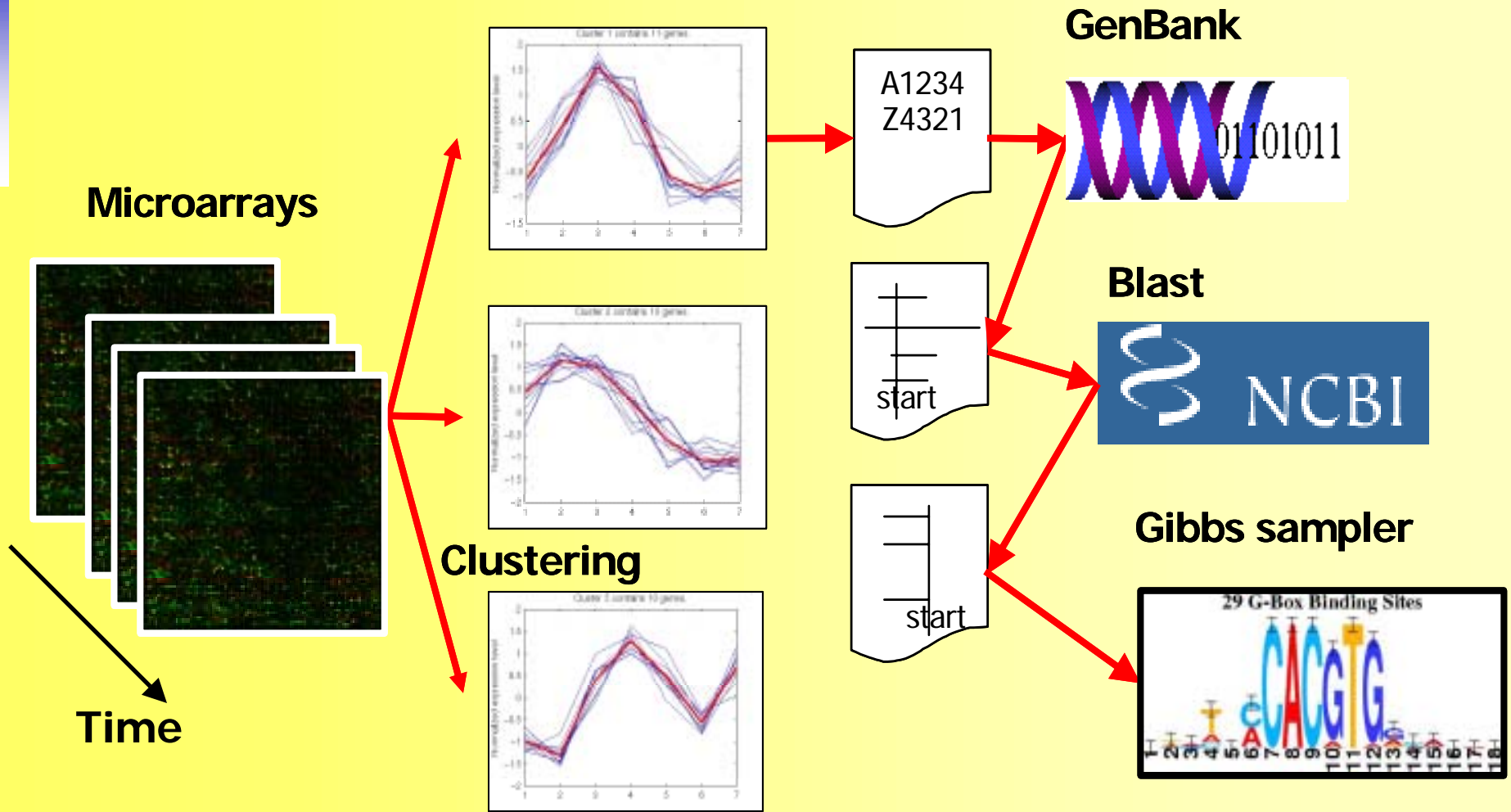


Identifying regulatory sequences

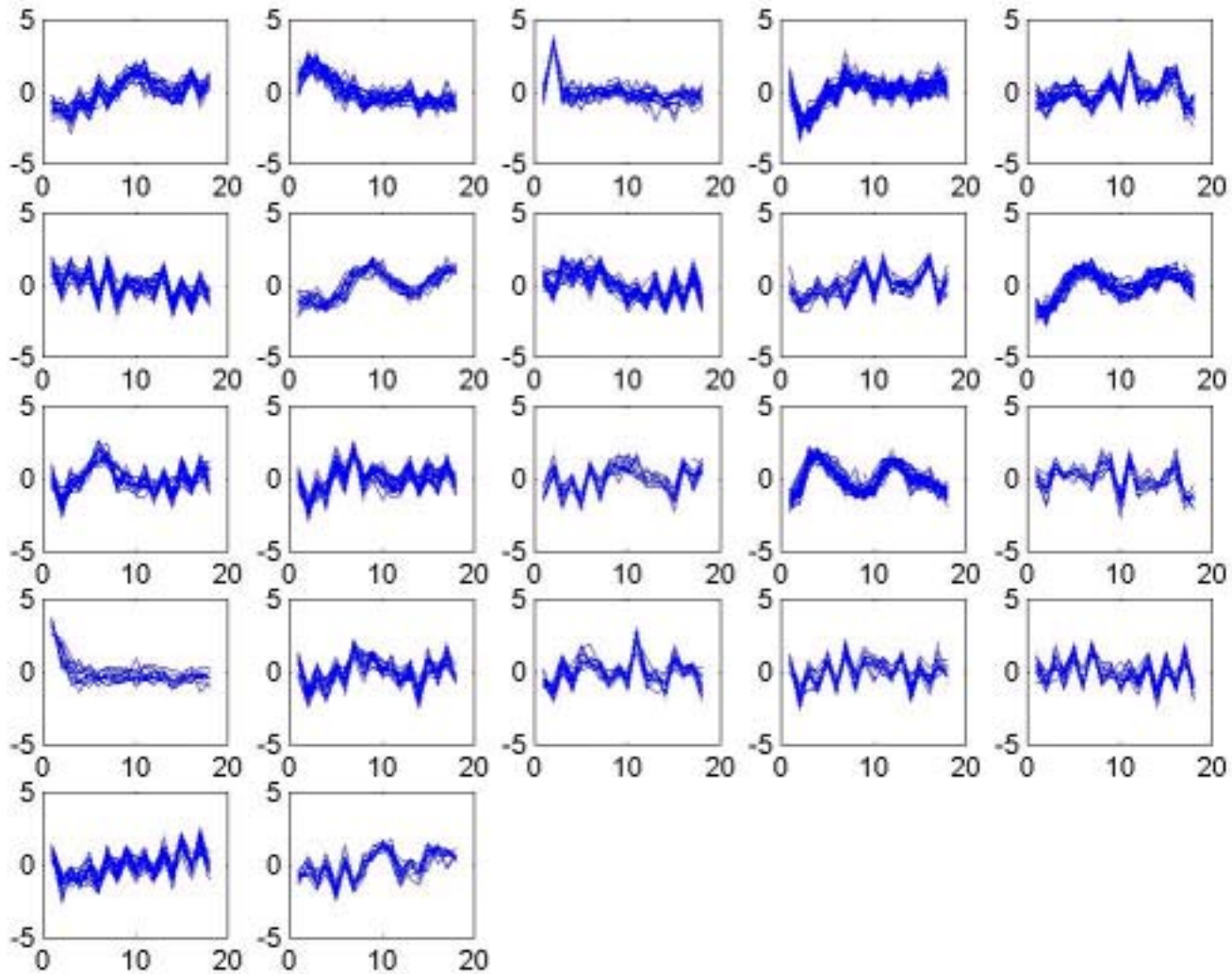
- Cluster genes from microarray expression data to build clusters of coexpressed genes
- Coexpressed genes may share regulatory mechanisms
- Most regulatory sequences are found in the upstream region of the genes (up to 2kb in *A. thaliana*)
- Motifs that are statistically overrepresented in the upstream regions are candidate regulatory sequences



Clustering then motif finding



Clusters: 'Guilt by association'



Zooming in on one cluster

Results: BioDemo_270369

Input Parameters:

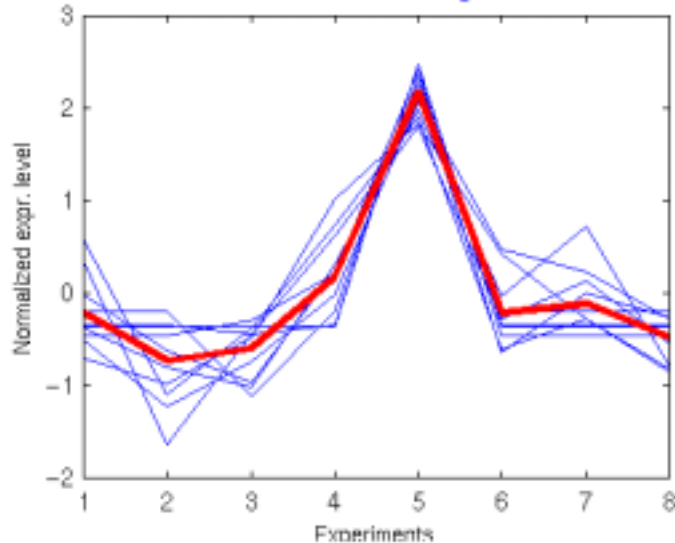
- Required probability of genes belonging to cluster: 0.95
- Minimal number of genes in cluster: 3
- Total number of genes in data set: 138

Results:

- Number of clusters: 12
- Number of genes not in a cluster: 68

Cluster 1

Cluster 1 contains 11 genes.



Accession Number	Gene Name
X98793	AWI31
L40031	CCOAMT
X67816	ELI3
X91957	GER2
AJ000470	GPX2
J04537	HMG1
AC001645	JIP
AF001168	LECRK
n.a.	MEKKK
U08958	NIT2
M34107	PR3AIII

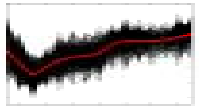
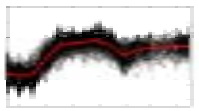



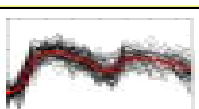
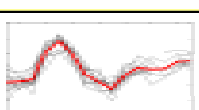
Fetch Sequences

Similarity measure
-Euclidean distance
-Euclidean angle

Relevancy of measure?
- Biologically ?
- Dynamics (e.g. distance between time responses)?



Results

Cluster number	Graphical representation of cluster	Number of ORFs	MIPS functional category (top-level)	ORFs within functional category	P-value ($-\log_{10}$)
1		426	energy transport facilitation	47 40	10 5
3		196	cell growth, cell division and DNA synthesis	48	5
4		149	protein synthesis cellular organisation	71 107	50 19
5		159	cell rescue, defense, cell death and ageing	20	4
6		171	cell growth, cell division and DNA synthesis	76	24
9		78	cell growth, cell division and DNA synthesis	23	4
37		11	metabolism	9	6

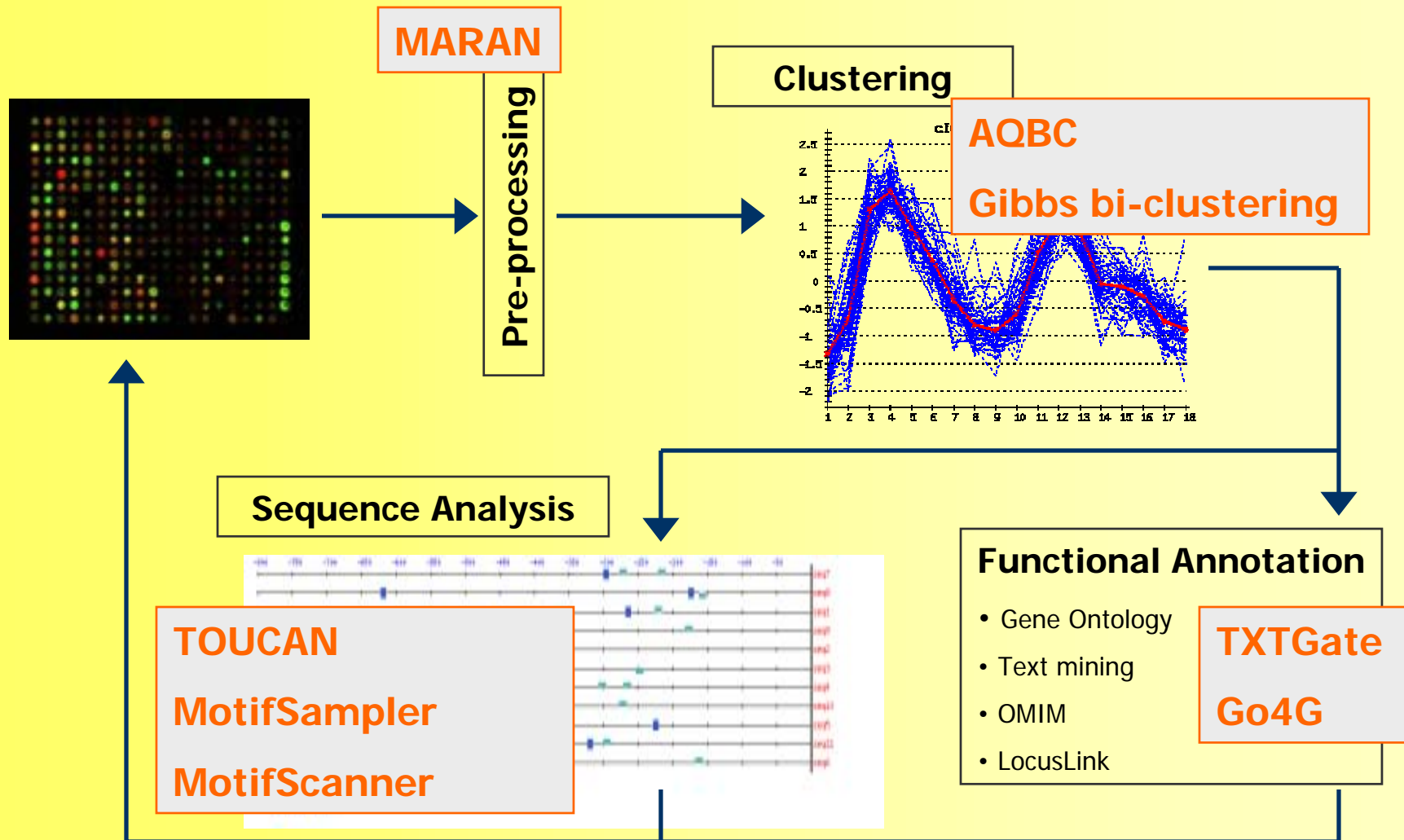


Arabidopsis Thaliana

Cluster	Consensus motif	Runs	PlantCARE	Description
1 [11 seq.]	TAArTAAGTCAC ATTCAAATTT CTTCTTCGATCT	7/10 8/10 5/10	TGAGTCA CGTCA ATACAAAT TTCGACC	Tissue specific GCN4-motif MeJA-responsive element element assoc. to GCN4-motif elicitor responsive element
2 [6 seq.]	TTGACyCGy mACGTCACCT	5/10 7/10	TGACG (T)TGAC(C) CGTCA ACGT	MeJa responsive element elicitor responsive element MeJA responsive element Abcissic acid response element
3 [5 seq.]	wATATATATmTT TCTwCnTC ATAAATAkGCnT	5/10 9/10 7/10	TATATA TCTCCCT -	TATA-box like element TCCC-motif,light response elem. -
4 [5 seq.]	yTGACCGTCCsA CACGTGG GCCTymTT AGAATCAAT	9/10 5/10 8/10 6/10	CCGTCC CCGTCC TGACG CGTCA CACGTG ACGT - -	meristem specific activation of H4 gene A-box, light or elicitor responsive element MeJA responsive element MeJA responsive element G-box light responsive element Abcissic acid response element - -



INCLUSive: online analysis of μ -array data



<http://www.esat.kuleuven.ac.be/inclusive/>



INCLUSive – web portal

INCLUSive Matrix

Maran service router: <http://www.esat.kuleuven.ac.be/maran/Maran>

General INCLUSive service router: <http://www.esat.kuleuven.ac.be/inclusive/ServiceRouter>

Web Application	Web Service	Stand Alone Executable	Info
Link	WSDL	Download	General
Preprocessing:			
MARAN	Maran.wsdl	MaranMain MaranClient	manual [6]
Clustering:			
AQBC	AQBC.wsdl	AQBCMain AQBCClient	[5]
Functional Scoring:			
GO-90			
Sequence Retrieval:			
Intergenic Select			
Sequence Comparison:			
Avid/Viola	Viola.wsdl	download	[4] [8]
Motif Detection:			
MotifSampler	MotifSampler.wsdl	download	help [7] [9] [10]
MotifScanner	MotifScanner.wsdl	download	help [1] [11]
MotifLocator	MotifLocator.wsdl		
FastPrinter	FastPrinter.wsdl	FastPrinterMain FastPrinterClient	manual [2] [3]

References

[1] Aerts, S., Thijs, G., Coessens, B., Staes, M., Moreau, Y., De Moor, B. (2003) TOUCAN: Deciphering the Co-Regulatory Logic of Coregulated Genes. *Nucleic Acids Res.*, **31**: 1753-1764

[2] Blanchette, M., Schwikowski, B., and Tompa, M. (2002) Algorithms for Phylogenetic Footprinting. *Journal of Computational Biology*, **9**: 211-223

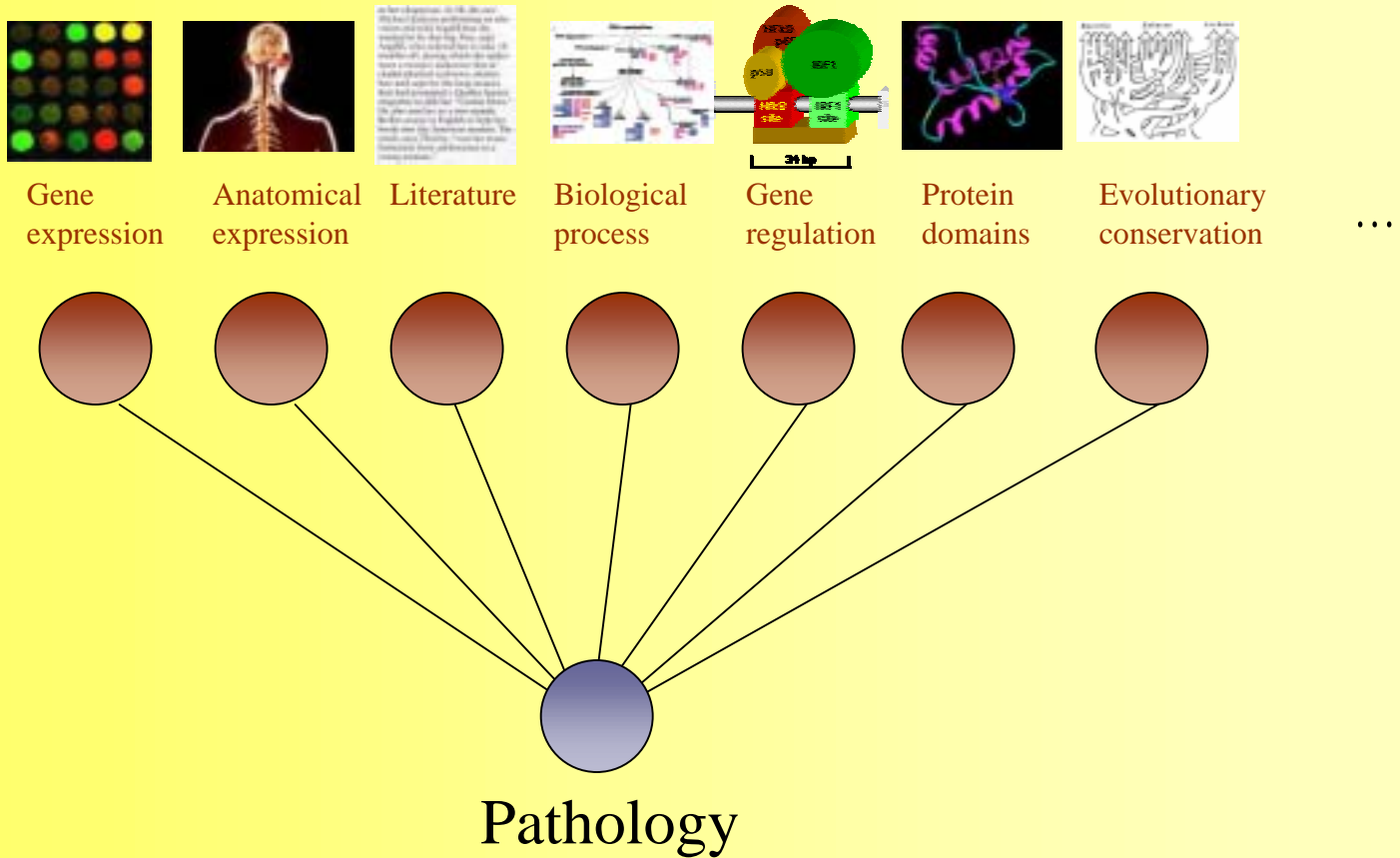
[3] Blanchette, M. and Tompa, M. (2002) Discovery of Regulatory Elements by a Computational Method for Phylogenetic Footprinting. *Genome Research*, **12**: 739-748

[4] Bray, N., Dabchok, I. and Pachter, L. (2003) AVID: A Global Alignment Program. *Genome Research*, **13**: 97

[5] De Smet, F., Mathys, J., Marchal, K., Thijs, G., De Moor, B., Moreau, Y. (2002) Adaptive quality-based clustering of gene expression profiles. *Bioinformatics*, **18**: 735-746

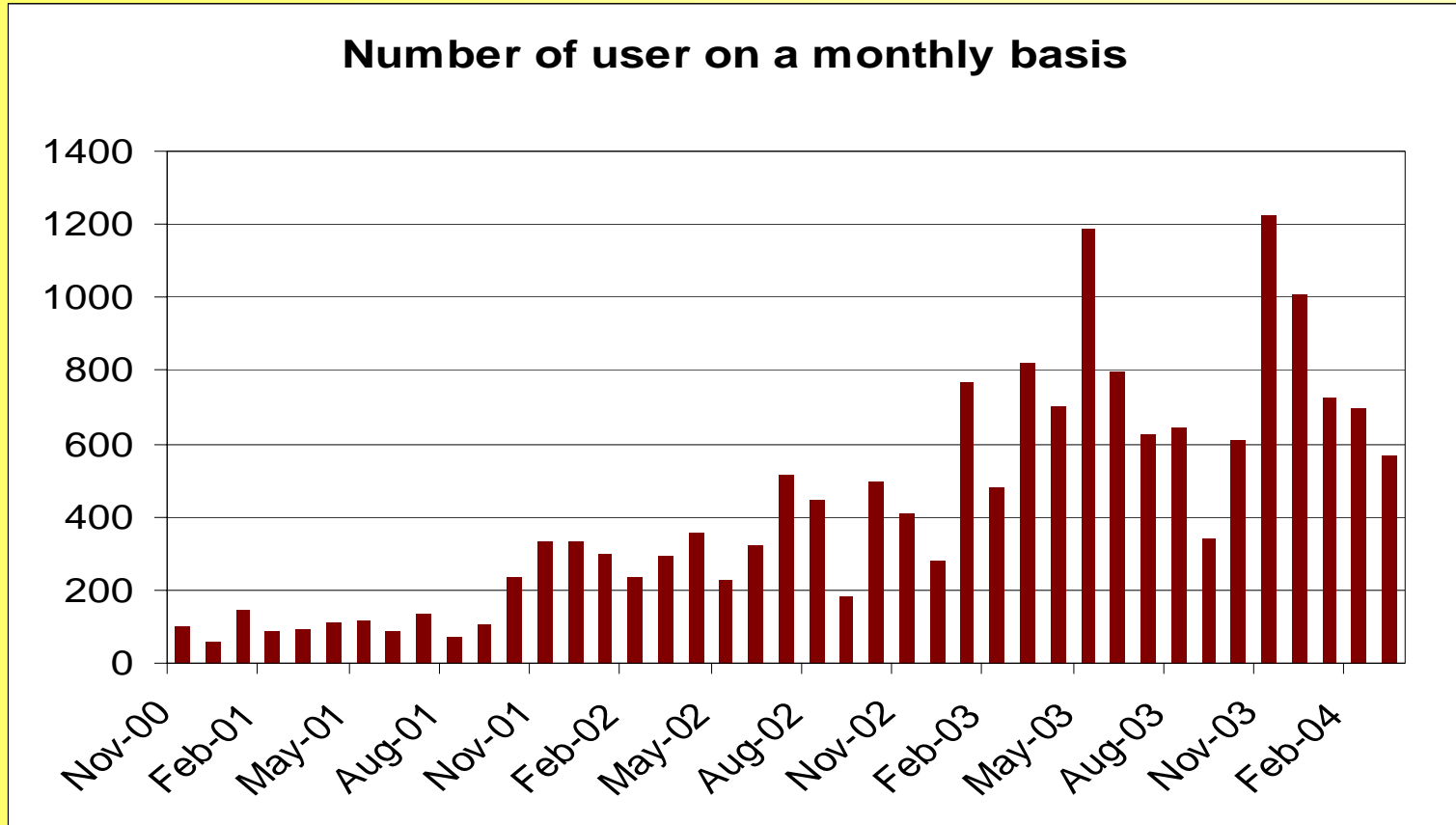


Endeavour: data & algorithm integration



Software statistics: example

Motif sampler



<http://www.esat.kuleuven.ac.be/~dna/BioI/Software.html>



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From Kepler to Newton

Kepler's laws:

Law 1: Orbit is ellipsis with Sun in focus

Law 2: Joing line sweeps out equal areas in equal time

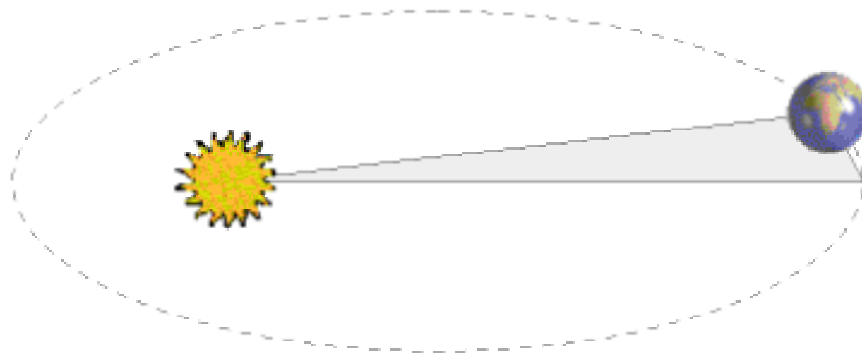
Law 3:
$$\frac{T_1^2}{T_2^2} = \frac{a_1^3}{a_2^3}$$



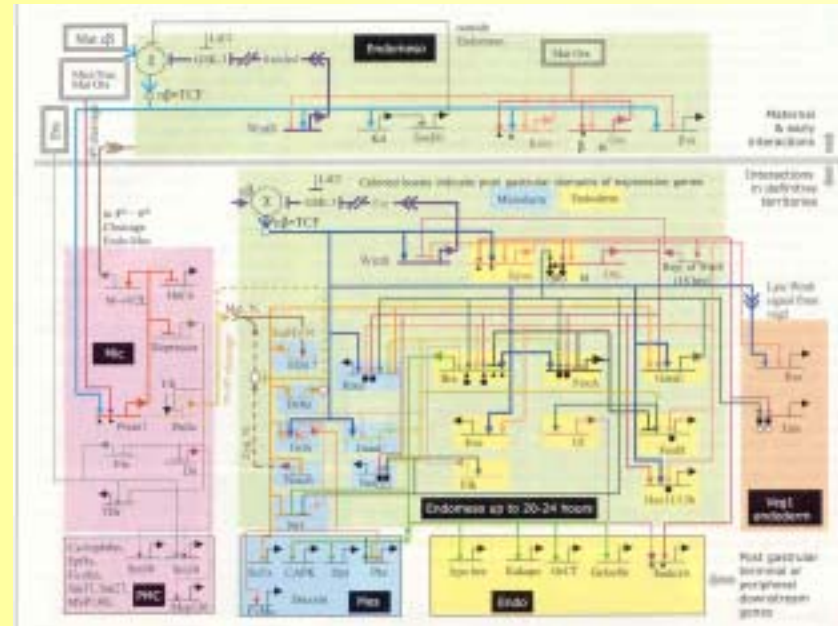
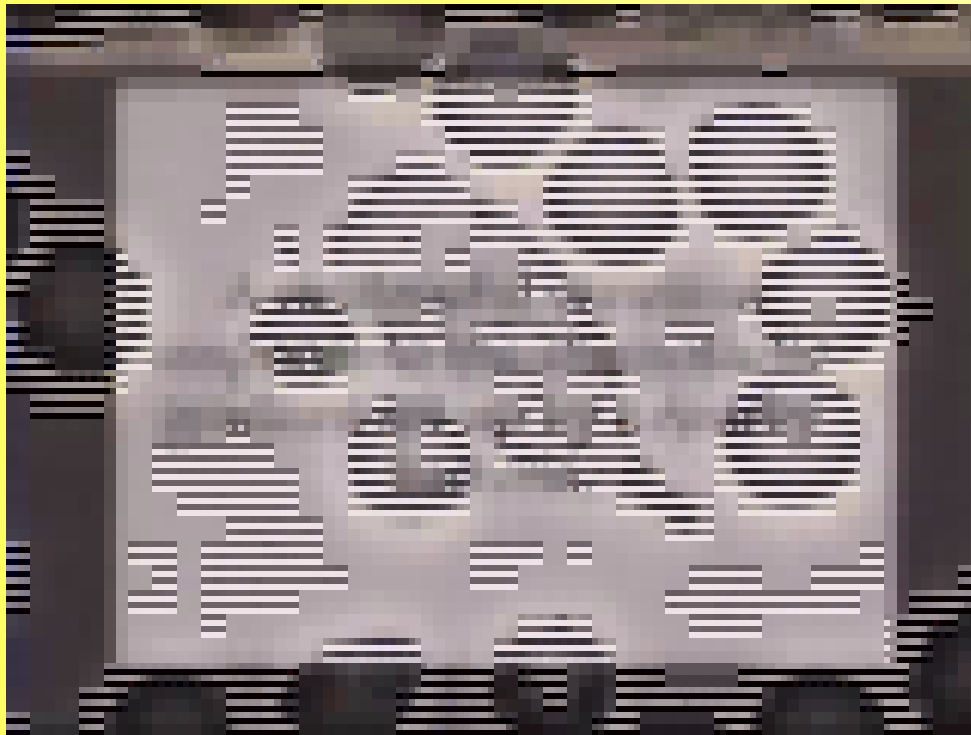
From conic sections to centripetal forces and states

$$F = m \cdot a$$

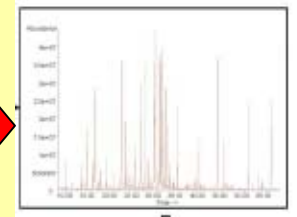
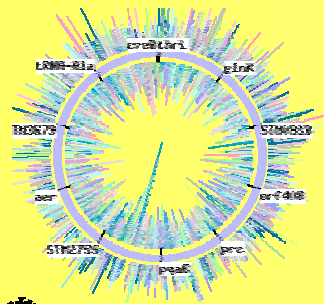
$$F = G \frac{m \cdot M}{r^2}$$



Example: Systems biology: Chemotaxis



'high throughput 'data



genome

transcriptome

proteome

metabolome

interactome



Frankenstein or the modern Prometheus ?

Venter Cooks Up a Synthetic Genome in Record Time

Elizabeth Pennisi , *Science*

When the U.S. Department of Energy (DOE) announced last week that sequencing maverick J. Craig Venter had taken just 2 weeks to build a viral genome from scratch, Secretary of Energy Spencer Abraham called the work "nothing short of amazing."

He predicted that it could lead to the creation of microbes tailored to deal with pollution or excess carbon dioxide or even to meet future fuel needs. But the \$3 million DOE project drew ho-hum reviews from some scientists. "I didn't think it was a big deal," says Ian Molineux, a molecular biologist at the University of Texas, Austin. And Richard Ebright, a molecular biologist at Rutgers University in Piscataway, New Jersey, agrees: "This is strictly a limited incremental advance over current technologies."

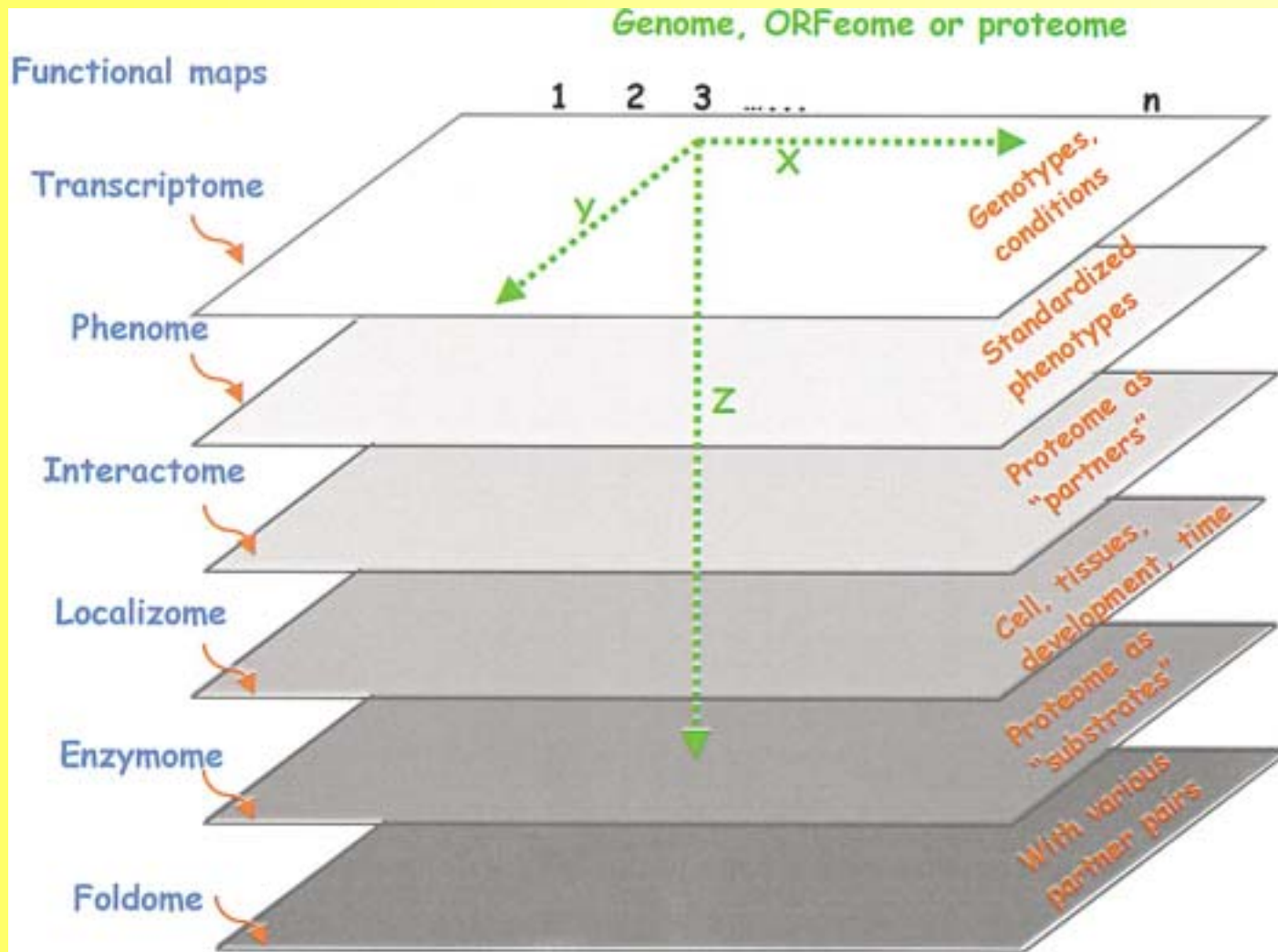
The skeptics focus on how hard it will be to go beyond the initial step, while Venter, head of the Institute for Biological Energy Alternatives (IBEA) in Rockville, Maryland, and former president of Celera Genomics, and his backers are proud to have gotten this far. All are in agreement, however, that the experiment demonstrated speed in converting raw ingredients into a functioning virus.

The genome synthesized by the Venter-led group belongs to a bacterial virus, called a phage; when it was tested in a lifelike situation, Venter reported, it infected and killed bacteria just as natural phages would.

Because his team stitched together the phage's DNA in just a few weeks instead of years, molecular virologist Eckard Wimmer of the State University of New York, Stony Brook, called the effort "a very smart piece of work."



Omic's world



© Vidal M. Cell. 2001 Feb 9;104(3):333-9.



Systems biology / Whole-istic / Integration



SYSTEMS BIOLOGY
 INTRODUCTION

Whole-istic Biology

If someone were to analyze current reviews and fashionable catchwords, he would find "systems" high on the list. The concept has pervaded all fields of biology, physics, and even music. In keeping with this trend, I will begin with emphasis on systems biology. And while giving praise, for the words quoted above were the culture in Ludwig von Bertalanffy's 1967 article, I'll also mention a few caveats, some of which date to Bertalanffy's original general systems theory not only in general science as well. In his view, old-fashioned sciences by reducing them to an interplay of elementary entities. Contemporary sciences, on the other hand, recognize that as "problems of organization, phenomena not reducible to the behavior of parts when in short, 'systems' of various orders not understandable by isolation." And this remains an effective definition of systems integration and application of mathematics, engineering, and a range of complex biological regulatory systems. The delay between the early pre-eminence of the theory may be primarily to accommodate sufficient dimensions of the whole. Kitzler's Review (p. 1662) in this field and identifies areas where attempts will flourish by the absence of required a "Chain and Dangle" (p. 1664) bring an biological complexity. Biologists can be better design control systems (such as a virus, every minute, a solution of safety genome). They provide insights into systems for the molecular plants (most that studying the function of cardiac cells and the one just beginning to solve the "sugar's gas" coming from cells in gene networks. Davidson et al. (p. 1668) demonstrate regulatory network can explain the early development of an embryo. These regulatory circuits prescribe the order of genes that determine the fate of developing cells and together form a one-way path to build a functional organism would be satisfied, as he notes in his book that this rational development of the new systems analysis, even from a holistic view (or fluid), was the very observation that "led the US through to antibiotic resistance, i.e., the discovery that viral proteins applicable in terms of natural science." Bertalanffy thought argued that in a system open to its environment, the state not determines the final steady state, and that "the system physical laws" proposed by Dirac's "stappans".

Also, at Science's Signal Transduction Knowledge (STKPL) Lab describes design similarities between electrical signaling networks. The refinement of cell system to the immune responses, can also be explained through models, as described by Chikraverty.

—LISA CHEN AND

Looking out Bertalanffy, General Systems Theory Association, Division of Systems Science, New York, 1995.

Science





www.sciencemag.org SCIENCE VOL.



Yeast protein-protein interactions

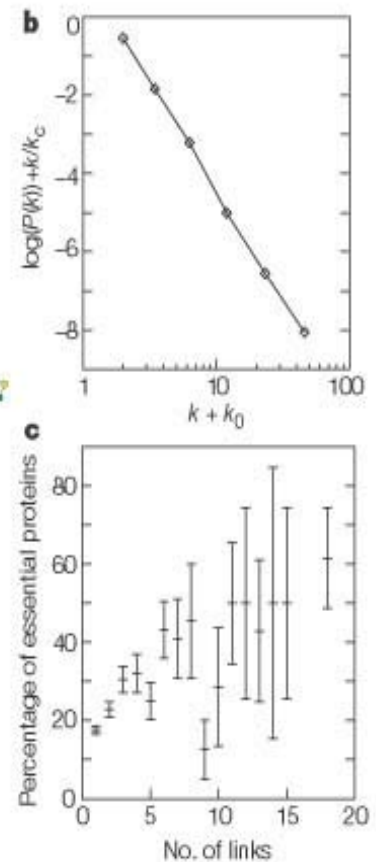
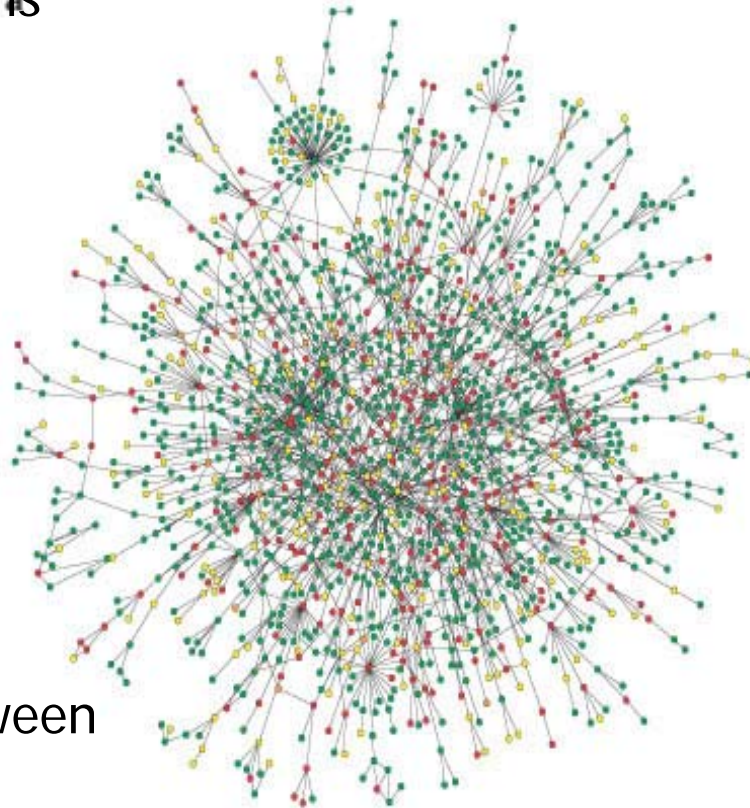
- 78% of proteins shown in giant component

- Protein-protein interactions

-  : lethal mutation
-  : slow growth
-  : non-lethal
-  : unknown

- Connectivity $P(k)$

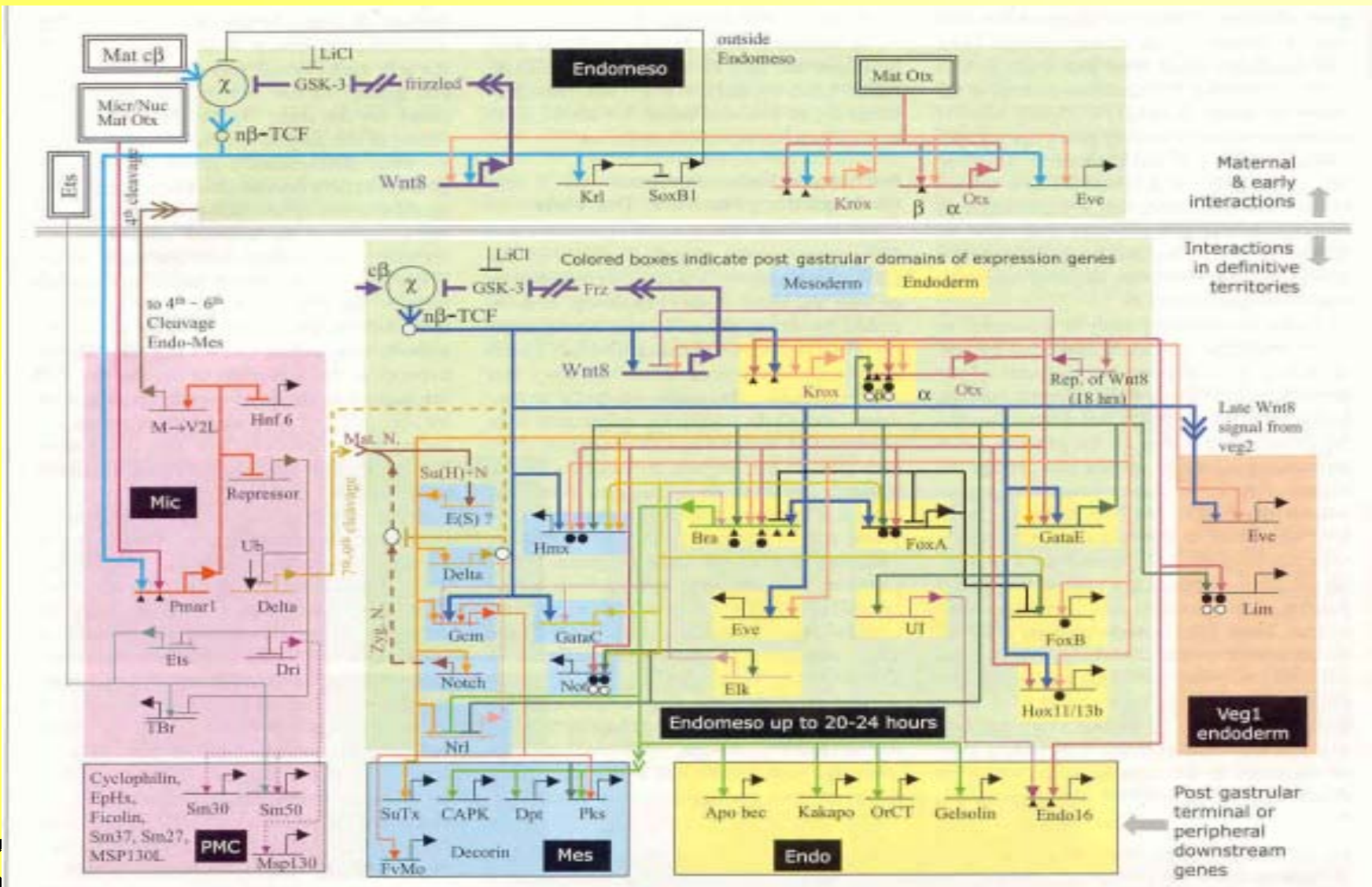
- Fragility: Correlation between connectivity and lethality



© Jeong H et al. Nature. 2001 May 3;411:41-2.



Unravelling genetic networks....



ODE model of cell cycle

Table 1. A mathematical model of the proposed mechanism (Fig. 1) for the fission yeast cell cycle

Differential equations*

$$\frac{d}{dt} Rum1 = k_5 - k'_6 \cdot Rum1 - k_p \cdot (MPF_a + \varepsilon_p \cdot SK \cdot mass) \cdot Rum1 + (k'_{pp} + k_{pp} \cdot PP) \cdot Rum1P - k_j \cdot MPF \cdot Rum1 + k_{JR} \cdot CR + k_{2c} \cdot CR$$

$$\frac{d}{dt} Rum1P = k_p \cdot (MPF_a + \varepsilon_p \cdot SK \cdot mass) \cdot Rum1 - (k'_{pp} + k_{pp} \cdot PP) \cdot Rum1P - (k'_6 + k_6) \cdot Rum1P - k_j \cdot MPF \cdot Rum1P + k_{JR} \cdot CRP + k_{2c} \cdot CRP$$

$$\frac{d}{dt} CR = k_j \cdot MPF \cdot Rum1 - k_{JR} \cdot CR - k_{2c} \cdot CR - k'_6 \cdot CR - k_p \cdot (MPF_a + \varepsilon_p \cdot SK \cdot mass)$$

$$\frac{d}{dt} CRP = k_p \cdot (MPF_a + \varepsilon_p \cdot SK \cdot mass) \cdot CR - (k'_{pp} + k_{pp} \cdot PP) \cdot CRP + k_j \cdot MPF \cdot Rum1P$$

$$\frac{d}{dt} MPF = k_1 \cdot mass - k_2 \cdot MPF - k_{Wee} \cdot MPF + k_{C25} \cdot preMPF - k_j \cdot MPF \cdot (Rum1 + Rum1P)$$

$$\frac{d}{dt} preMPF = k_{Wee} \cdot MPF - k_{C25} \cdot preMPF - k_2 \cdot preMPF$$

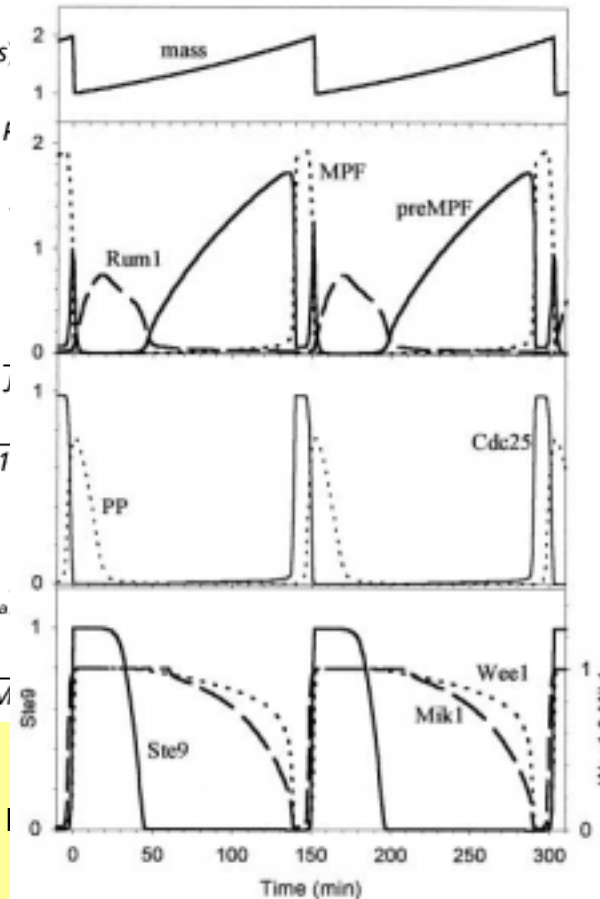
$$\frac{d}{dt} Ste9 = (k'_{Ste9R} + k_{Ste9R} \cdot PP) \cdot \frac{1 - Ste9}{J_{Ste9R} + 1 - Ste9} - k_{Ste9} \cdot (MPF_a + SK \cdot mass) \cdot Ste9$$

$$\frac{d}{dt} Mik1 = (k_5 + k'_{MR} + k_{MR} \cdot PP) \cdot \frac{1 - Mik1}{J_{MikR} + 1 - Mik1} - k_M \cdot MPF_a \cdot \frac{Mik1}{J_{Mik} + Mik1}$$

$$\frac{d}{dt} Wee1 = (k'_{WR} + k_{WR} \cdot PP) \cdot \frac{1 - Wee1}{J_{WeeR} + 1 - Wee1} - k_W \cdot MPF_a \cdot \frac{Wee1}{J_{Wee} + Wee1}$$

$$\frac{d}{dt} Slp1 = k_{as} \cdot MPF_a - k_{ad} \cdot Slp1, \quad \frac{d}{dt} Slp1_a = (k'_{aa} + k_{aa} \cdot MPF_a) \cdot (Slp1 - Slp1_a)$$

$$\frac{d}{dt} PI = k_i \cdot Inh \cdot PP - k_{IR} \cdot PI - k_d \cdot PI, \quad \frac{d}{dt} mass = \mu \cdot mass, \quad \frac{d}{dt} R_{DNA} = \frac{K}{1 + Y \cdot M}$$



$$k_{25R} \cdot PP \cdot \frac{Cdc25}{J_{25R} + Cdc25}$$

$$k_{IR} \cdot PI - k_d \cdot Inh$$

4): 7865–7870

© Sveiczler A et al. |



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Innovation through multidisciplinary

'Enlightment': Split up sciences

Dr. Eric Lander

"For me as a scientist in the world of genomics, watching this amazing convergence of biology, medicine, computer science and technology, is tremendously exciting."

20th Century

21th Century



Nano-Technology

Micro-Electronics

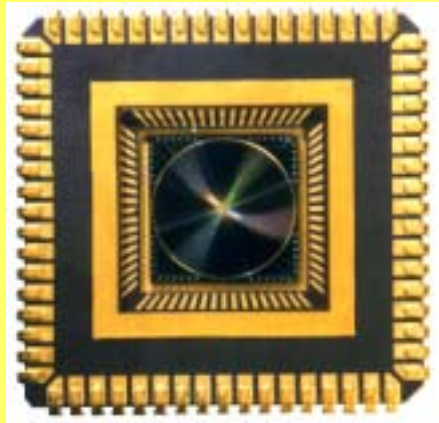
Nano-technology

Biotechnology

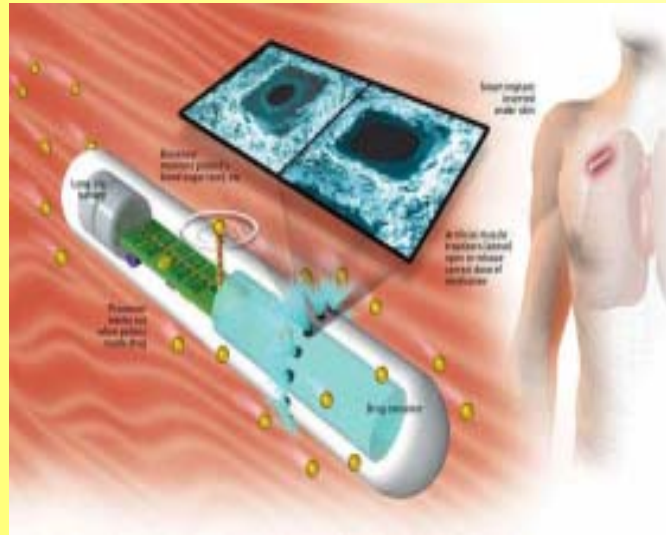
'Renaissance': Merge sciences



Nano-Sensoren en Actuatoren



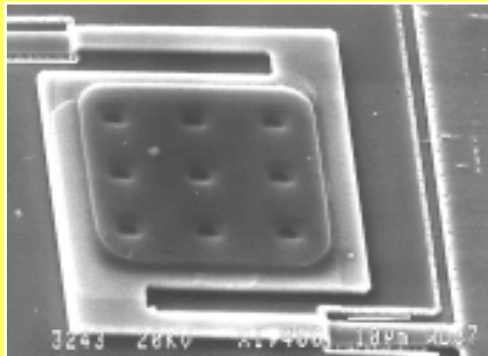
CMOS Imager



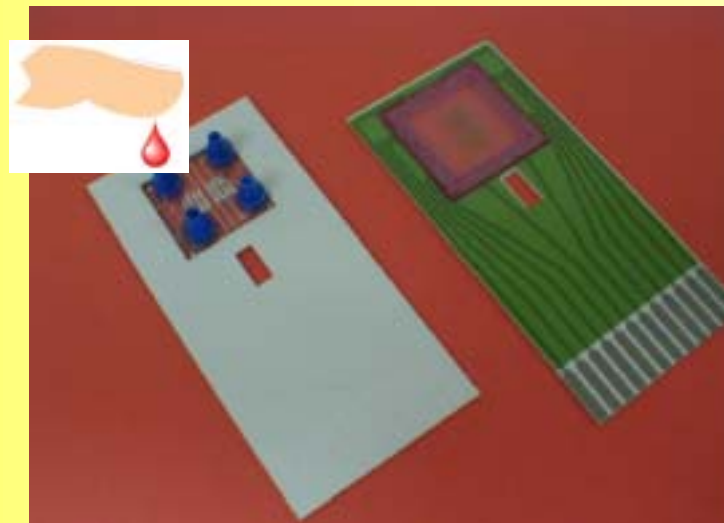
Smart Pill (Ohio State Univ)



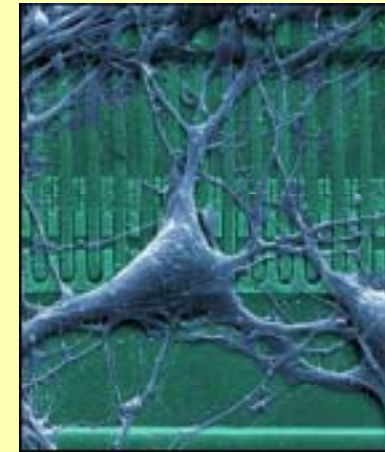
Blood gas sensor (IMEC)



IR Sensor (IMEC)



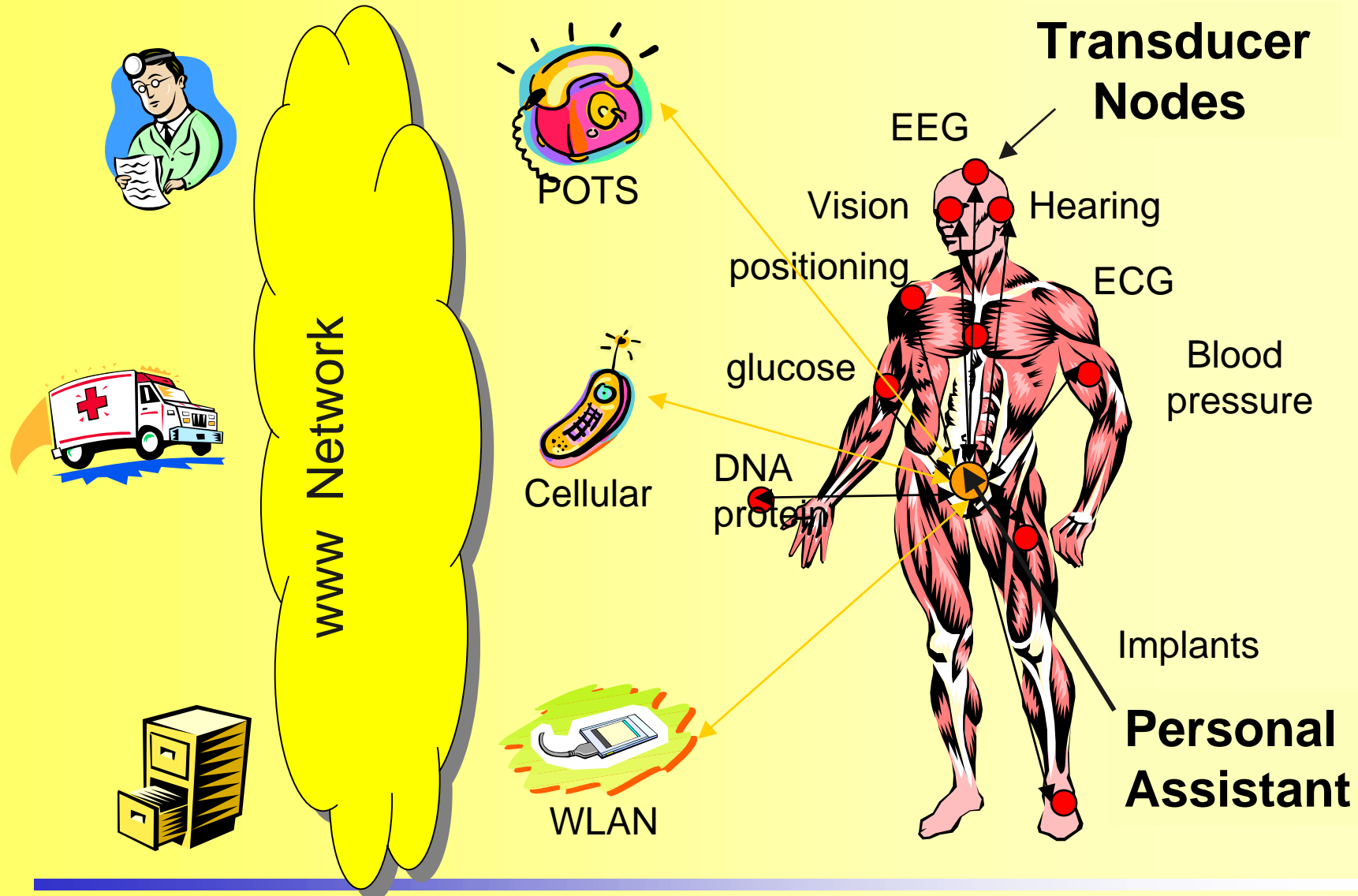
Prostate cancer diagnosis (IMEC)



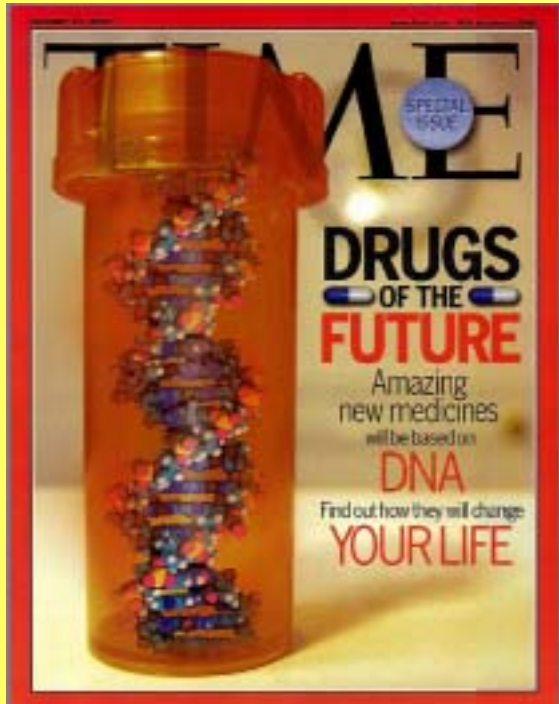
NeuronSensor (KNS)



Human++ programma IMEC



This is the (very near) future...

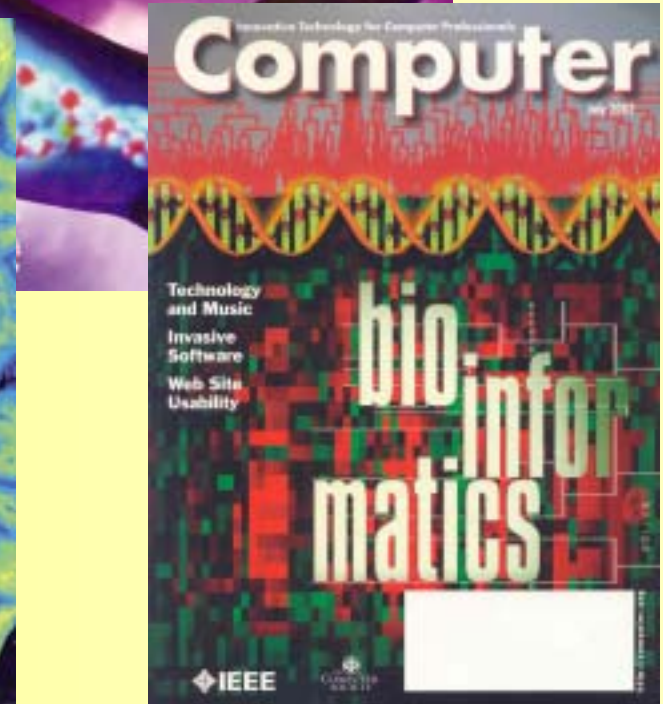
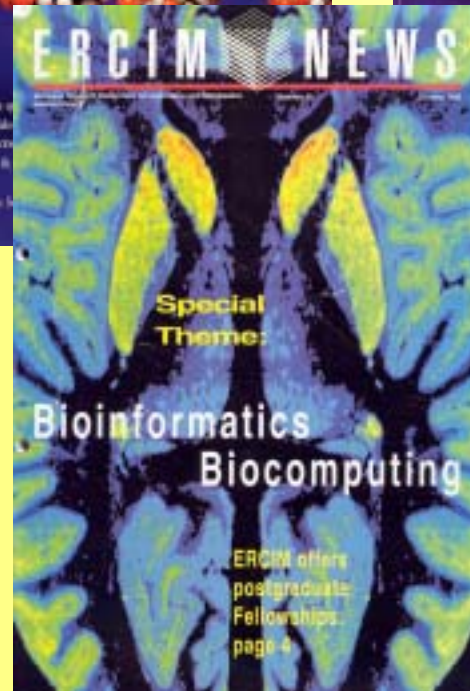
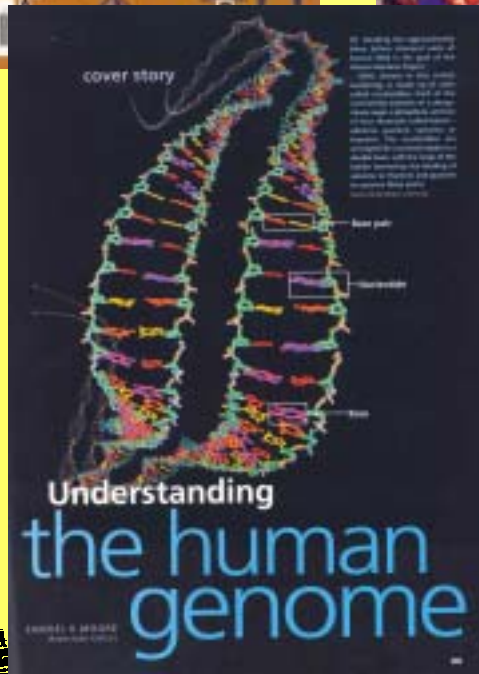
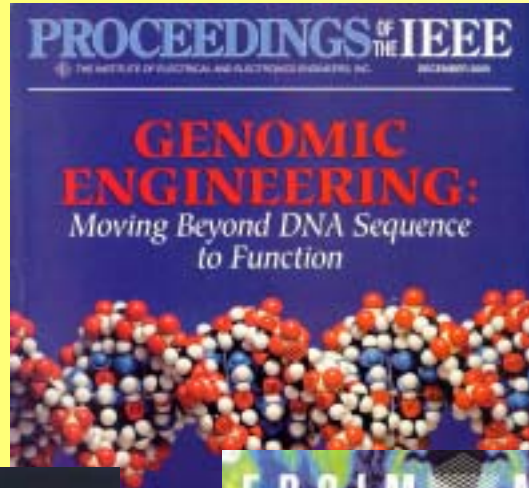


'Customized'
medicine

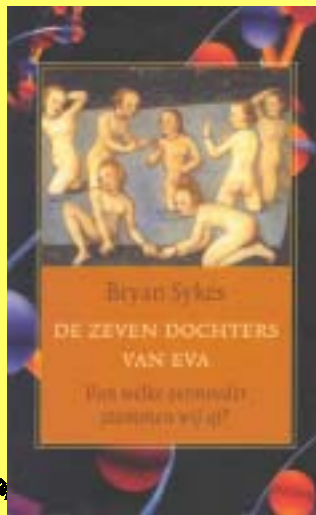
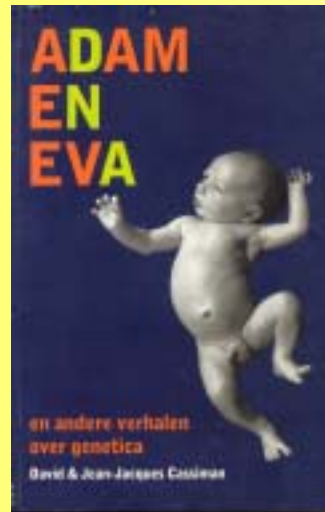
Massive automated genetic
screening of 1000s of assays



What to read and study (the specialist) ?



What to read and study ?



...and finally...

The Human Genome Project has catalyzed striking paradigm changes in biology - *biology is an information science*. [...] Systems biology will play a central role in the 21st century; there is a need for global (high throughput) tools of genomics, proteomics, and cell biology to decipher biological information; and *computer science and applied math* will play a commanding role in converting *biological information into knowledge*.

Leroy Hood, Institute for Systems Biology, Seattle, WA, 2002

