
Bioinformatics and Computerscience

Systems Biology

Data collection

1. ARRAY BASED
2. NEXT-GEN SEQUENCING
RNA-Seq analysis
ChIP-seq
Bulked segregant analysis

Network Inference

1. Sequence-based data analysis
MotifSuite
ModuleDigger
Crossed
2. Network reconstruction
Lemone
Distiller
Comodo
Bayesian network reconstruction

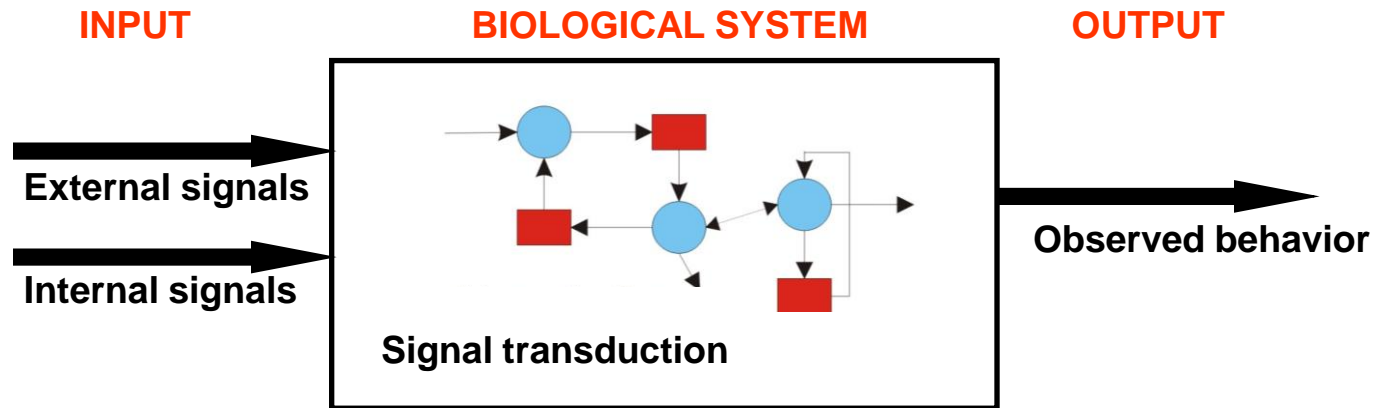
Network-based data integration

1. Network-based analysis of unstructured gene lists
2. Network-based gene prioritization
3. Network-based eQTL analysis
4. Network-based subtyping

- Development of methods that assist systems biologists
- Methods based on data-mining, statistics
- Unique in combining biologically relevant assumptions with rigorous statistical and data mining framework (pragmatic but not too much ad hoc)
- All tools have been validated on real biological cases

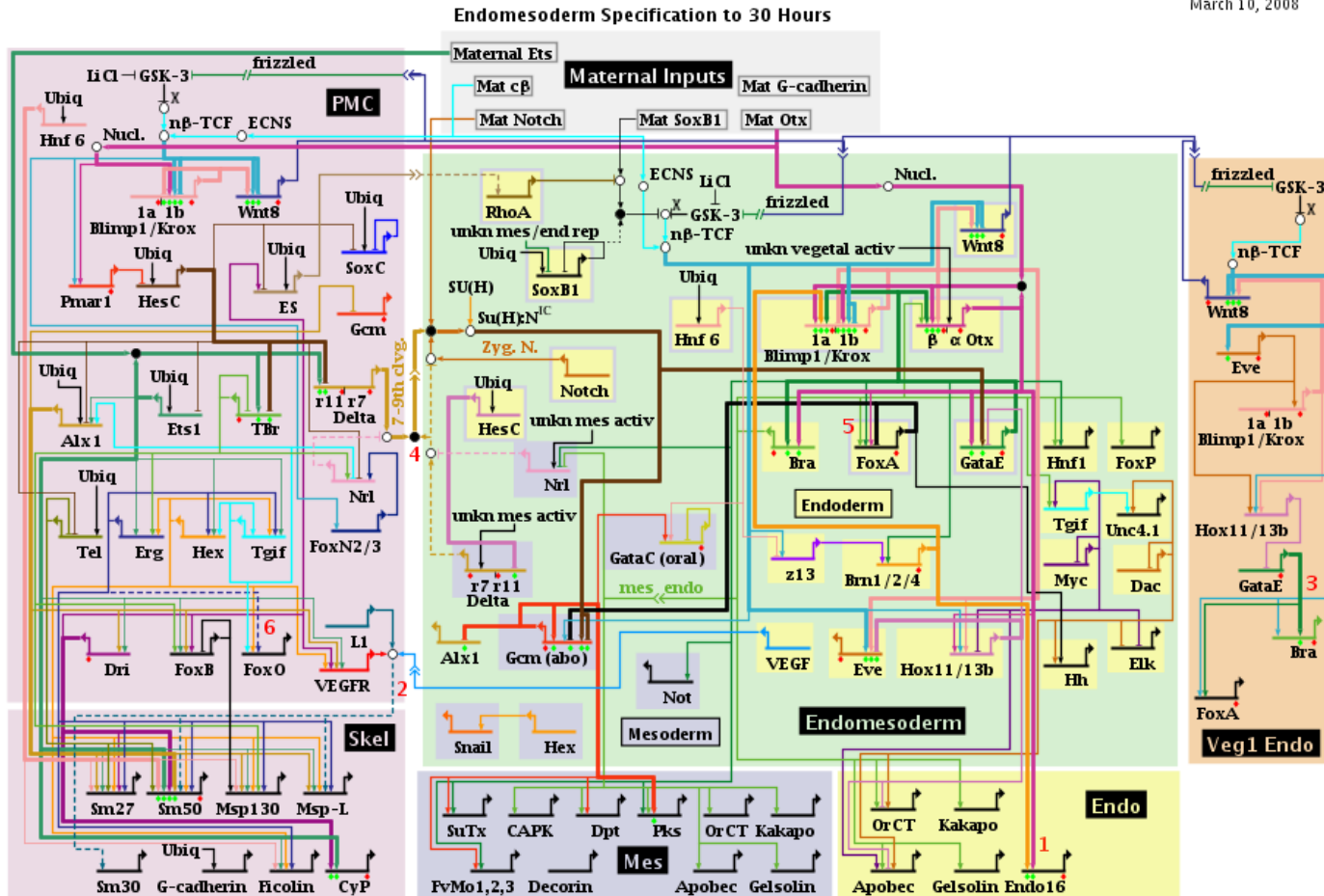
<http://bioinformatics.psb.ugent.be/DBN/dbn/software>

Systems Biology



Systems Biology

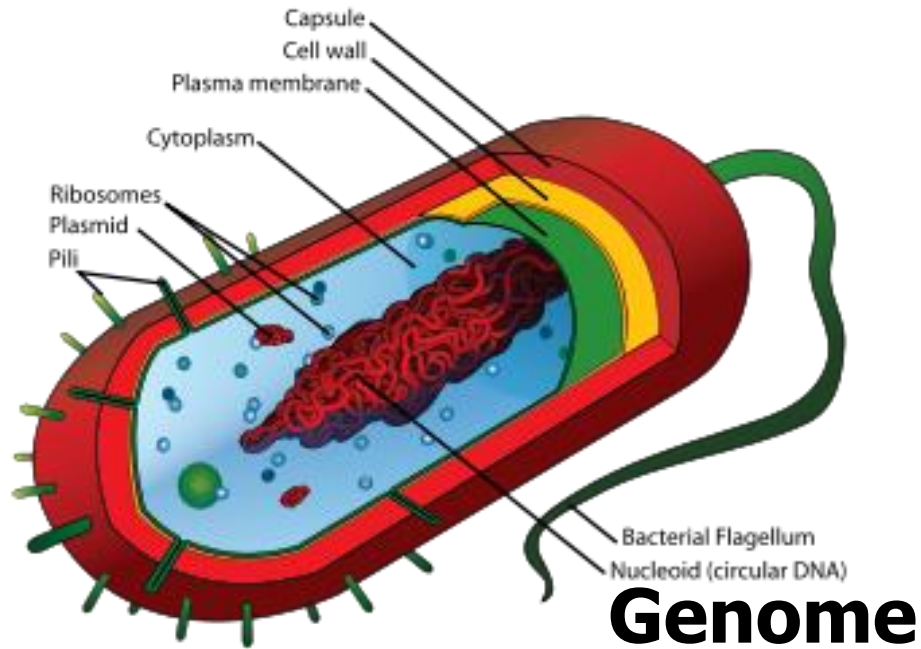
March 10, 2008



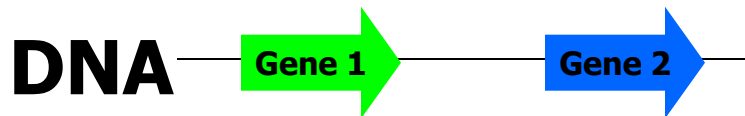
Ubiq=ubiquitous; Mat = maternal; activ = activator; rep = repressor;
 unkn = unknown; Nucl. = nuclearization; χ = β-catenin source;
 nβ-TCF = nuclearized b-β-catenin-Tcf1; ES = early signal;
 ECNS = early cytoplasmic nuclearization system; Zyg. N. = zygotic Notch

Copyright © 2001-2008 Hamid Bolouri and Eric Davidson

Systems Biology

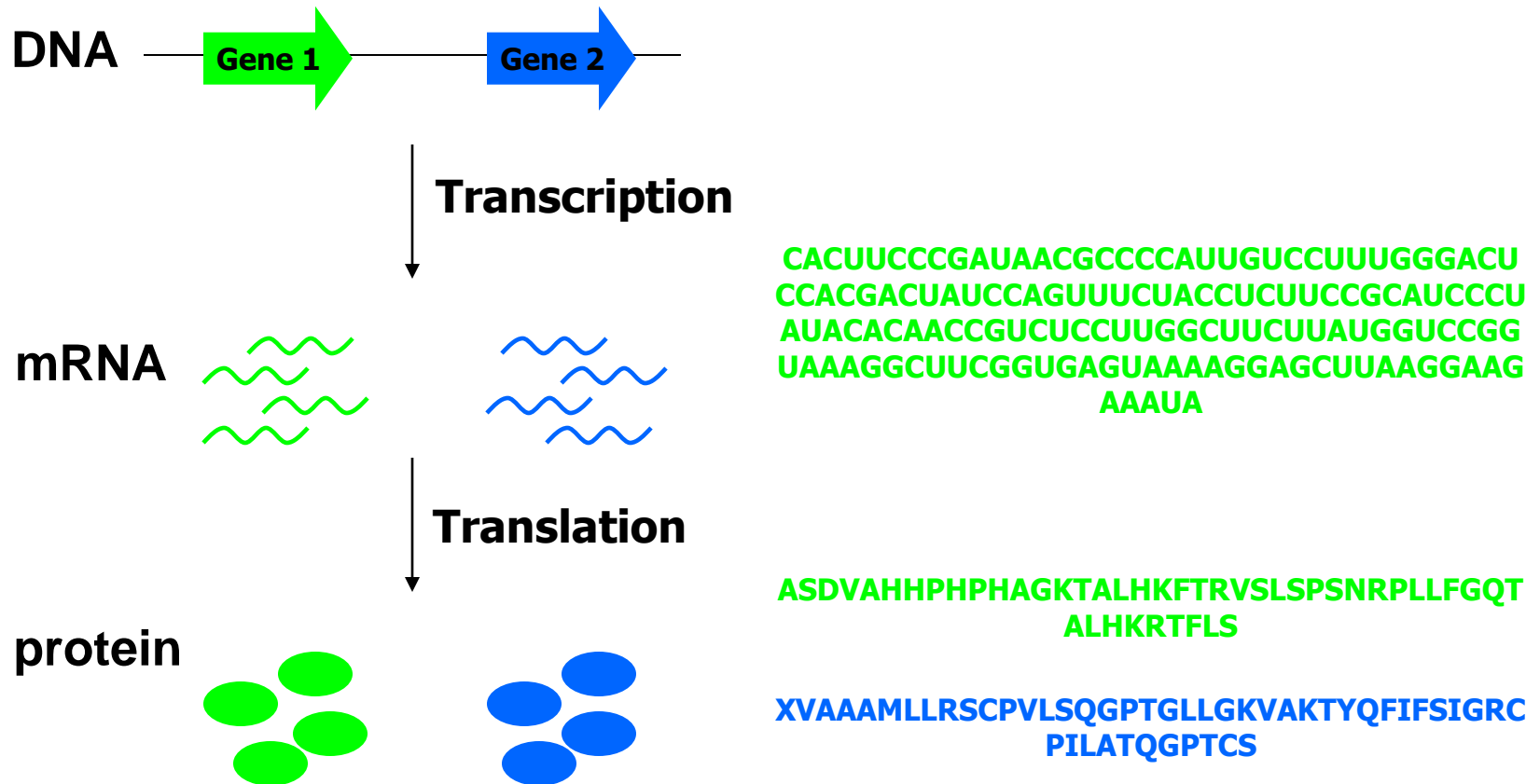


Systems Biology

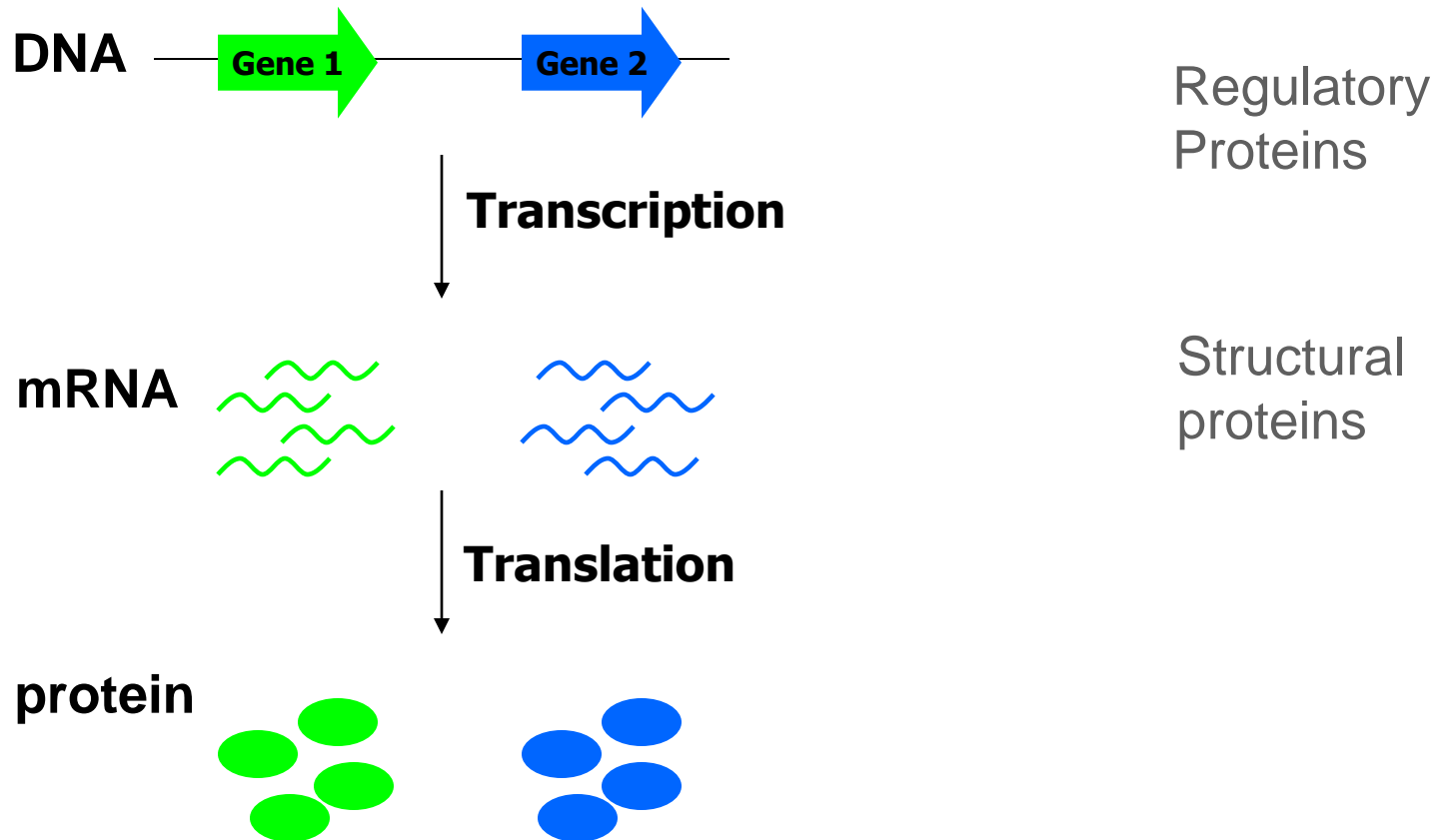


AGCACTGTCCACTGCATGGTGAGGATGGGGGTGAGCTCCCT
TTGTGGCTAGGTGCTTAAACGTCTATCGGACGCTCAG**GTGAA**
GGGCTATTGCGGGGTAACAGGAAACCCTGAGGTGCT
GATAGGTCAAAGATGGAGAAGGCCTAGGGATATGTG
TTGGCAGAGGAACCGAAGAATACCAGGCCATTCCG
AAGCCACTCATTTCCTCGAATTCCTTCTTTATGCCTTC
AGTCTCTATTGACCGTAAATTTGGTTGTTGTCTCCCAGCTGT
TTATTTCTGTAACAGATCTTGGAGGCTGCGGTCTGGATCCCT
CGCCAAGAACCAGATCCAGGAGAAAACGTGCTCAACGTGCA
GCTCTGCTCCTACTGATTATAGCCCCACAGATGACATCGCTC
CATAGTCACACCAAGTCTCCTGTGGGAGTCTTGCTCCTCGTT
CTCAGTGTCTGTTACAGCTCGGTATTTAGTGTGTCAGGACGTC
GGCTCCCAGCCCGCATCTCCGCTCAGCAATGCCATTATCTTC
TCAGCCAAGTCCTAGAAATGGGTTG**GCTTCCCATTGCAA**
AAACATCGTCCATAGTCACACCAAGTCTCCTGTGGG
AGTCTTGCTCCTCGTTCTCAGTGTCTGTTACAGCTCG
GTATTTAGTGTGTCAGGACGTCGGCTCCCAGCCCGCAT
CTCCGCTCAGCAATGCCATTATCTTCTCAGCCAAGTCCT
AGAAATGGGTTGGCTTCCCATTTGCAAAAACATCGCTCCATA
GTCACACCAAGTCTCCTGTGGG

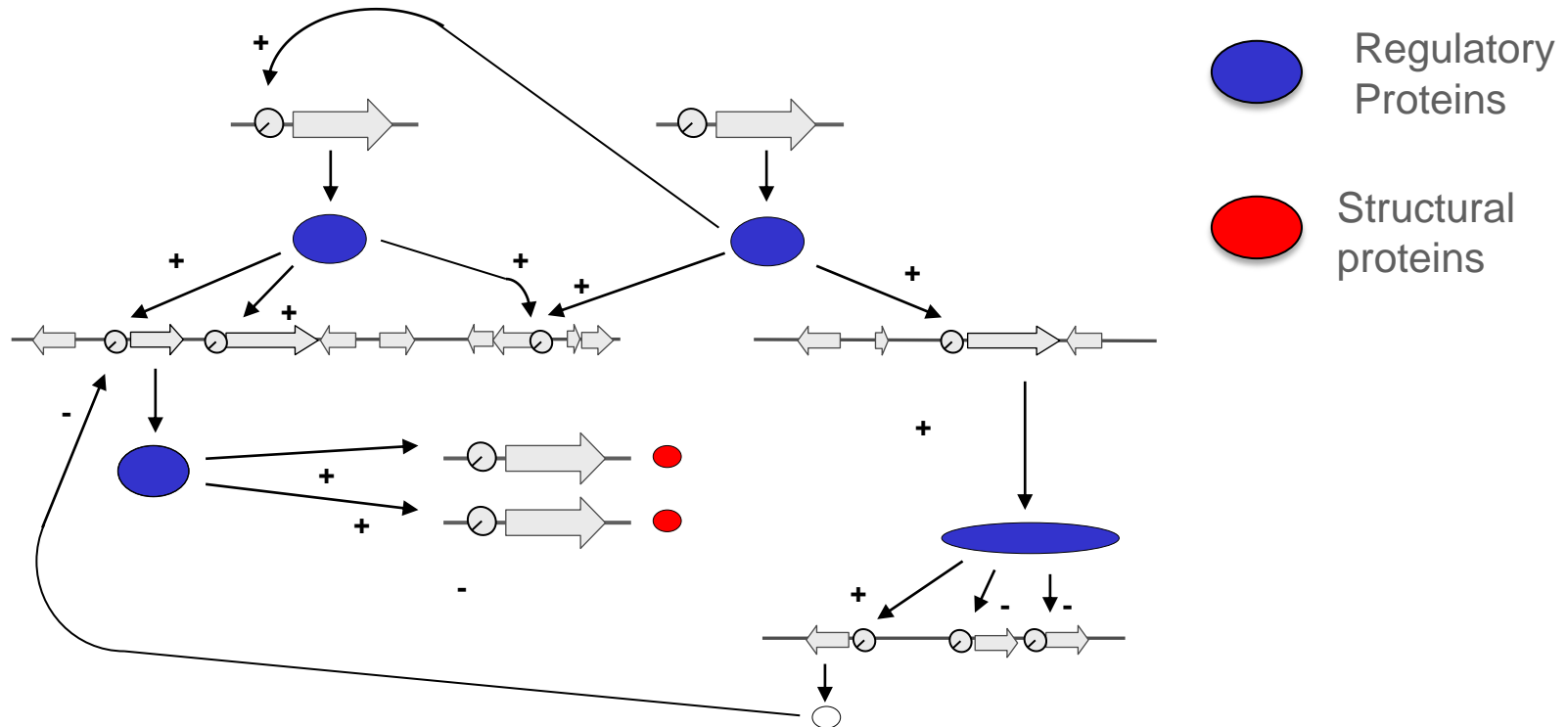
Systems Biology



Systems Biology



Systems Biology



Static network is encoded in the genome

Not all proteins are made at all times (network is condition-dependent)

State of the network can be measured through gene/protein expression

Systems Biology



computer cluster



cell culture robot



HPLC



PCR



microarray platform



MALDI-TOF mass spectrometer



DNA sequencers

Encoding of the network

Genome

AGCACTGTCCACTGCATGGTGAGGATGGGGGTGAGCTCCCT
TTGTGGCTAGGTGCTTAAACGTCTATCGGACGCTCA**GTGAA**
GGGCTATTGCGGGTAACAGGAAACCCTGAGGTGCT
GATAGGTCAAAGATGGAGAAGGCGTAGGGATATGTG
TTGGCAGAGGAACCGAAGAATACCAGGCCATTTCCG
AAGCCACTCATTTTCCTCGAATTCCTTCTTTATGCCTTC
AGTCTCTATTGACCGTAAATTTGGTTGTTGTCTCCCAGCTGT
TTATTTCTGTAACAGATCTTGGAGGCTGCGGTCTGGATCCCT
CGCCAAGAACCAGATCCAGGAGAAAACGTGCTCAACGTGCA
GCTCTGCTCCTACTGATTATAGCCCCACAGATGACATCGCTC
CATAGTCACACCAAGTCTCCTGTGGGAGTCTTGCTCCTCGTT
CTCAGTGTCTGTTACAGCTCGGTATTTTAGTGTGACGACGTC
GGCTCCCAGCCCGCATCTCCGCTCAGCAATGCCATTATCTTC
TCAGCCAAGTCCTAGAAATGGGTTG**GCTTCCCATTTGCAA**
AAACATCGCTCCATAGTCACACCAAGTCTCCTGTGGG
AGTCTTGCTCCTCGTTCTCAGTGTCTGTTACAGCTCG
GTATTTTAGTGTGACGACGTCGGCTCCCAGCCCGCAT
CTCCGCTCAGCAATGCCATTATCTTCTCAGCCAAGTCCT
AGAAATGGGTTGGCTTCCCATTTGCAAAAACATCGCTCCATA
GTCACACCAAGTCTCCTGTGGG

Encoding of the network

THE HUMAN GENOME

The Sequence of the Human Genome

J. Craig Venter,^{1*} Mark D. Adams,¹ Eugene W. Myers,¹ Peter W. Li,¹ Richard J. Mural,¹ Granger G. Sutton,¹ Hamilton O. Smith,¹ Mark Yandell,¹ Cheryl A. Evans,¹ Robert A. Holt,¹ Jeannine D. Gocayne,¹ Peter Amanatides,¹ Richard M. Ballew,¹ Daniel H. Huson,¹ Jennifer Russo Wortman,¹ Qing Zhang,¹ Chinappa D. Kodira,¹ Xiangqun H. Zheng,¹ Lin Ch. Marian Skupski,¹ Gangadharan Subramanian,¹ Paul D. Thomas,¹ Jinghui Zhang,¹ George L. Gabor Miklos,² Catherine Nelson,³ Samuel Broder,¹ Andrew G. Clark,⁴ Joe Nade Victor A. McKusick,⁵ Norton Zinder,⁷ Arnold J. Levine,⁷ Richard J. Roberts,⁸ Mel Simon Carolyn Slayman,¹⁰ Michael Hunkapiller,¹¹ Randall Bolanos,¹ Arthur Delcher,¹ Ian Dew,¹ Danie Michael Flanagan,¹ Liliana Florea,¹ Aaron Halpern,¹ Sridhar Hannenhalli,¹ Saul Kravitz,¹ Samu Clark Mobarry,¹ Knut Reinert,¹ Karin Remington,¹ Jane Abu-Threideh,¹ Ellen Beasley,¹ Kendra Vivien Bonazzi,¹ Rhonda Brandon,¹ Michele Cargill,¹ Ishwar Chandramouliswaran,¹ Rosane Ci Kabir Chaturvedi,¹ Zuoming Deng,¹ Valentina Di Francesco,¹ Patrick Dunn,¹ Karen Elbec Carlos Evangelista,¹ Andrei E. Gabrielian,¹ Weiniu Gan,¹ Wangmao Ge,¹ Fangcheng Gong,¹ Zhi Ping Guan,¹ Thomas J. Helman,¹ Maureen E. Higgins,¹ Rui-Ru Ji,¹ Zhaoxi Ke,¹ Karen A. Ketch Zhongyu Lai,¹ Yiding Lei,¹ Zhenya Li,¹ Jayin Li,¹ Yong Liang,¹ Xiaoying Lin,¹ Fu Lu,¹ Gennady V. Merkulov,¹ Natalia Milshina,¹ Helen M. Moore,¹ Ashwinkumar K Naik,¹ Vaibhav A. Narayan,¹ Beena Neelam,¹ Deborah Nusskern,¹ Douglas B. Rusch,¹ Steven Salzberg Wei Shao,¹ Bixiong Shue,¹ Jingtao Sun,¹ Zhen Yuan Wang,¹ Aihui Wang,¹ Xin Wang,¹ Jian V Ming-Hui Wei,¹ Ron Wides,¹² Chunlin Xiao,¹ Chunhua Yan,¹ Alison Yao,¹ Jane Ye,¹ Ming Zi Weiqing Zhang,¹ Hongyu Zhang,¹ Qi Zhao,¹ Liansheng Zheng,¹ Fei Zhong,¹ Wenyuan Zhor Shiaoqing C. Zhu,¹ Shaying Zhao,¹² Dennis Gilbert,² Suzanne Baumhueter,¹ Gene Spier,¹ ...

Human Genome Project 16/02/2001

articles

Initial sequencing and analysis of the human genome

International Human Genome Sequencing

*A partial list of authors appears on the opposite page.

The human genome holds an extraordinary Here we report the results of an international effort to sequence and analyze the human genome. We also present an initial analysis

The rediscovery of Mendel's laws of heredity the 20th century¹ sparked a scientific of nature and content of genetic information biology for the last hundred years. The se falls naturally into four main phases, corres four quarters of the century. The first establi heredity: the chromosomes. The second defi of heredity: the DNA double helix. The third tional basis of heredity, with the discovery of son by which cells read the information cont the invention of the recombinant DNA tech sequencing by which scientists can do the s. The last quarter of a century has been mar to decipher first genes and then entire geno of genomics. The fruits of this work alrea fangs, two animals and one plant.

Here we report the results of a collaborat from the United States, the United Ki Germany and China to produce a draft s genome. The draft genome sequence was a map covering more than 96% of the euchro genome and, together with additional sequ it covers about 94% of the human geno produced over a relatively short period, wi about 10% to more than 90% over rougl sequence data have been made availabl updated daily throughout the project. The finished sequence, by closing all gaps and r Already about one billion bases are in fin bringing the vast majority of the sequen straightforward and should proceed rapidl. The sequence of the human genome i respects. It is the largest genome to be exte being 25 times as large as any previousl eight times as large as the sum of all such vertebrate genome to be extensively sequen the genome of our own species.

Much work remains to be done to prod sequence, but the vast trove of inform available through this collaborative effort all on the human genome. Although the det sequence is finished, many points are alac. The genomic landscape shows marked v tion of a number of features, includin elements, GC content, CpG islands and re gives us important clues about function. I omentally important HOX gene clusters a regions of the human genome, probably tel

Genome Sequencing Centres (Listed in order of total genomic sequence contributed, with a partial list of personnel. A full list of contributors at each centre is available as Supplementary Information.)

Whitehead Institute for Biomedical Research, Center for Genome Research: Eric S. Lander¹, Lauren M. Linton¹, Bruce Birren¹, Chad Nusbaum¹, Michael C. Zody¹, Jennifer Baldwin¹, Kerl Devon¹, Ken Dewar¹, Michael Doyle¹, William Fitzhugh¹, Riel Funke¹, Diane Gage¹, Katrina Harris¹, Andrew Heathrow¹, John Howland¹, Lisa Kann¹, Jessica Lebecky¹, Rosie Levine¹, Paul McEwan¹, Kevin McKernan¹, James Meldrum¹, Jill P. Mesirov¹, Cher Miranda¹, William Morris¹, Jerome Naylor¹, Christina Raymond¹, Mark Rosetti¹, Ralph Santos¹, Andrew Sheridan¹, Carrie Sougnez¹, Nicole Stange-Thomann¹, Nikola Stojanovic¹, Aravind Subramanian¹ & Dudley Wyman¹

The Sanger Centre: Jane Rogers², John Sulston², Rachael Ainscough², Stephen Beck², David Bentley², John Burton², Christopher Clew², Nigel Carter², Alan Coulson², Rebecca Deadman², Panos Deloukas², Andrew Dunham², Ian Dunham², Richard Durbin², Lisa French², Darren Grafham², Simon Gregory², Tim Hubbard², Sean Humphray², Adrienne Hunt², Matthew Jones², Christine Lloyd², Amanda McMurray², Lucy Matthews², Simon Mercer², Sarah Milne², James C. Mullikin², Andrew Mungall², Robert Plumb², Mark Ross², Ratna Showkneet² & Sarah Sims²

Washington University Genome Sequencing Center: Robert H. Waterston³, Richard K. Wilson³, Ludwina W. Hillier³, John D. McPherson³, Marco A. Marra³, Elaine R. Mardis³, Lucinda A. Fulton³, Asif T. Chinnwalla³, Kymberlie H. Pepin³, Warren R. Gish³, Stephanie L. Chissoe³, Michael C. Wendl³, Kim D. Delahunty³, Tracie L. Miner³, Andrew Delabaunty³, Jason B. Kramer³, Lisa L. Cook³, Robert S. Fulton³, Douglas L. Johnson³, Patrick J. Minx³ & Sandra W. Clifton³

US DOE Joint Genome Institute: Trevor Hawkins⁴, Elbert Brancome⁴, Paul Preduel⁴, Paul Richardson⁴, Sarah Wenning⁴, Tom Slezacek⁴, Norman Doggett⁴, Jan-Fang Cheng⁴, Anne Olsen⁴, Susan Lucas⁴, Christopher Elkin⁴, Edward Uberbacher⁴ & Marvin Frazier⁴

Baylor College of Medicine Human Genome Sequencing Center: Richard A. Gibbs⁵, Donna M. Muzny⁵, Steven E. Scherer⁵, John B. Bouck⁵, Erica J. Sodergren⁵, Kim C. Worley⁵, Catherine M. Rives⁵, James R. Gorecki⁵, Michael L. Metzker⁵, Susan L. Naylor⁵, Raja S. Kucherlapati⁵, David L. Nelson⁵, & George M. Weinstock⁵

RIKEN Genomic Sciences Center: Yoshiyuki Sakaki⁶, Asao Fujiyama⁶, Masahira Hattori⁶, Tetsushi Yada⁶, Atsushi Toyoda⁶, Takehiko Itoh⁶, Chiharu Kawagoe⁶, Hidemi Watanabe⁶, Yasushi Totoki⁶ & Todd Taylor⁶

Genoscope and CNRS UMR-8030: Jean Weissenbach⁷, Roland Heilig⁷, William Saurin⁷, Francois Artiguenave⁷, Philippe Brottier⁷, Thomas Bruls⁷, Eric Pelletier⁷, Catherine Robert⁷ & Patrick Wincker⁷

GTC Sequencing Center: Douglas R. Smith⁸, Lynn Doucette-Stamm⁸, Marc Rubenfield⁸, Keith Weinstock⁸, Hong Mei Lee⁸ & JoAnn Dubois⁸

Department of Genome Analysis, Institute of Molecular

Biotechnology: André Rosenthal⁹, M. Gerald Nyakatura⁹, Stefan Taudien⁹

Beijing Genomics Institute/Human Genome Project: Huangming Yang¹⁰, Jun Yu¹⁰, Jian Wan & Jun Gu¹⁰

Multibase Sequencing Center, Biology: Leroy Hood¹¹, Lee Rowen¹¹

Stanford Genome Technology Center: Nancy A. Federspiel¹², A. Pia Abola¹²

Stanford Human Genome Center: Ric Jeremy Schmutz¹³, Mark Dickson¹³, J. & David R. Cox¹³

University of Washington Genome Center: Rajinder Kaul¹⁴ & Christopher Raymon

Department of Molecular Biology, Ke Medicine: Nobuyoshi Shimizu¹⁵, Kazuo & Shinsai Minoshima¹⁵

University of Texas Southwestern Me Glen A. Evans¹⁶, Maria Athanasio¹⁶

University of Oklahoma's Advanced C Technology: Bruce A. Roe¹⁷, Feng Che

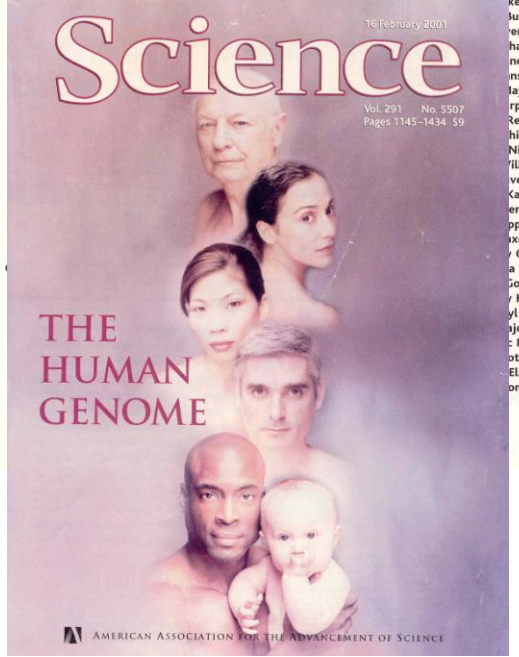
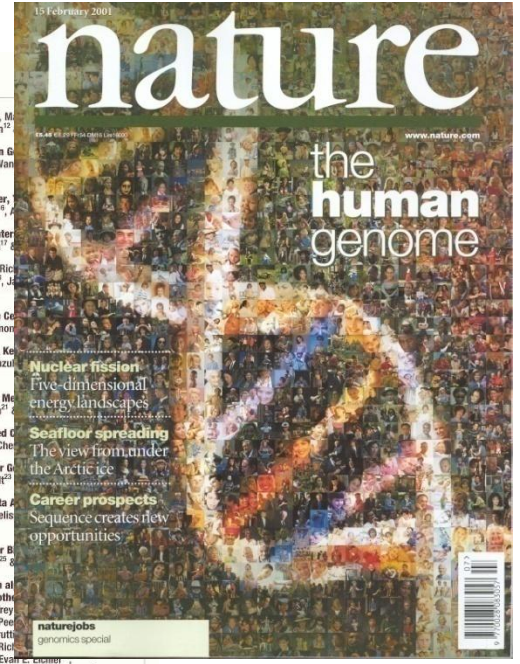
Max Planck Institute for Molecular G Hans Lehrach¹⁸ & Richard Reinhardt¹⁸

Cold Spring Harbor Laboratory, Lita Center: W. Richard McCombie¹⁹, Melis & Nelsa Dedisi¹⁹

GBF—German Research Center for B Helmut Blöcker²⁰, Klaus Hornischer²⁰

Genome Analysis Group (listed in al includes individuals listed under othe Richa Agarwala²¹, L. Aravind²¹, Jeffrey Serafini Batzoglou²¹, Ewan Birney²¹, Peo Christopher B. Burge²¹, Lorenzo Cerutti Deanna Church²¹, Michele Clamp²¹, Ric Tobias Doerks²¹, Sean R. Eddy²¹, Evan Ewinger²¹, Terrence S. Furey²¹, James Galagan²¹, James G. R. Gilbert²¹, Cyrus Harmon²¹, Yoshihide Hayashizaki²¹, David Haussler²¹, Henning Hermjakob²¹, Karsten Hokamp²¹, Wonhee Jang²¹, L. Steven Johnson²¹, Thomas A. Jones²¹, Simon Kasif²¹, Arik Kasprzyk²¹, Scott Kennedy²¹, W. James Kent²¹, Paul Kitts²¹, Eugene V. Koonin²¹, Ian Korf²¹, David Kulp²¹, Doron Lancet²¹, Todd M. Lowe²¹, Anile McLysaght²¹, Tarjei Mikkelson²¹, John V. Moran²¹, Nicola Mulder²¹, Victor J. Pollara²¹, Chris P. Ponting²¹, Greg Schuler²¹, Jörg Schultze²¹, Guy Slater²¹, Arfan F. A. Smit²¹, Eita Stupka²¹, Joseph Szustakowski²¹, Danielle Thierry-Mieg²¹, Jean Thierry-Mieg²¹, Lukas Wagner²¹, John Wallis²¹, Raymond Wheeler²¹, Alan Williams²¹, Yuri I. Wolf²¹, Kenneth H. Wolfe²¹, Shiao-Pyng Yang²¹ & Ru-Fang Ye²¹

Scientific management: National Human Genome Research Institute, US National Institutes of Health: Francis Collins²², Mark S. Guyer²², Jane Peterson²², Adam Feisenfeld²², & Kris A. Wetterstrand²², Office of Science, US Department of Energy: Astrides Patrinos²², The Wellcome Trust: Michael J. Morgan²²



1304

NATURE VOL 409 16 FEBRUARY 2001 www.nature.com

861

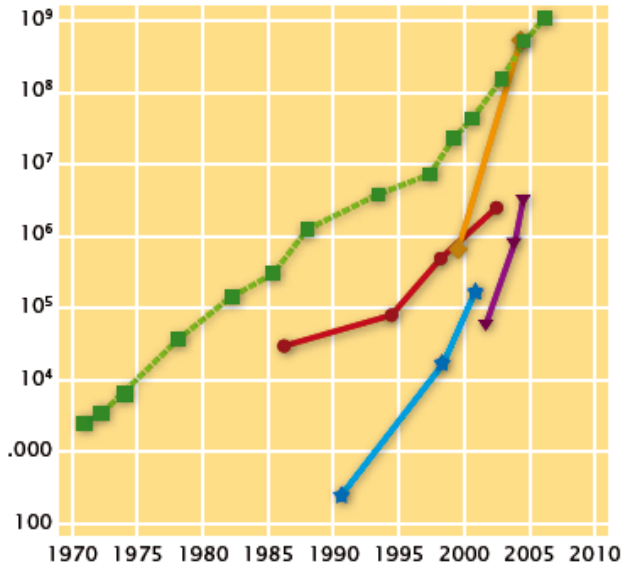


Sequencing human genome: 13 years/ 3 miljard dollar



Encoding of the network

Next generation sequencing follows Moore Law



DNA-Analyse

ABI Sequence machine

Pyrosequence machine

Aantal transistoren per chip

DNA-Synthese

ABI Synthesizer

Egea GeneWriter

De wet van Moore (exponentiële toename van de snelheid) is ook van toepassing op de analyse en synthese van DNA

454

Solid: 50 Gb/run;

Helicos

Illumina: 25 Gb/run; 75 bp reads

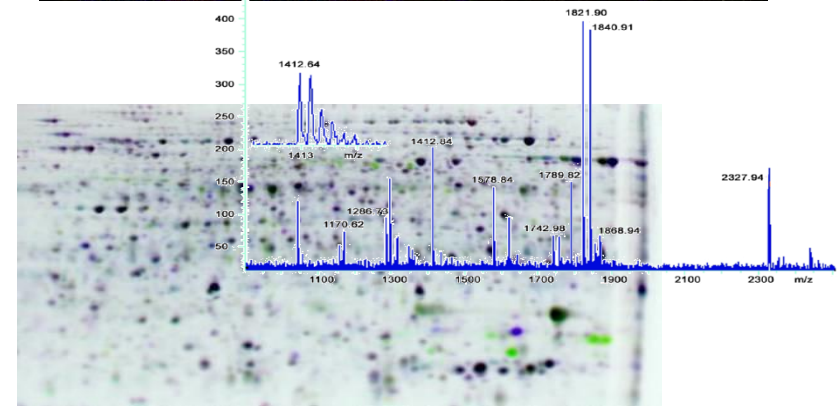
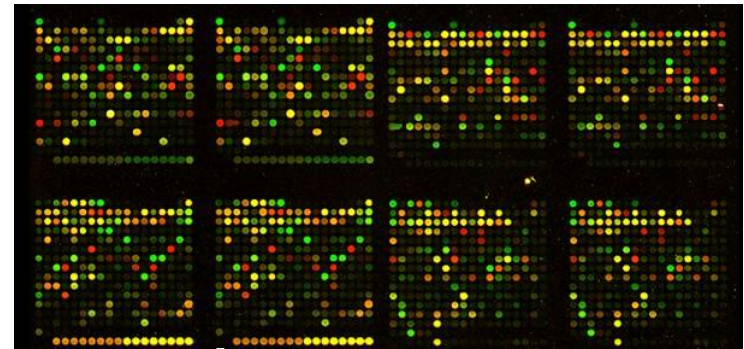
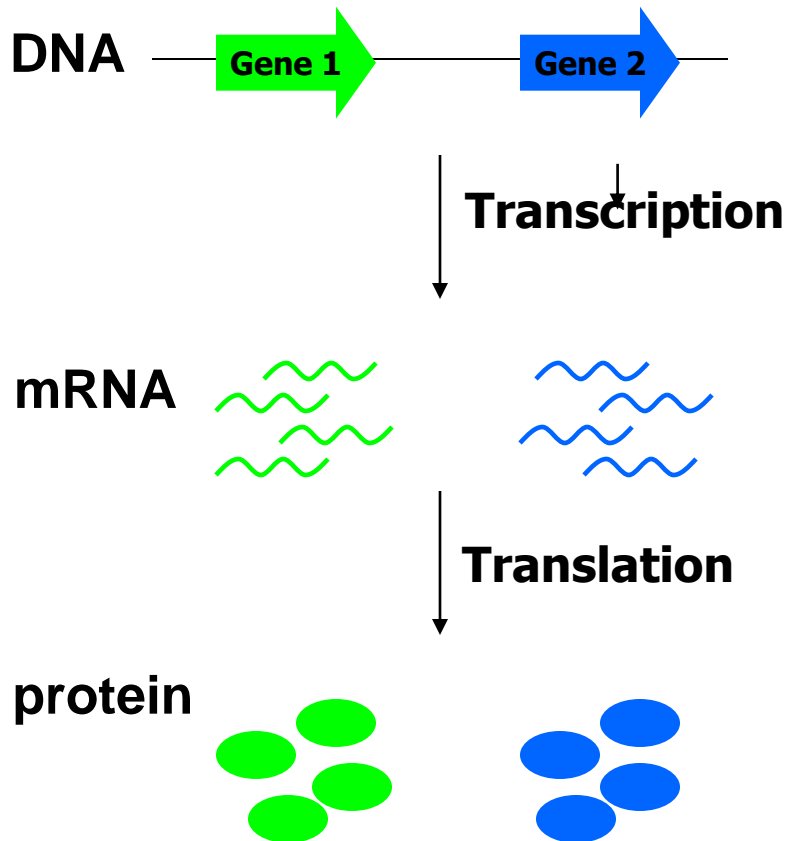
Encoding of the network

Next generation sequencing technology

- **Sequencing human genome: 13 years/ 3 miljard dollar**
- **Genome Watson (454 techn): 20 persons/2 months. Totaal 1.000.000 dollar**
- **Now: 1000 dollar human genome**

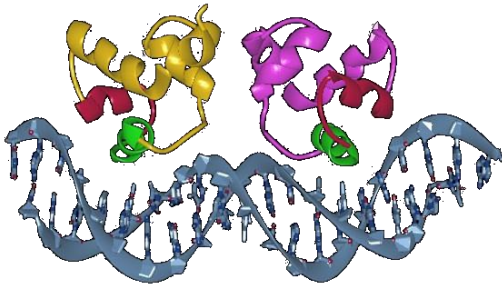
State of the network

Functional data: transcriptome, proteome, metabolome

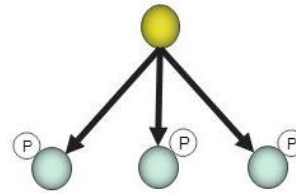


Interactions of the network

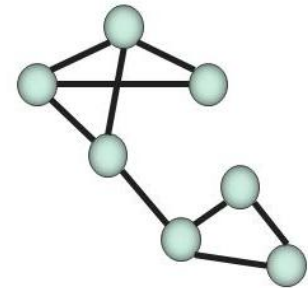
Physical data:



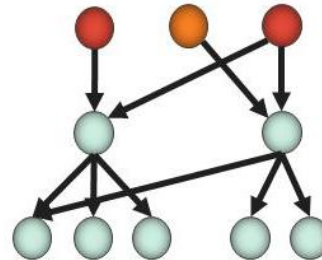
Signaling network



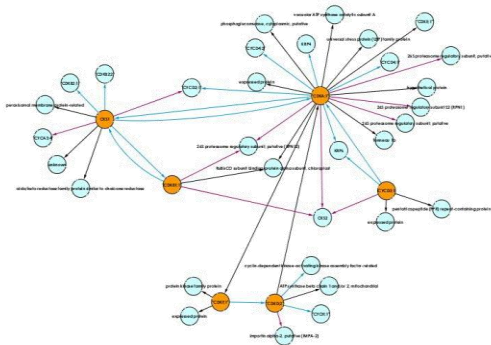
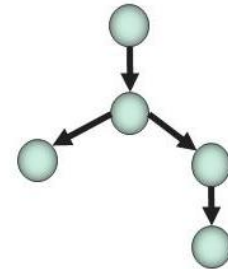
Protein interaction network



(Post)Transcriptional network

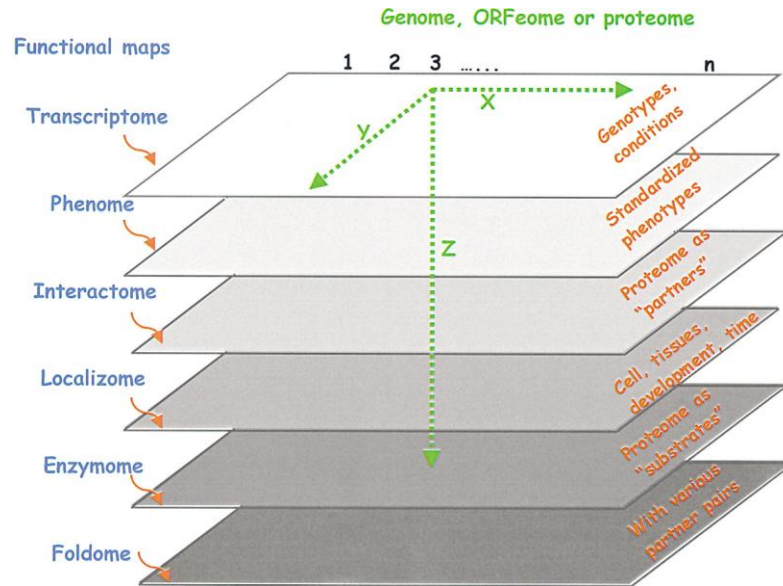


Metabolic network

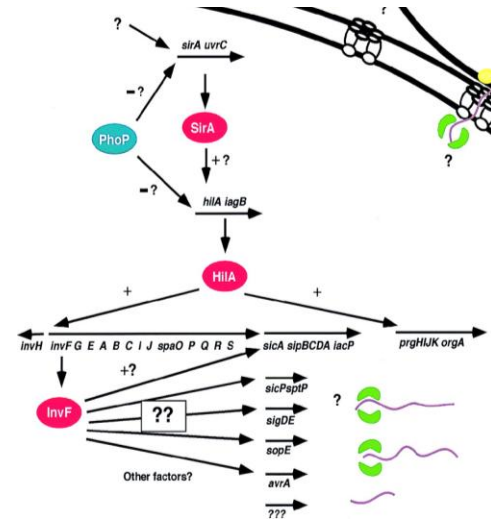


PHYSICAL INTERACTION
NETWORKS

Inferring the network



High throughput data



Mechanistic insight in the biological system at molecular biological level (holistic insight)

Inferring the network

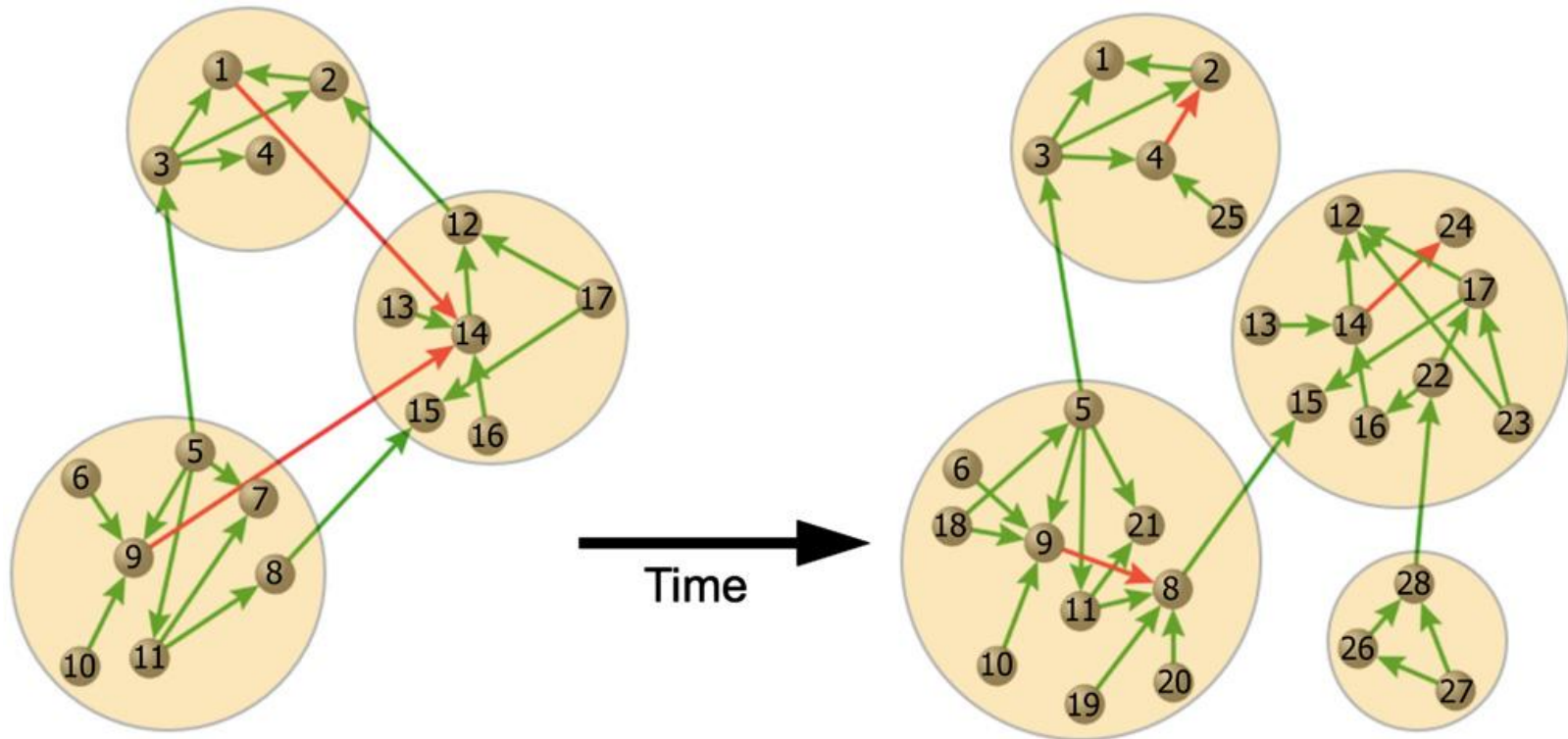
- **Omics data are noisy**
- **Omics data are incomplete**

- **Integrate different data to obtain higher precision and coverage**

- **Reconstruct network**
 - Different datasources
 - Different Molecular layers

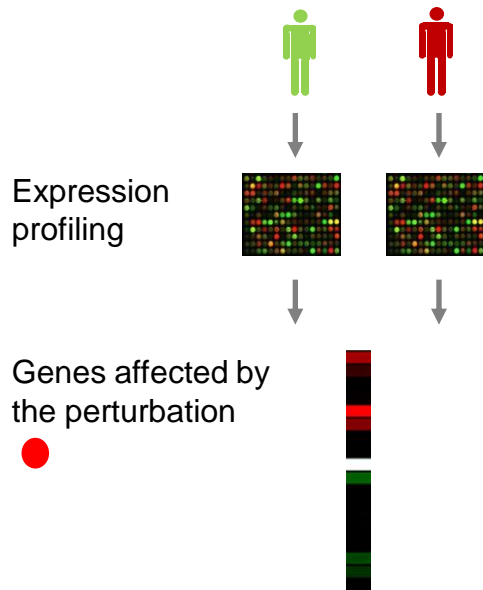
Fundamental knowledge

Evolution: comparing network between species or over time

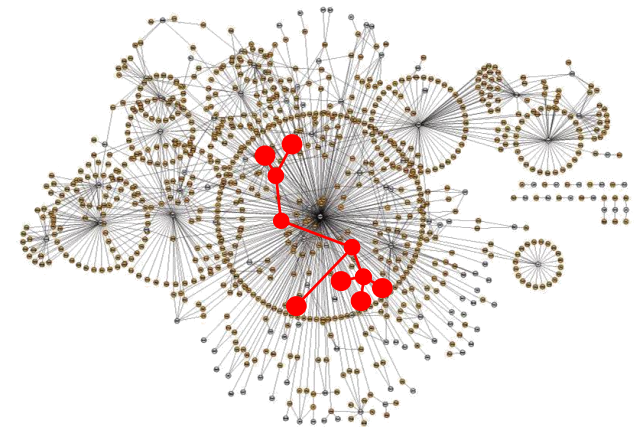


Using the network to interpret data

In house data



Physical interaction network



Infer the **hidden paths** between genomic variations and expression alteration

Network-based analysis of unstructured gene lists

Bioinformatics and datamining

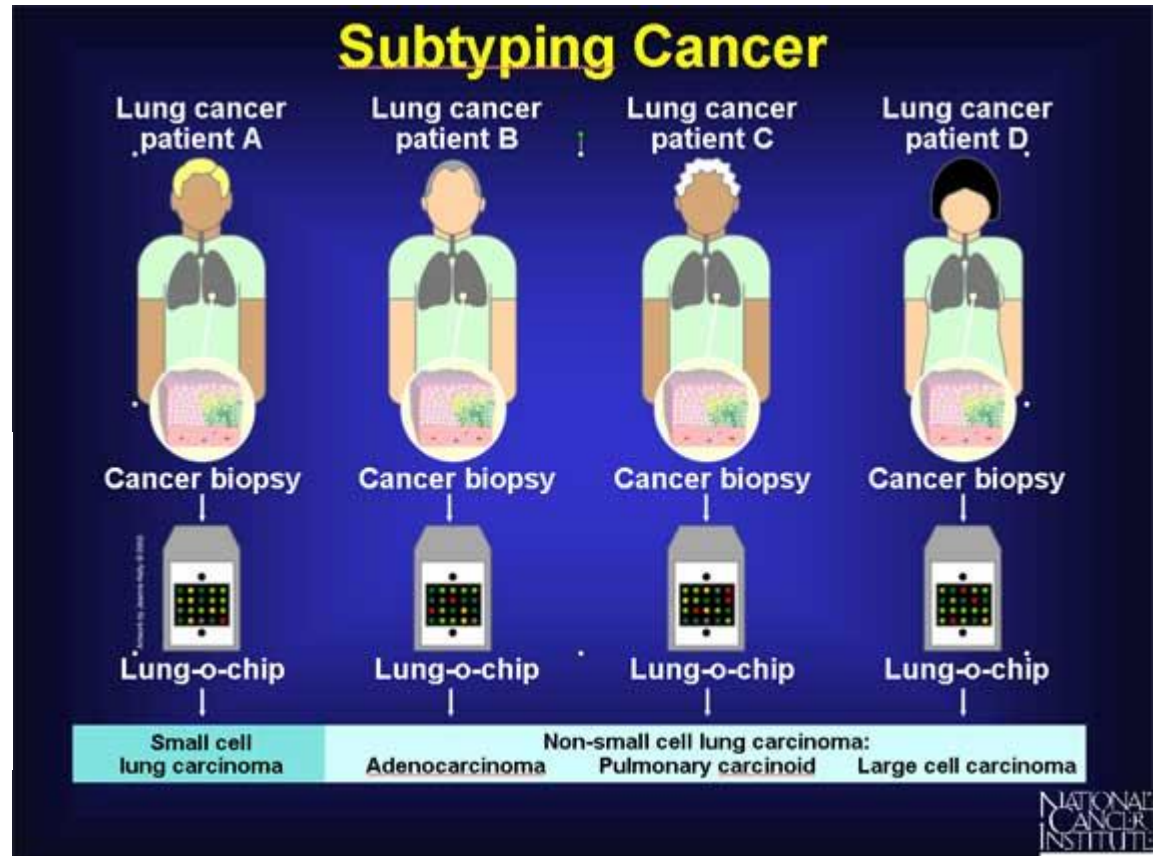
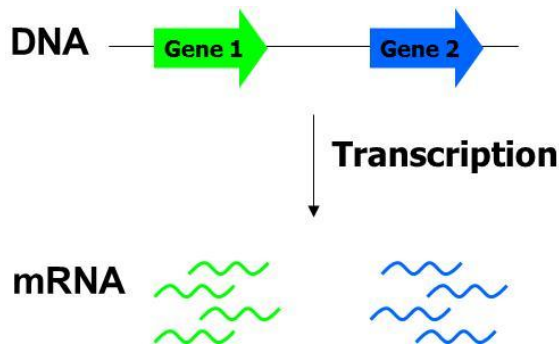
What does it require applying a computer science framework to bioinformatics?

Fast solution

- **The wet lab scientist rules**
- **Competition is fierce**
 - **Often the high impact papers are the conceptual ones**
 - **You tackle a research problem for the FIRST time**
- **Biological message is more important than the method used to analyse the data**
 - **Code is sloppy , undocumented**
 - **Unsustainable code /tool development**

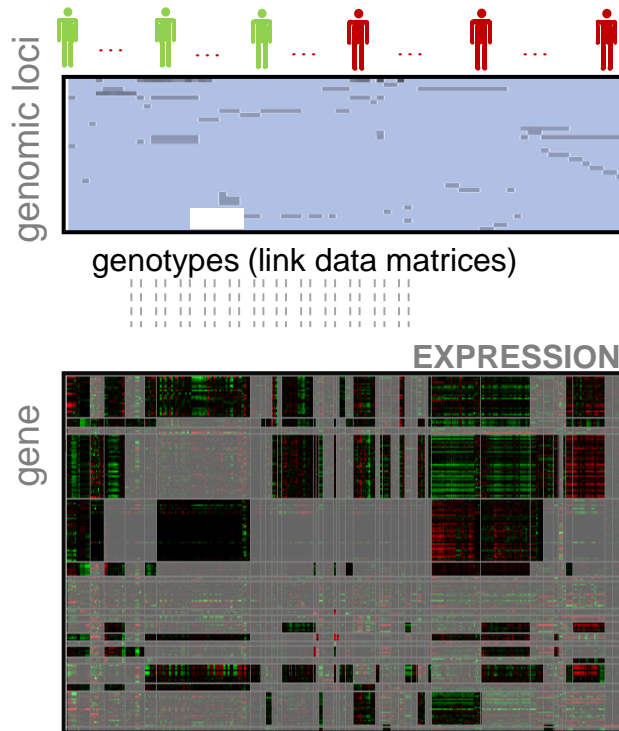
Problems are complex

Cancer subtyping & biomarker identification



Classification problem

Problems are complex

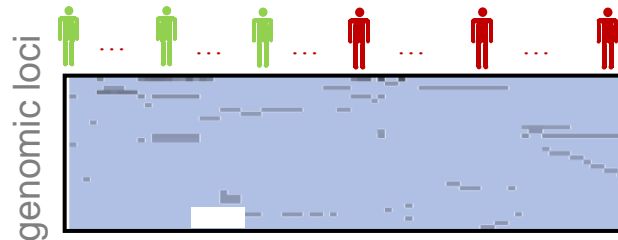


Dataintegration problem

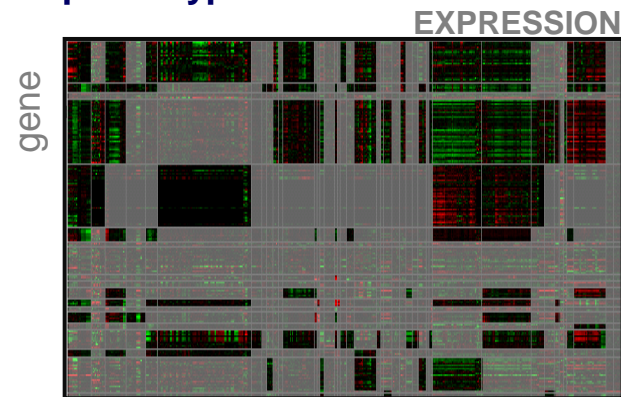
Problems are complex

Cancer subtyping

Preprocess the data

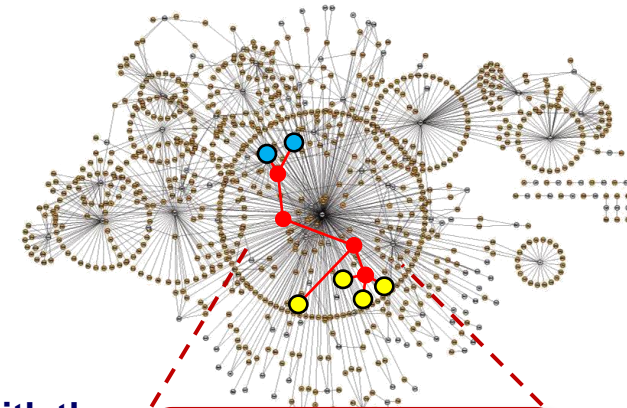


Associate genotypes to phenotype

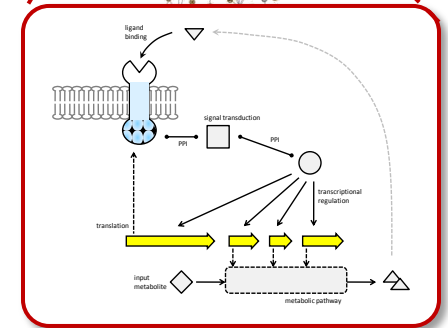


Data integration problem

Infer the interaction network from the omics data



Integrate all data with the interaction network

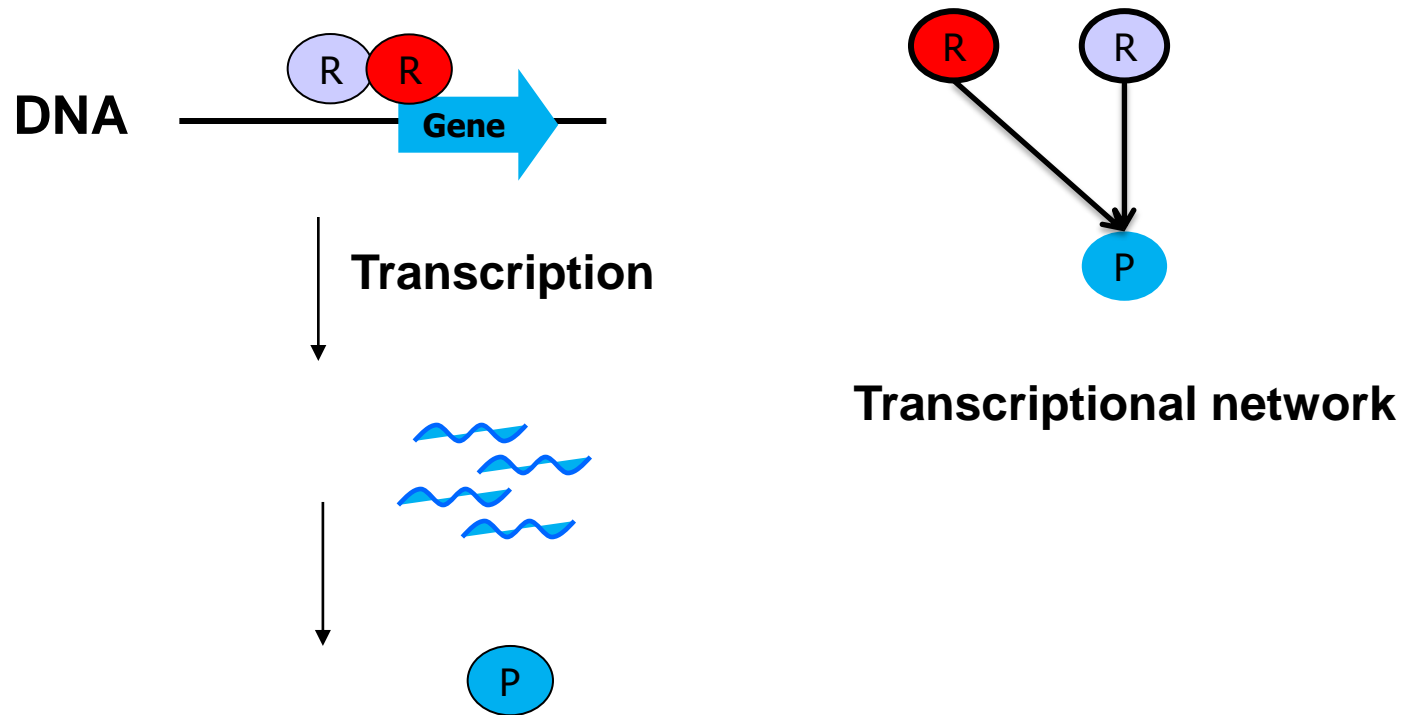


Bioinformatics and datamining

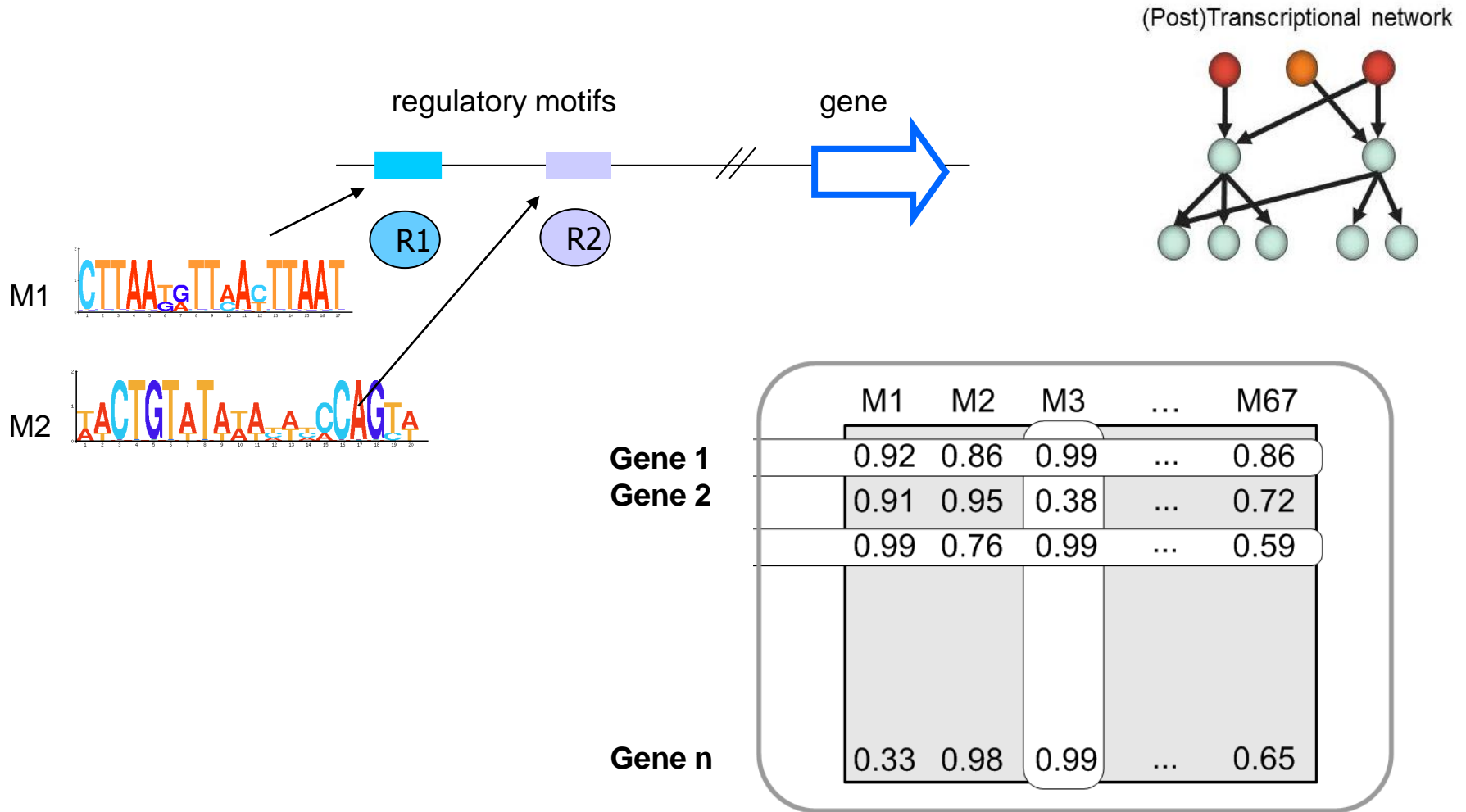
- Problems need a **fast solution**
- Problems are **increasingly complex** and can not be solved by one particular datamining tool (generic knowledge needed)
- Datamining in bioinformatics requires a quite **thorough understanding of biology**
- Problems are **underdetermined and ill defined**

Network inference

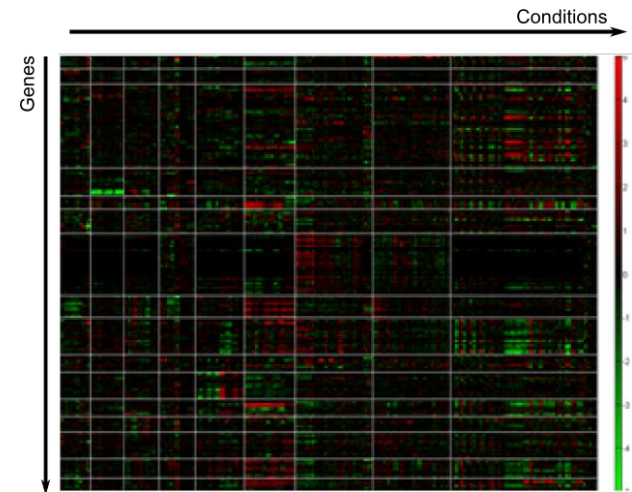
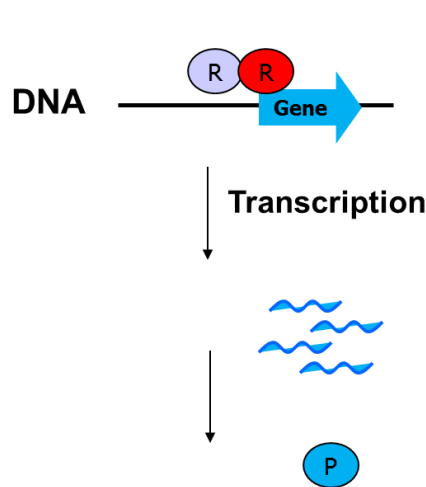
Transcriptional network inference



Network inference



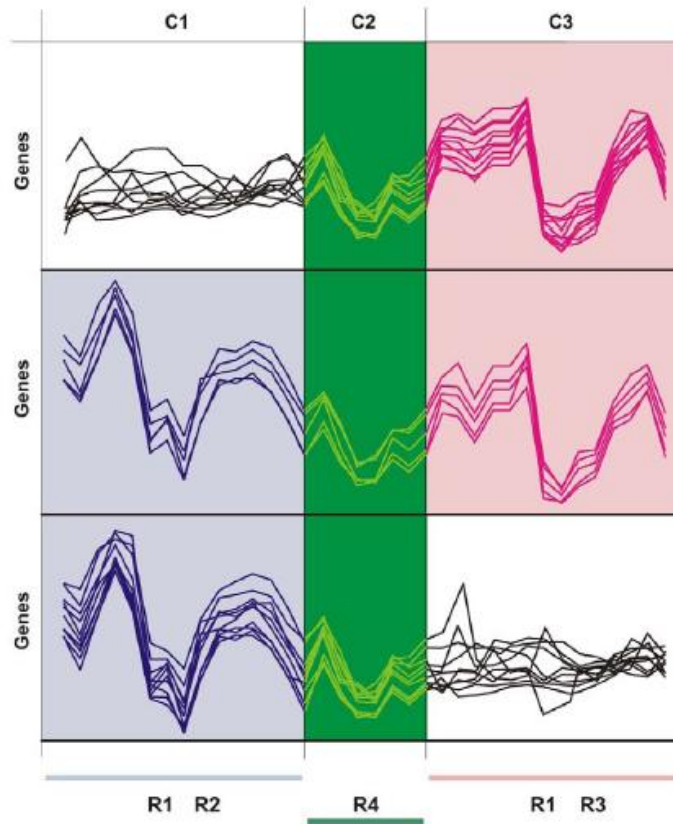
Network inference



	A1	A2	A3	...	A870
gene 1	4.92	1.09	-3.34	...	2.06
gene 2	-2.21	0.35	-4.98	...	0.12
gene 3	4.45	-2.51	-3.98	...	-3.41
...					
gene k	-4.56	-0.13	4.29	...	2.05

Network inference

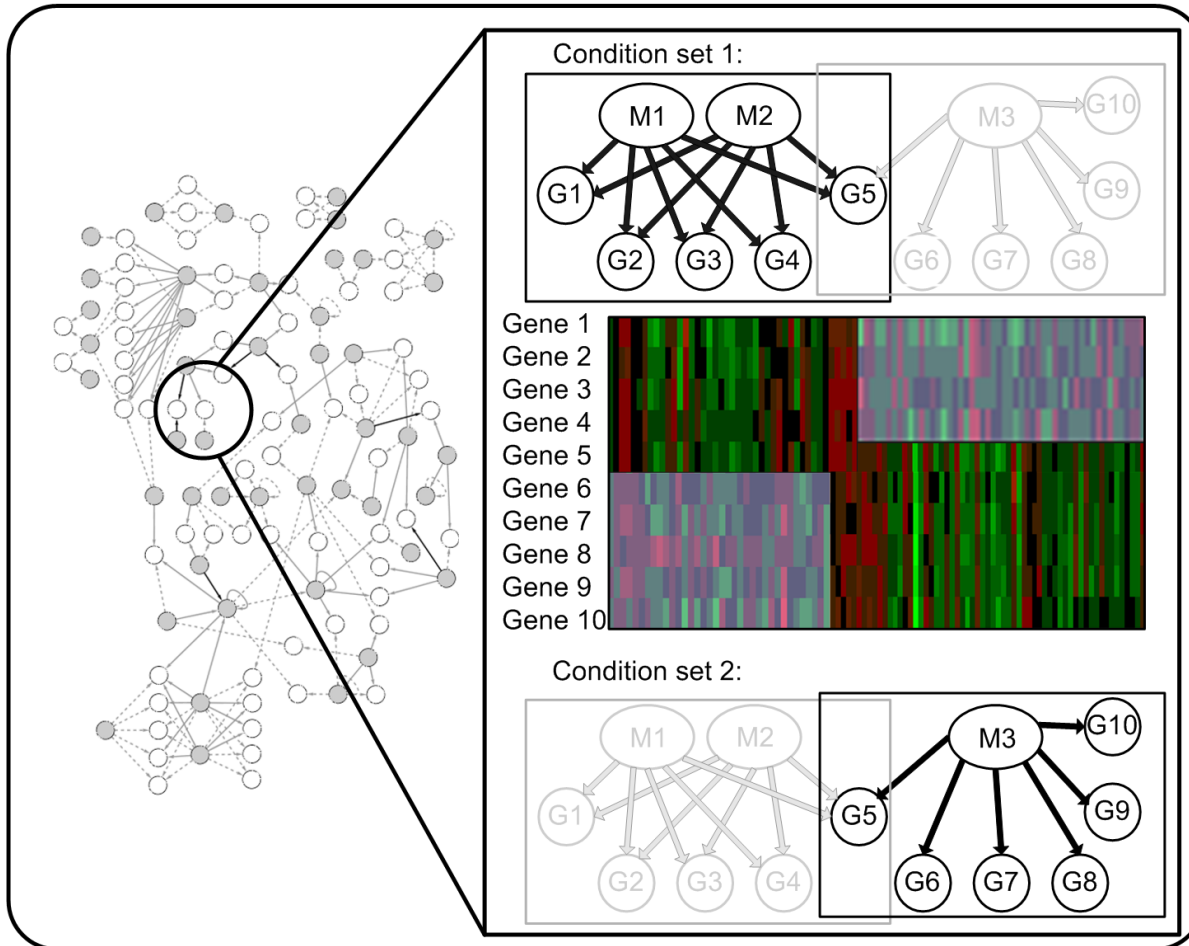
	A1	A2	A3	...	A870
gene 1	4.92	1.09	-3.34	...	2.06
gene 2	-2.21	0.35	-4.98	...	0.12
gene 3	4.45	-2.51	-3.98	...	-3.41
...					
gene k	-4.56	-0.13	4.29	...	2.05



Guilt by association

Coexpressed target genes are coregulated

Network inference



Network inference

DISTILLER

Data Integration System

Lemmens et al. Genome Biol. 2009

To Identify Links in Expression in Expression regulation

Search co expression modules that meet minimal requirements:

- All genes are significantly co-expressed in a sufficiently large, *a priori* unspecified set of experimental conditions CC
- All genes contain motif instances for a sufficient number of common, *a priori* unspecified regulators CR

	A1	A2	A3	...	A870		M1	M2	M3	...	M67
gene 1	4.92	1.09	-3.34	...	2.06		0.92	0.86	0.99	...	0.86
gene 2	-2.21	0.35	-4.98		0.12		0.91	0.95	0.38	...	0.72
gene 3	4.45	-2.51	-3.98	...	-3.41		0.99	0.76	0.99	...	0.59
...											
gene k	-4.56	-0.13	4.29	...	2.05		0.33	0.98	0.99	...	0.65

Network inference

- Items = genes
- Transactions = conditions, motifs

Transactions	Items
M1	G1 G3 G5
M2	G3 G5 G9
M3	G3 G5 G11
M4	G1 G3 G5

Itemset 1: G3, G5 supported by 4 motifs

Itemset 2: G1,G3,G5 supported by 2 motifs

Tidset $t(G3,G5) = M1, M2, M3, M4$

Tidset $t(G1, G3,G5) = M1, M4$

Minimal support = 3

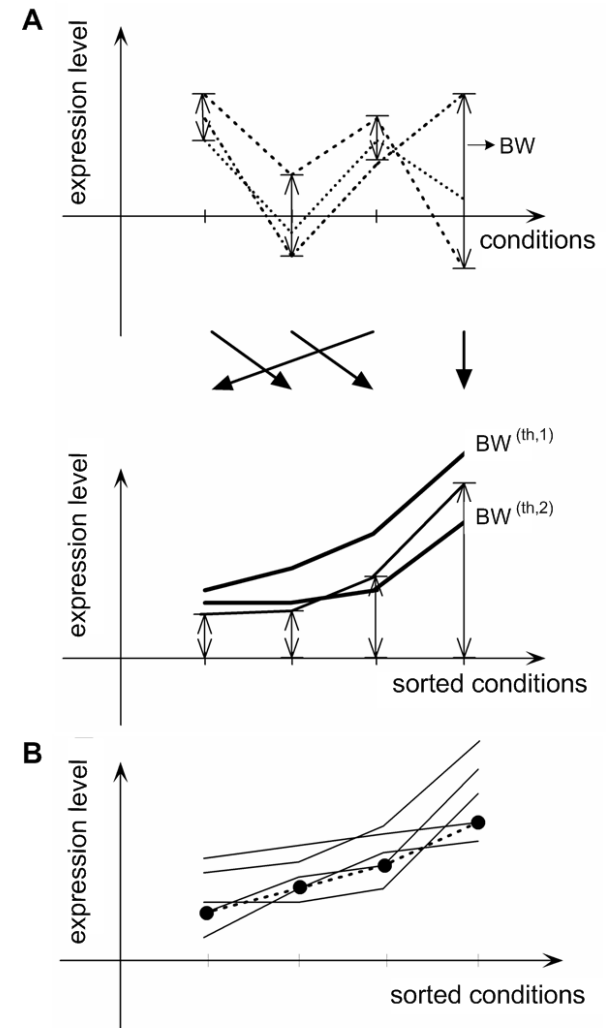
Itemset G3, G5 is frequent

- Supports:
 - All genes in the module (itemsets) are significantly co-expressed in a sufficiently large (**minimum support**), a priori unspecified set of experimental conditions
 - All genes in the module (itemsets) contain motif instances for a sufficient number R (**minimum support**) of common, a priori unspecified regulators

Network inference

Expression support
co-expression in a significant
number of experimental
conditions

- BW for each condition
- Rank BW in increasing order
- Check if BW sequence is within threshold BW sequence
- BW threshold sequence is determined by randomization



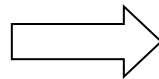
Network inference

Rank modules by assigning interest score

- $p(\text{module motif content})$:
The chance that a module with the same number of genes and the same number of motifs is found at random
- $p(\text{module expression pattern})$:
The chance that a module with at least the same number of genes and containing the same number of conditions is found by chance

Interest score:

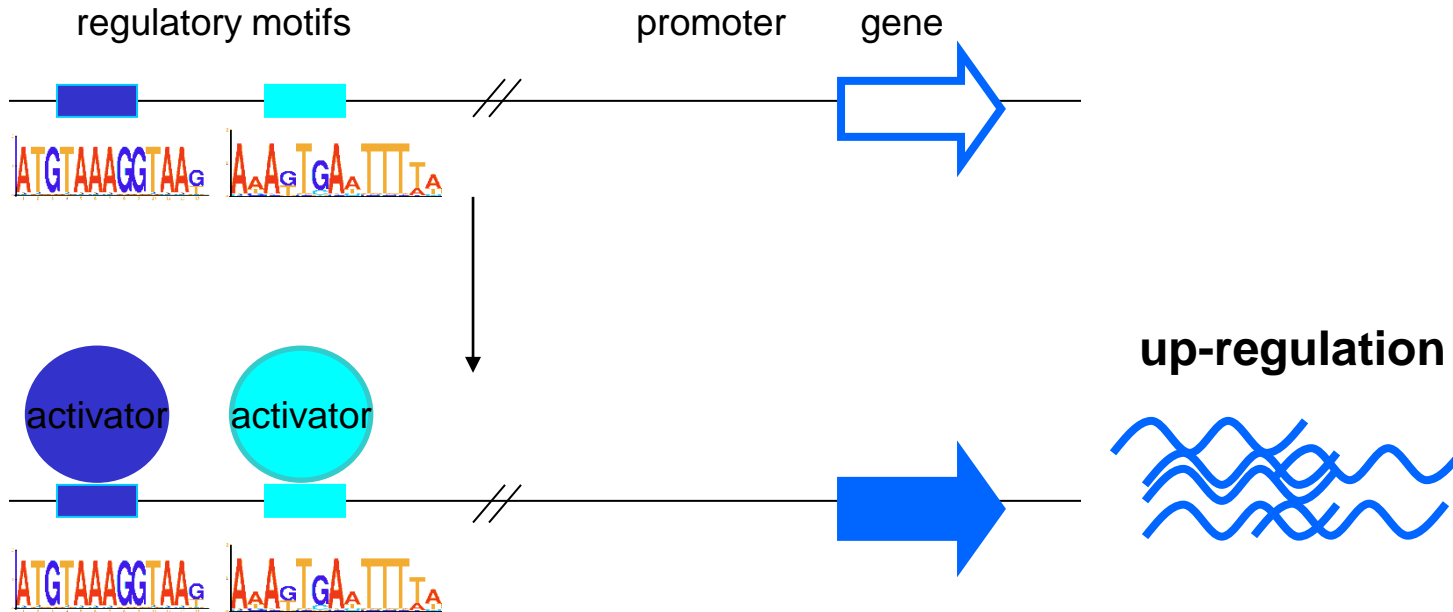
$$p(M|m,g) \times p(M|e,g)$$



Rank the modules

- Modules are selected iteratively such that they add as much as possible new information to the already selected modules

CRM detection



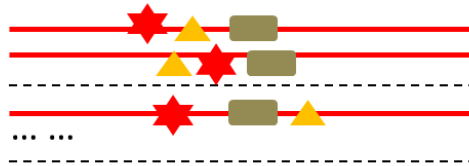
CRM: cis acting regulatory module
Combination of TF binding sites

CRM detection

Genes that are needed together in the cell usually are activated together = coregulation

Genome-wide enrichment score calculation and ranking

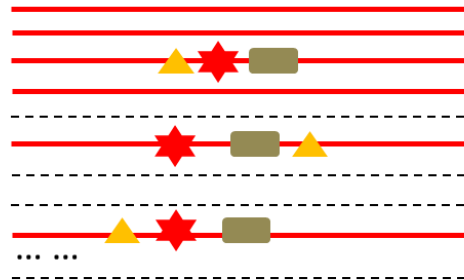
Input sequences



Valid CRM



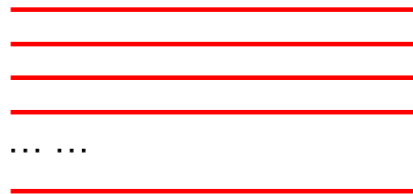
Background sequences



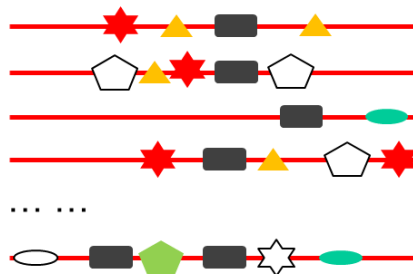
CRM detection

Input:

- 1) TRANSFAC PWM library
- 2) Set of sequences



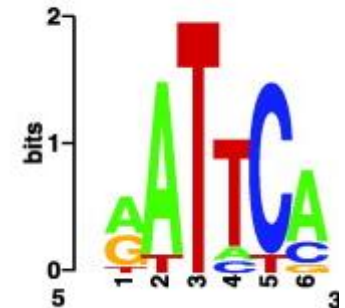
Motif screening and filtering



(C) PWM matrix

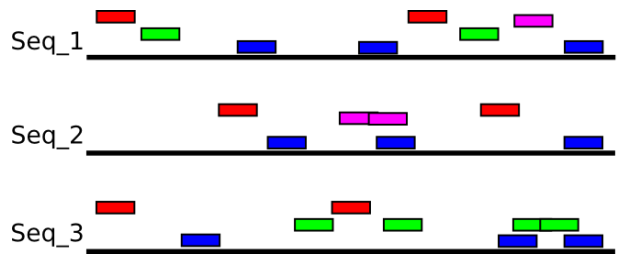
	1	2	3	4	5	6
A	0.90	1.71	-2.94	-1.06	-2.94	1.36
C	-2.94	-2.94	-2.94	-1.06	1.71	-0.28
G	0.61	-2.94	-2.94	-2.94	-2.94	-1.06
T	-1.06	-1.06	1.85	1.54	-1.06	-2.94

(D) sequence logo



- CRM is valid if motifs occur in each others neighbourhood
- Order needs to be conserved?

CRM detection



	m1	m2	m3	...	m66
seq 1	0	0	1	...	0
seq 2	0	1	0	...	0
seq 3	1	0	1	...	0
...					
seq k	0	1	1	...	0

DISTILLER

	Motif_1	Motif_2	Motif_3	Motif_4
Seq_1	(1,10) (65,74)	(12,20) (80,88)	(43,51) (53,61) (90,98)	(72,78)
Seq_2	(33,42) (85,94)		(49,57) (56,64) (91,99)	(50,56) (52,58)
Seq_3	(1,10) (82,91)	(45,53) (58,66) (75,83) (90,88)	(24,32) (72,80) (89,97)	

CPMODULE

Convert screening results in table with (start, stop) positions, for every sequence/motif pair

CRM detection

Constraint Programming (CP)

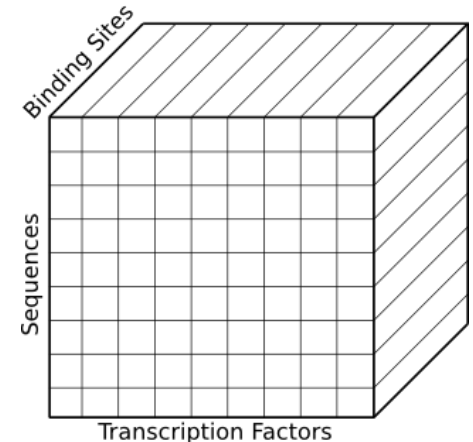
Model (by user):

Problem
specification
in terms of
constraints

Search (by solver):

Propagation: in which a constraint is used to remove values from the domain of variables that would violate it

Branching: in which a variable is assigned a value from its domain $D(v)$

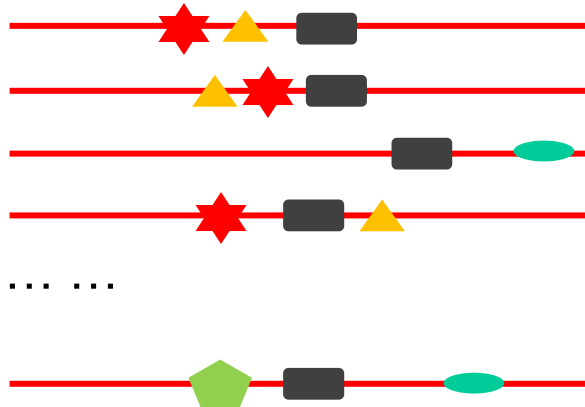


De Raedt et al. 2008 *KDD*

CRM detection

Frequency constraint

Frequency constraint = 2



Invalid CRM



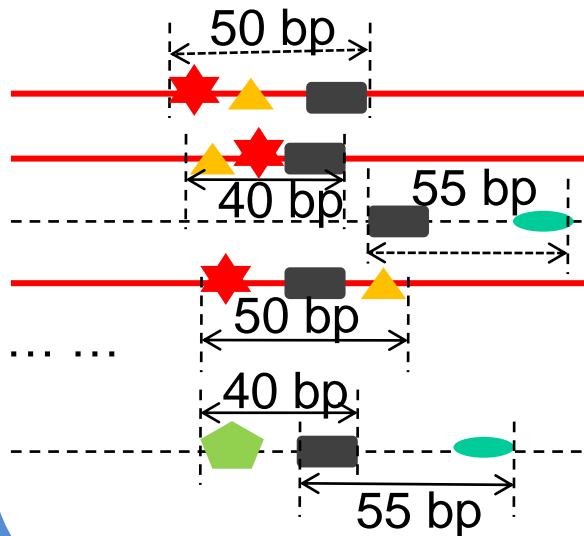
Valid CRM



The motif set should occur in a sufficient number of sequences (but not all) to be considered valid (support in itemset mining)

CRM detection

Proximity constraint



Invalid CRM



Valid CRM



Only motif instances that occur in each others proximity can contribute to a valid motif set (CRM)

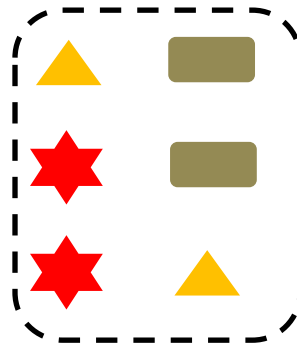
CRM detection

Redundancy constraint

Valid CRM



Invalid CRM



When enumerating all motif sets that meet the frequency and proximity constraint many subsets will occur in the same sequences and be composed of the same instances

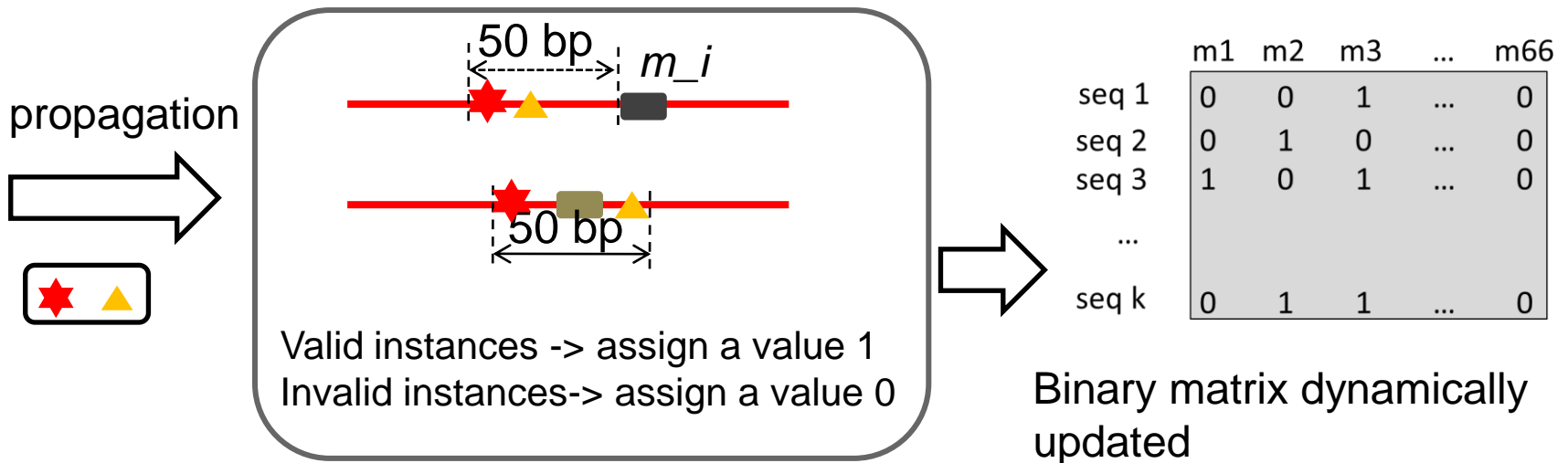
Only the motif set with more motifs will be retained

Removing redundant motif sets (CRMs) drastically increases the computation time (closeness in itemset mining)

CRM detection

Propagation: using a constraint to remove values from the domain of variables that would violate the constraint

Illustrate with the proximity constraint



Whether motif m_i is in the proximity of the motifs in motifset on sequence j ?

CRM detection

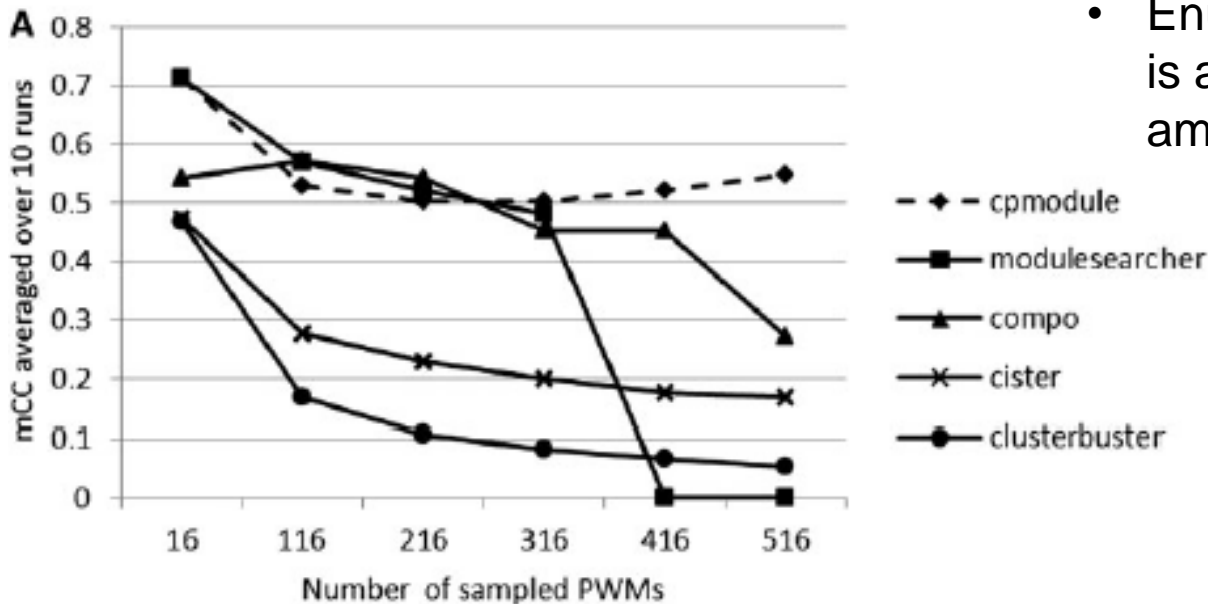
Benchmarked on a synthetic dataset

- Xie et al. 2008 (22 sequences)
- 516 TRANSFAC PWMs
- Motifs inserted from 3 known PWMs

CPModule: performances similar to state-of-the-art algorithms on a synthetic dataset

BUT

- Able to deal with much larger sequence sets
- Enumerating all solutions allows it is able to rank the true solution amongst all solution



Guns et al. 2010 BIBM
Sun et al. NAR, 2011

Bioinformatics and datamining

Optimal bioinformatics tool

- Right heuristics
- Proper biological assumptions
- Room for experimenting with different assumptions

- Modular code
- Sustainable code
- High performance
- Latest algorithmic developments

Usefulness of declarative framework (Prolog, ASP, Constrained based programming)

Acknowledgements

KUL/CMPG

- Ivan Ischukov*
- Hong Sun*
- Valerie Storms*
- Pieter Meysman*
- Kristof Engelen*
- Lore Cloots*
- Peyman Zarrineh*
- Riet De Smet*
- Karen Lemmens*
- Abeer Fadda*

UGENT/KUL/DBN

- Carolina Fierro
- Yan Wu
- Aminael Sanchez
- Marleen Claeys
- Dries De Maeyer
- Sergio Pullido
- Qiang Fu

University of Bristol

- Tijl Debie

KUL/Computer science

- Luc De Raedt
- Siegfried Nijssens
- Joris Renkens
- Tias Guns
- Tan Levan

<http://bioinformatics.psb.ugent.be/DBN/>