

Complex Network Analysis Jan.18th, 2018

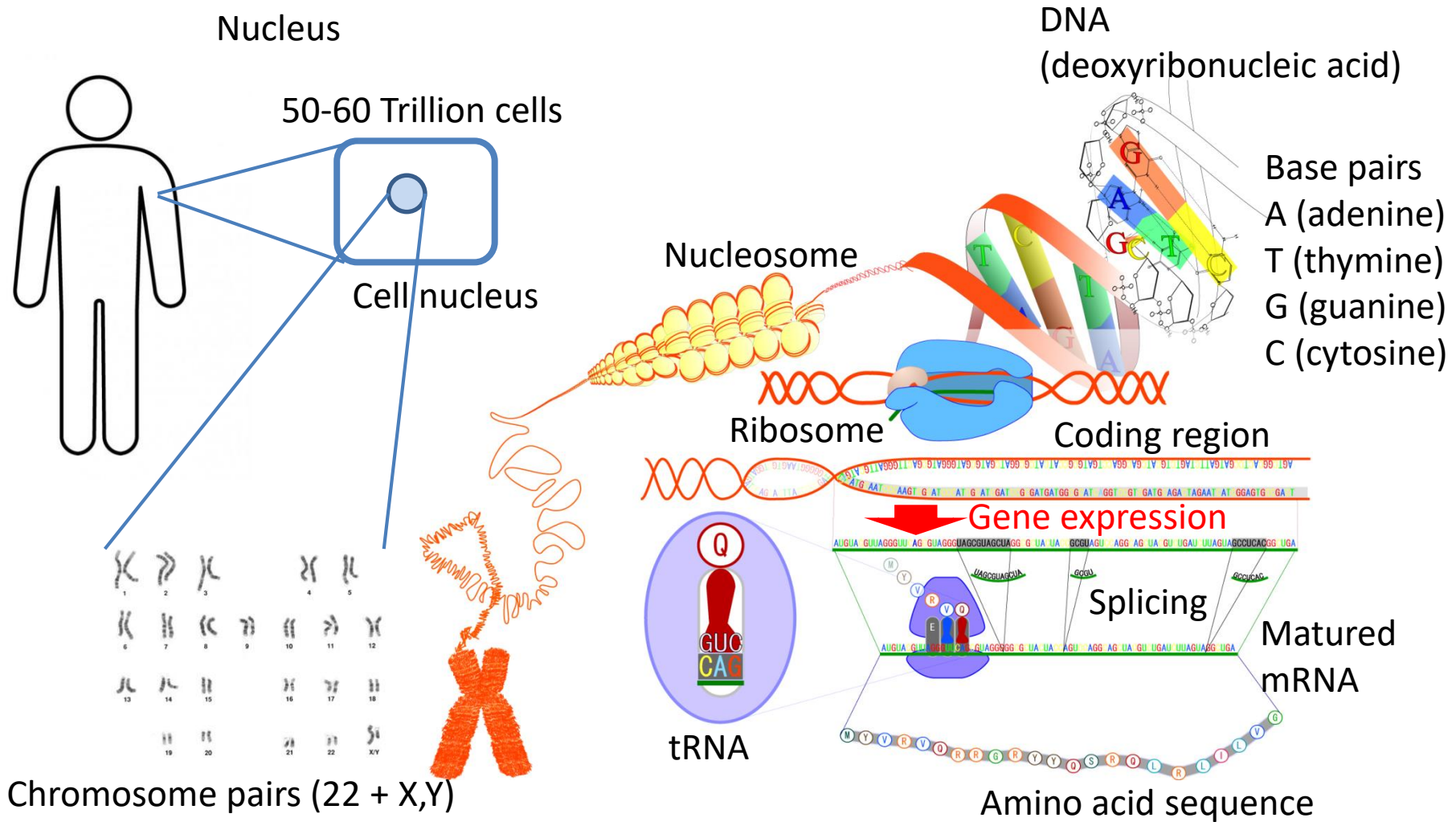
Gene Expression Regulatory Network Analysis

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Contents

- What is gene?
- What is gene expression?
- Gene expression regulatory network
- Complex Network Analysis
- Summary

What is gene?



Gene expression regulation

From DNA to Protein

<https://www.youtube.com/watch?v=gG7uCskUOrA> 2m:41s

Signal transduction => Gene expression

https://www.youtube.com/watch?v=vi-zWoobt_Q 3m:37s

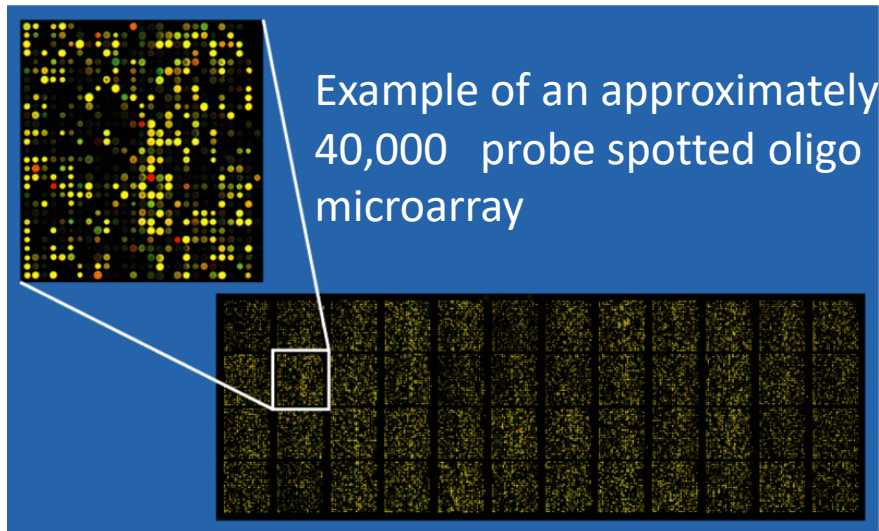
Transcription factors

<https://www.youtube.com/watch?v=MkUgkDLp2iE> 2m:55s

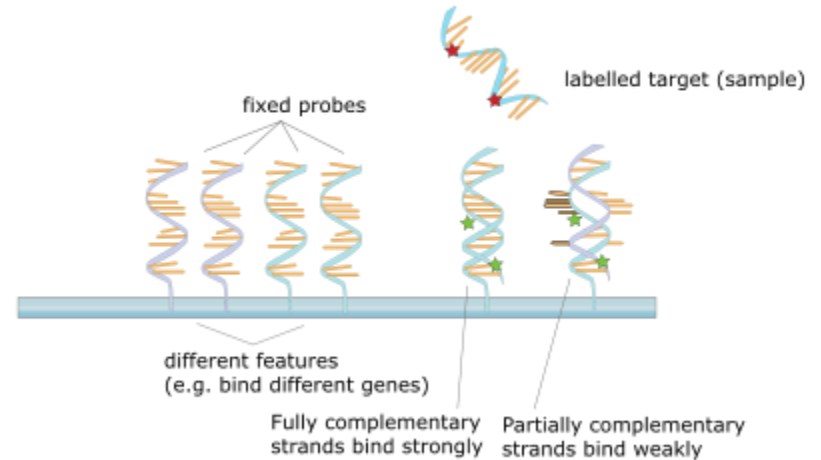
Issues

- Measurement of genome wide expression data
- Construction of gene expression regulatory network
- **Biologically meaningful interpretation** of the gene expression regulation network

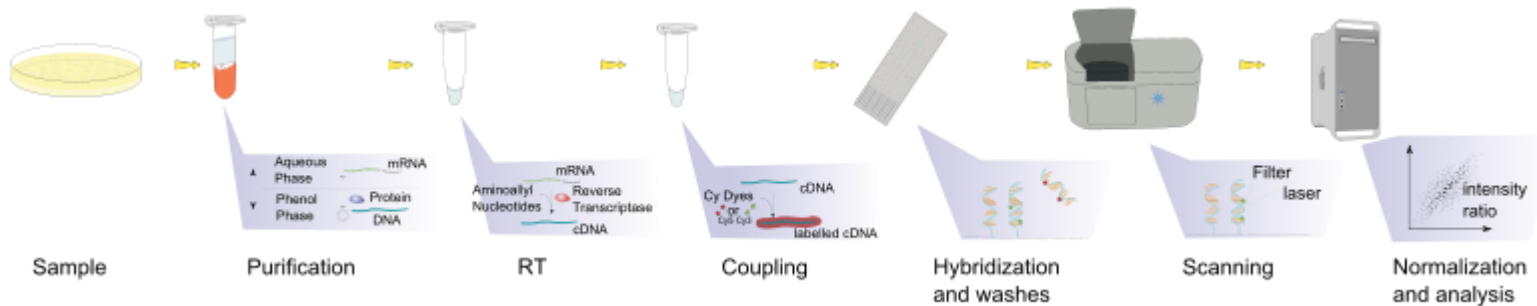
DNA Micro Array



DNA Hybridization

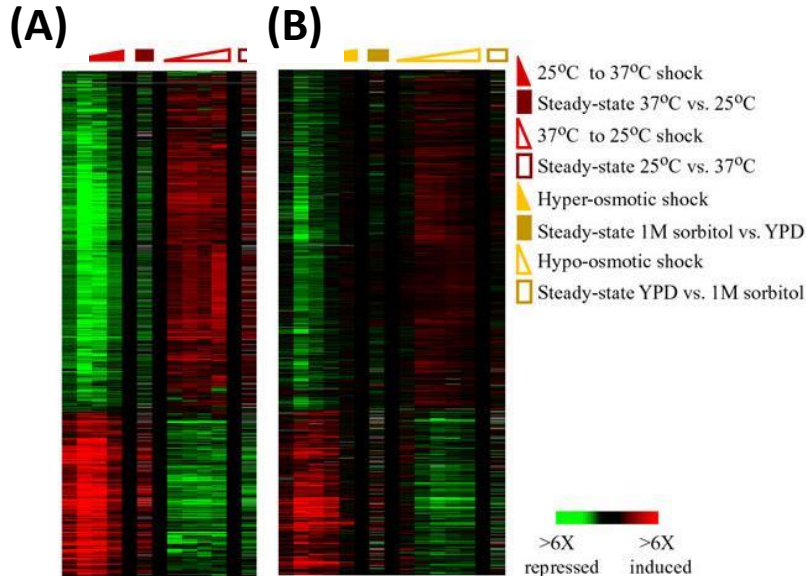


Experimental Protocol



http://en.wikipedia.org/wiki/DNA_microarray

Genome-wide Gene Expression Analysis



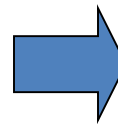
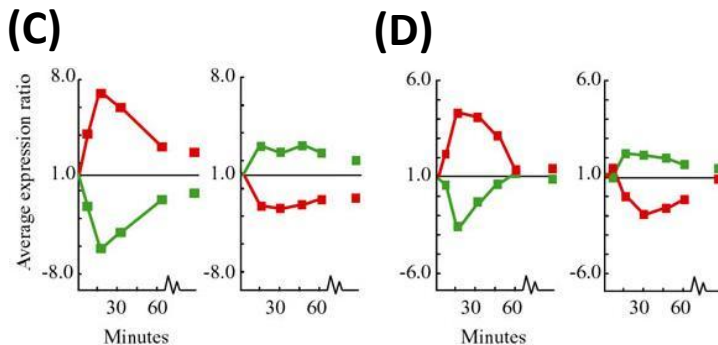
Gene Expression Regulation

(A) 25° C to 37° C
37° C to 25° C

(B) Low Osmotic Pressure \Rightarrow High
High Osmotic Pressure \Rightarrow Low

(C) Average Expression Levels in (A)

(D) Average Expression Levels in (B)

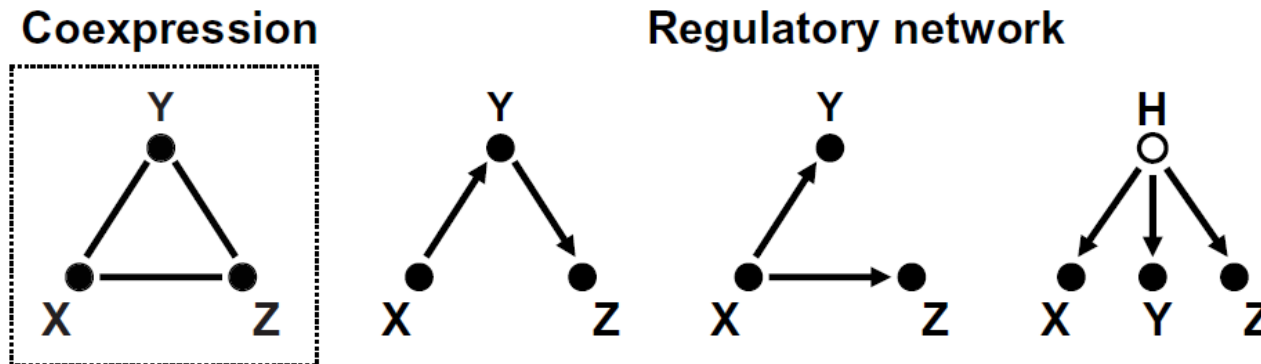


Clustering is useful to identify genes responsible for the stimuli

Gasch AP, Spellman PT,..., Botstein D, Brown PO : Genomic Expression Programs in the Response of Yeast Cells to Environmental Changes, Molecular Biology of the Cell, vol. 11, 4241-4257, (2000)

Limitation of Co-expression Analysis

Genes which have a similar expression profile may be regulated by the same transcription factors.



Issues: There are at least three regulation mechanisms even if X, Y and Z show the same expression profile.

Florian Markowetz: *BMC Bioinformatics* 2007, **8(Suppl 6):S5** doi:10.1186/1471-2105-8-S6-S5

Discovery of Driver Genes

How to identify driver genes such as

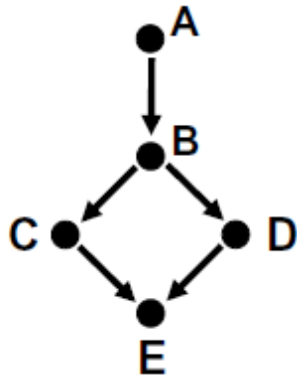
Yap1p (Transcription factor for oxidant stress)

Msn2p / Msn4p (Transcription factor for heat and osmotic stress)

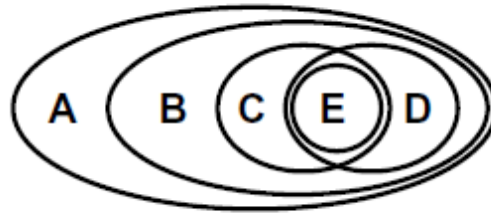
from Gene Expression Data

Knock out of **A**
affects to **B, C, D and E**

Pathway
structure



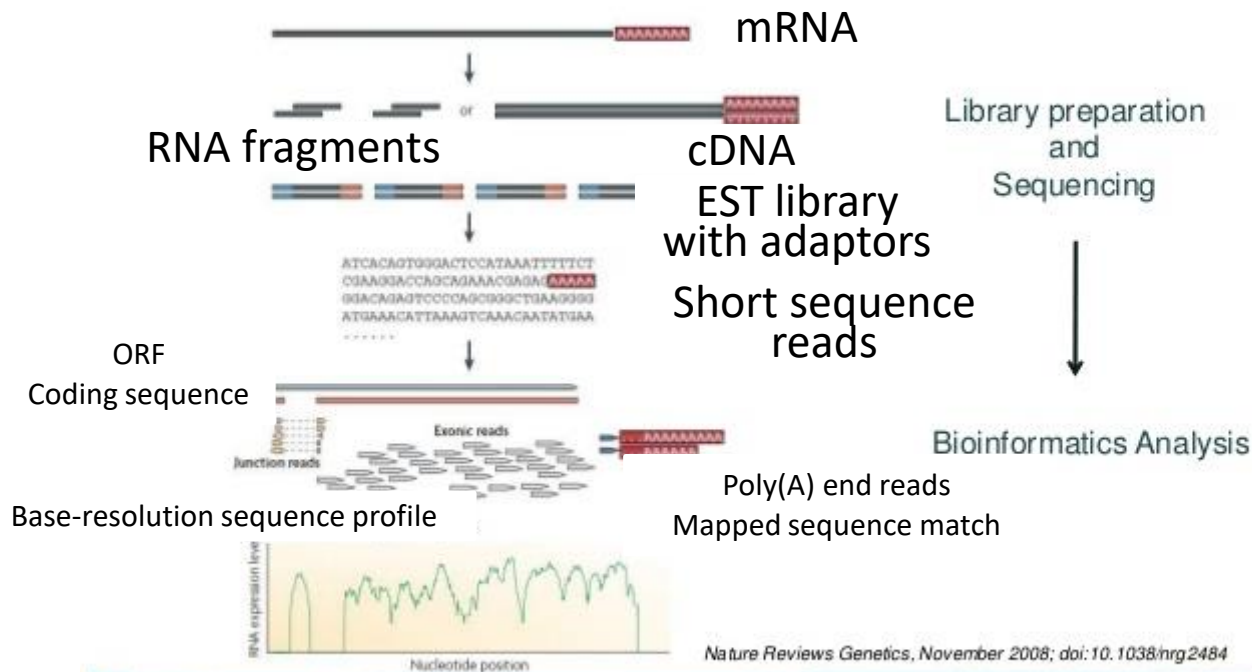
Downstream effects
of interventions



RNA Sequencing

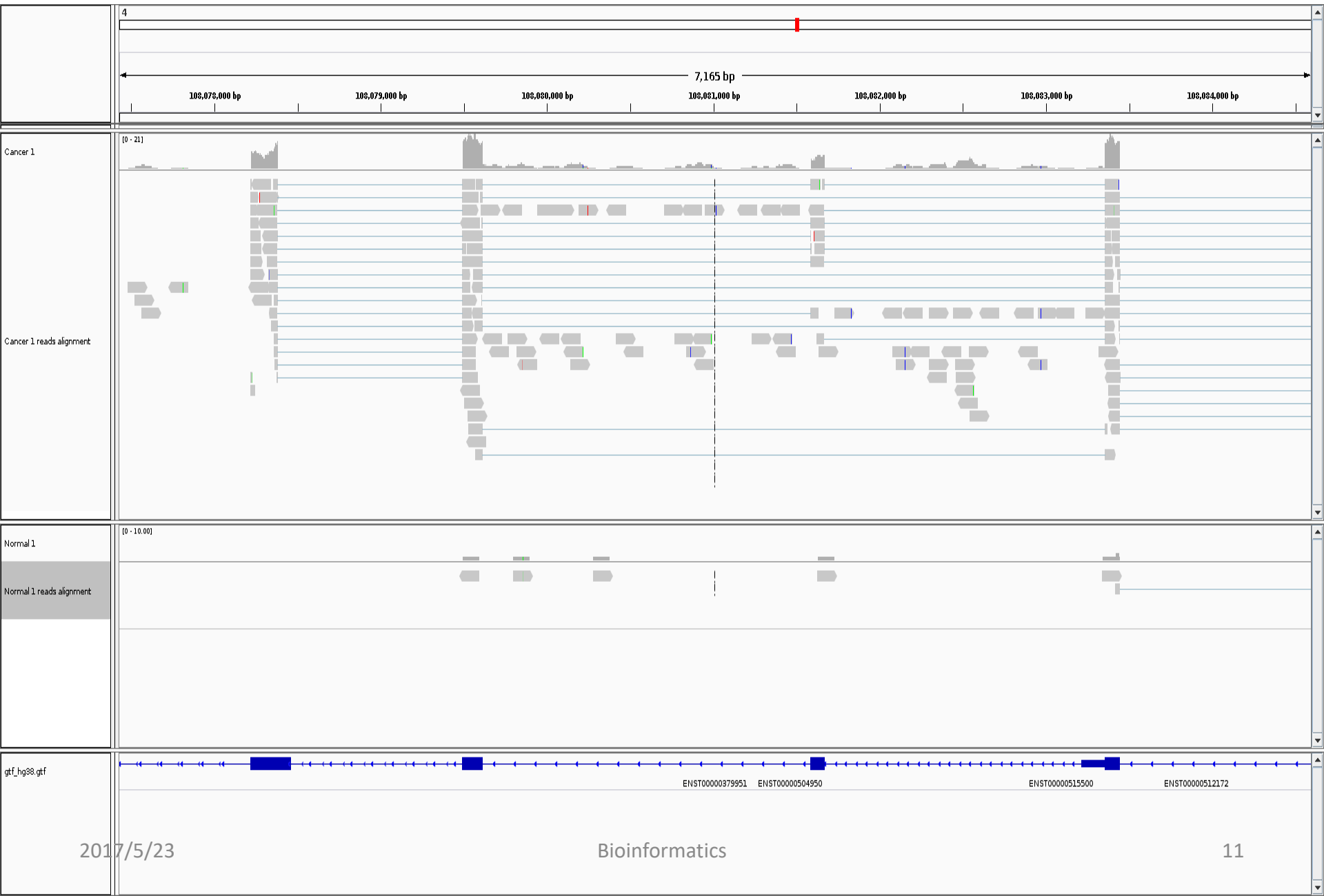
Ex) Next Generation Sequencer illumine NextSeq 550

A typical RNA-seq experiment

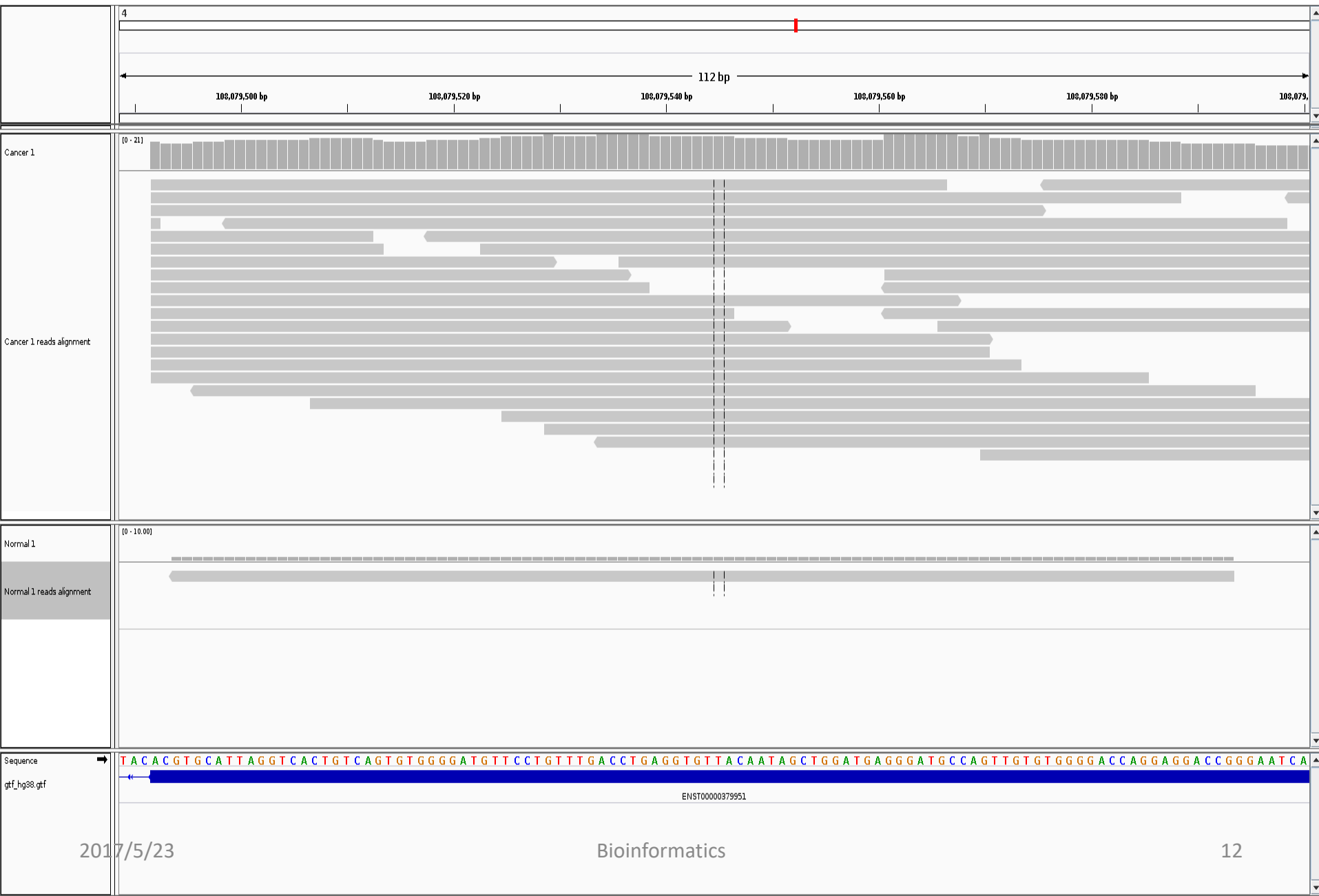


>10 M Reads (75bp)
40 samples
11 hours for sequencing

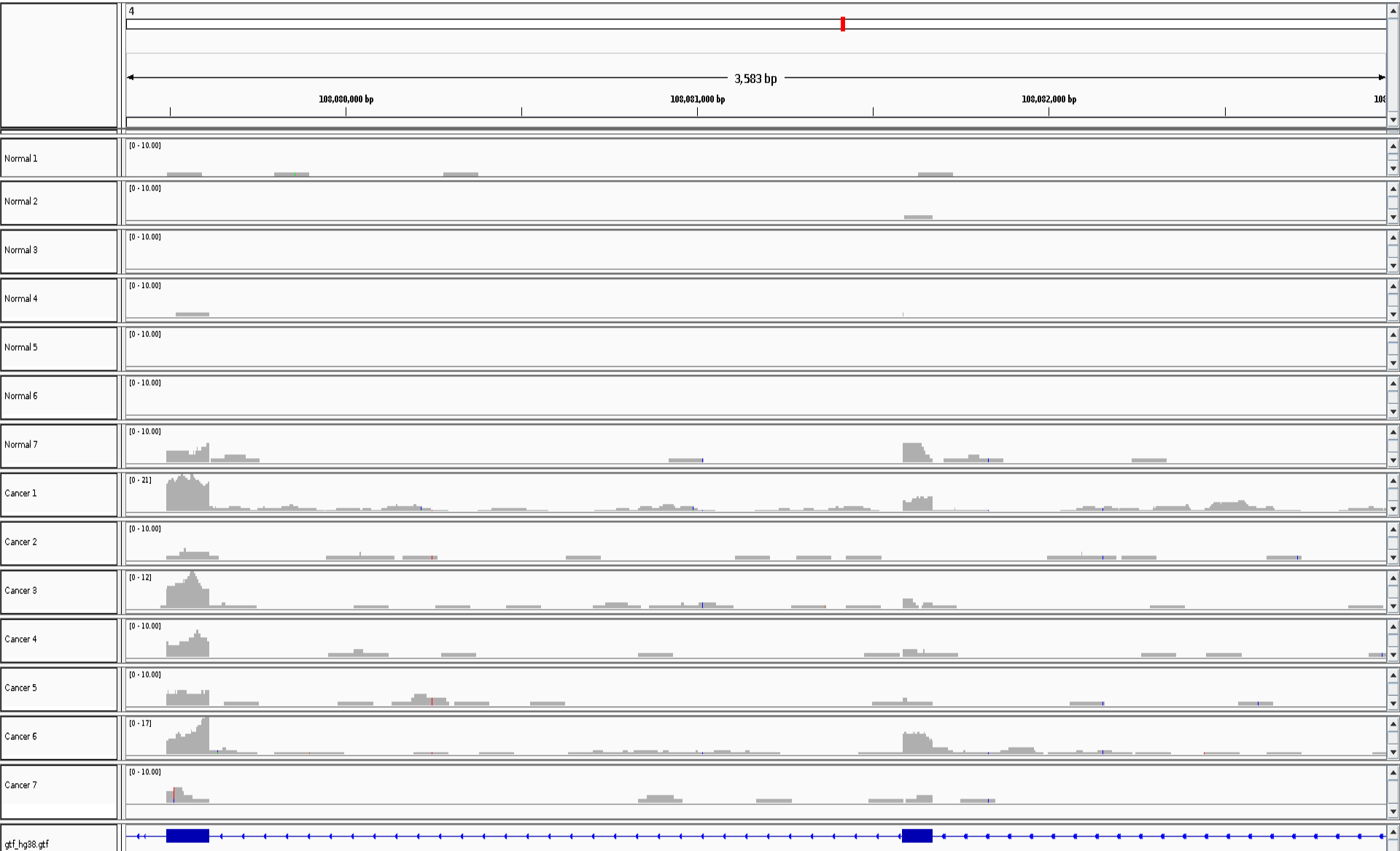
Visualization of reads alignment with genome reference



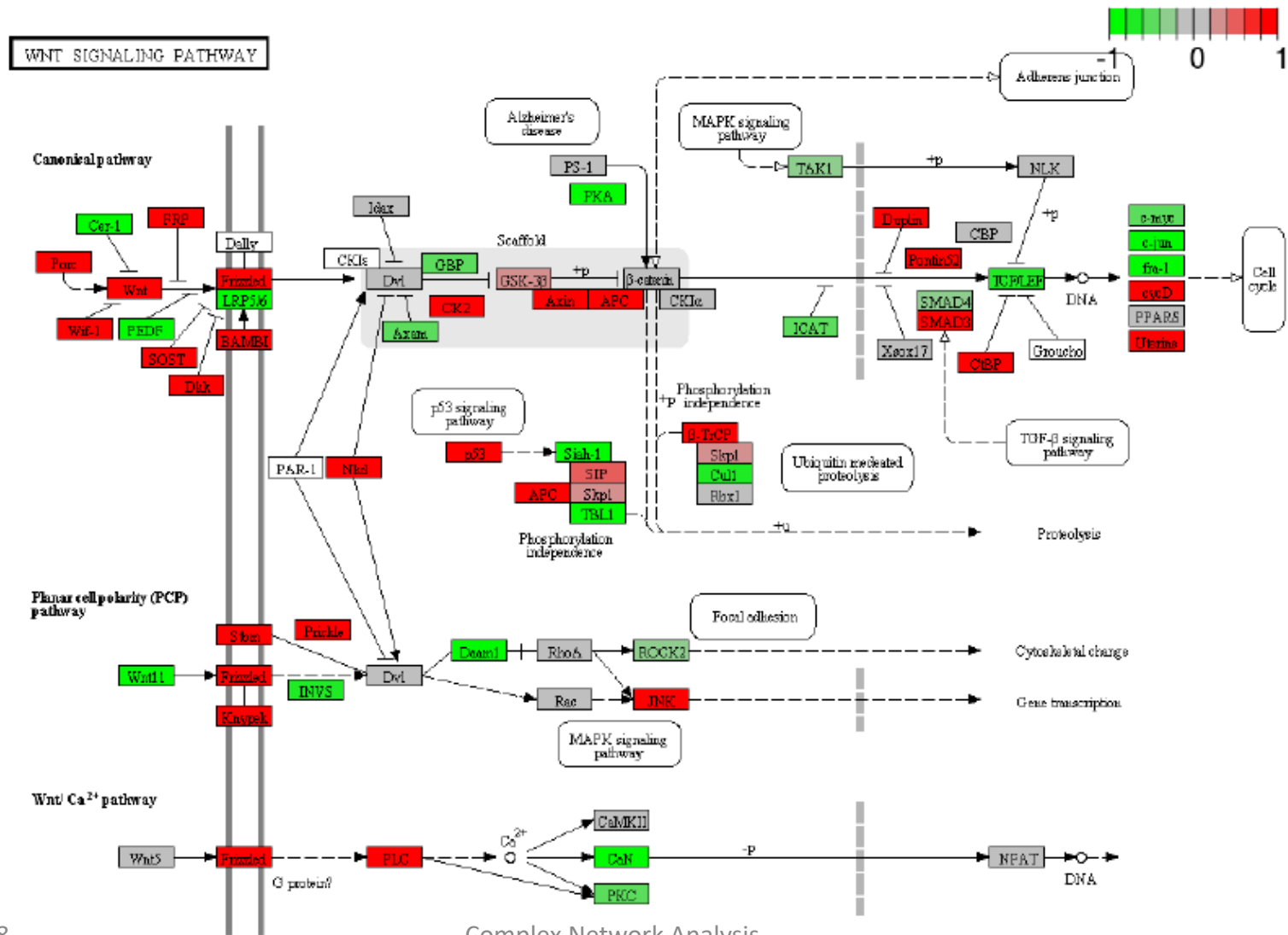
Visualization of reads alignment with genome reference (zoomed)



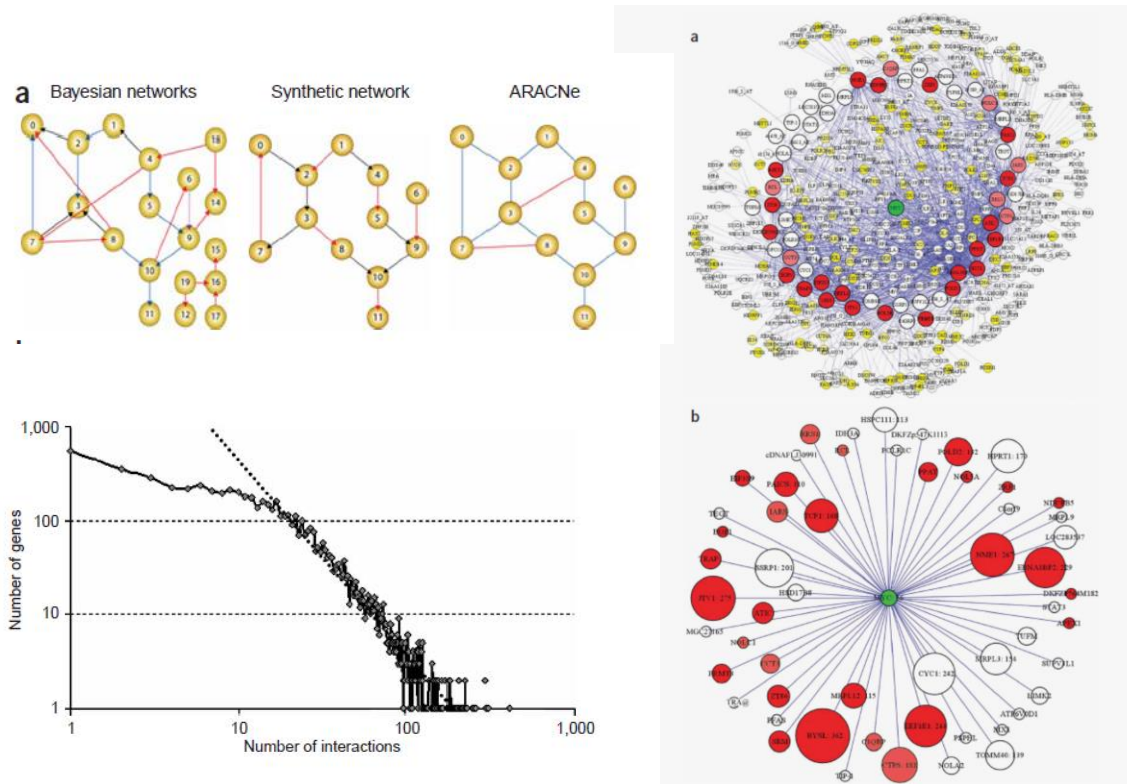
Alignment coverage of normal and cancer if gene LEF1 (upregulated in cancer)



Gene Regulatory Network of WNT Signaling pathway with RNA Seq. Data



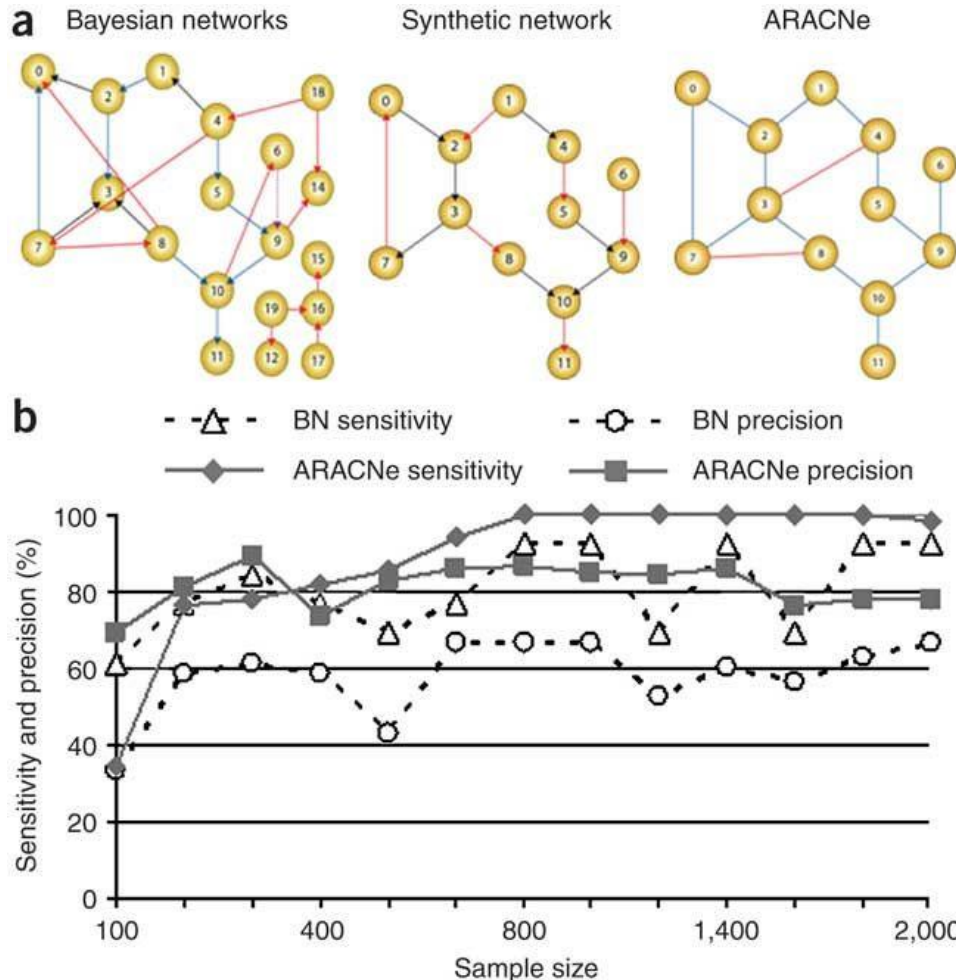
Reverse engineering of regulatory networks in human B cells



Identification of a master transcription regulation factor (MYC) with ARACNe (algorithm for the reconstruction of accurate cellular networks) which uses mutual information to eliminate indirect interactions.

Basso K, Margolin AA, ..., Dalla-Favera R, Califano A: Reverse Engineering of Regulatory Networks in Human B Cells, Nature Genetics, vol.37, no.4, pp.382-389 (2005)

Comparison between Bayesian network and ARACNe on the synthetic network



Bayesian network:

Blue arrows: correct detection
Red arrows: incorrect detection
Black arrows: correct but opposite direction

Synthetic network:

Black arrows: up regulation
Red arrows: down regulation

ARACNe:

Blue lines: correct detection
Red lines: incorrect detection

Gene simulator

$$Y_{t+1} = f(Y_t) = A(Y_t - T) + \varepsilon$$

Good for test data generation and
evaluation of network extraction algorithms

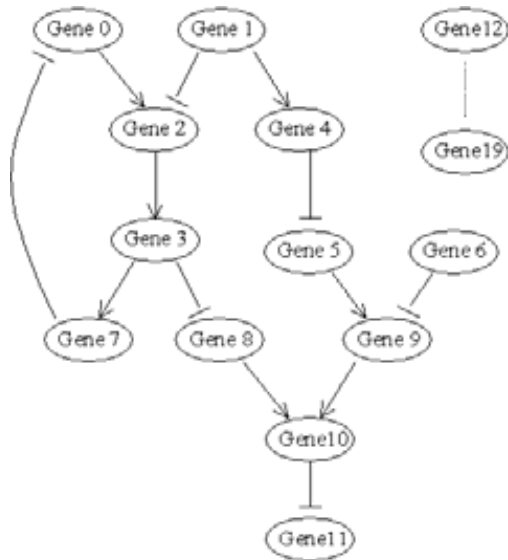
Y_t : gene expression level at t (from 0 to 100)

A : a matrix for gene interaction

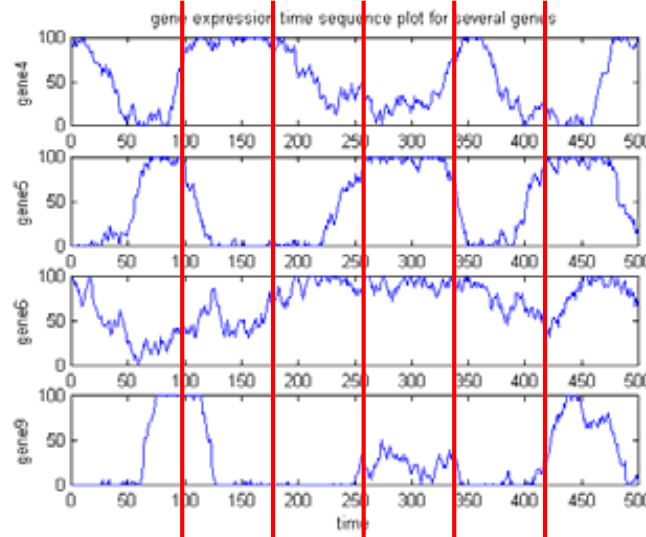
T : a vector of threshold regulating values (50)

ε : white noise (-10 to 10)

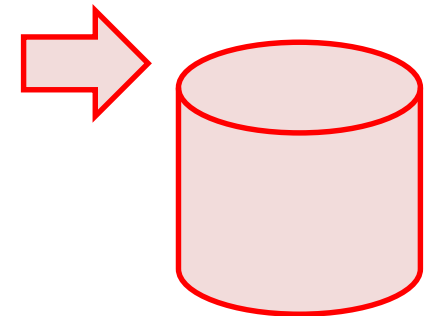
Given gene network



Generated gene expression



Generated
Time-course
Gene expression
samples

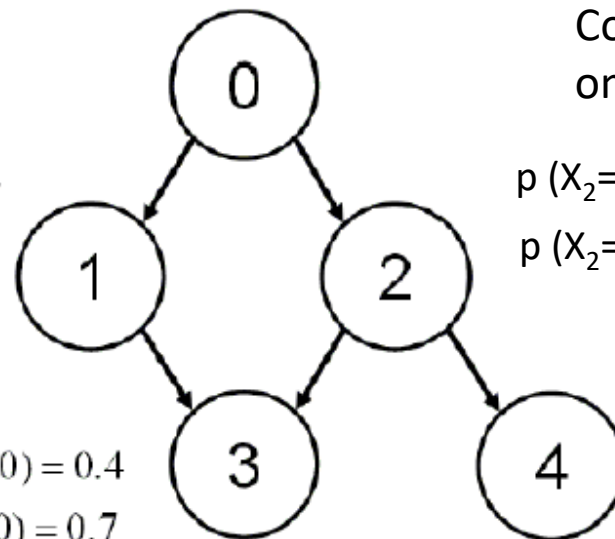


Jing Yu, et al: Using Bayesian Network Inference Algorithms
to Recover Molecular Genetic Regulatory Networks, in 3rd International
Conference on Systems Biology, Sweden (2002)

Bayesian Network

$$p(x_1, x_2, \dots, x_N) = \prod_{i=1}^N p(x_i | x_1, x_2, \dots, x_{i-1}) \quad \text{Good for node dependencies}$$

$$p(x_0 = 1) = 0.5$$



Conditional probability of node 2
only depend on the state of node 1.

$$p(x_1 = 1 | x_0 = 0) = 0.2$$

$$p(x_1 = 1 | x_0 = 1) = 0.6$$

$$p(X_2 = 1 | x_0 = 0) = 0.5$$

$$p(X_2 = 1 | x_0 = 1) = 0.2$$

Conditional probability of node 4
only depend on the state of node 2.

$$p(x_3 = 1 | x_1 = 0, x_2 = 0) = 0.4$$

$$p(x_3 = 1 | x_1 = 1, x_2 = 0) = 0.7$$

$$p(x_3 = 1 | x_1 = 0, x_2 = 1) = 0.2$$

$$p(x_3 = 1 | x_1 = 1, x_2 = 1) = 0.4$$

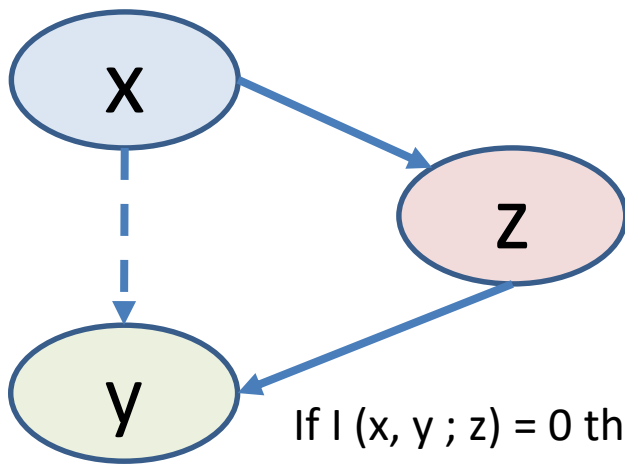
$$p(x_4 = 1 | x_2 = 0) = 0.7$$

$$p(x_4 = 1 | x_2 = 1) = 0.4$$

Conditional probability of node 3
only depend on the state of node 1
and node 2.

ARACNe : Improves gene network considering “indirect interaction” between genes

Two genes, x and y , as indirectly interacting through a third gene, z , if the conditional mutual information $I(x, y | z)$ is equal to zero.



If $I(x, y ; z) = 0$ there is an indirect link between X and Y

$$\begin{aligned} I(X; Y) &= H(X) - H(X|Y) \\ &= H(Y) - H(Y|X) \\ &= H(X) + H(Y) - H(X, Y) \end{aligned}$$

$H(X)$: entropy of X ; ambiguity of X

$H(X|Y)$: conditional entropy of X of Y ;
ambiguity of X when Y is available

X, Y : independent

$$\Leftrightarrow H(X|Y) = H(X)$$

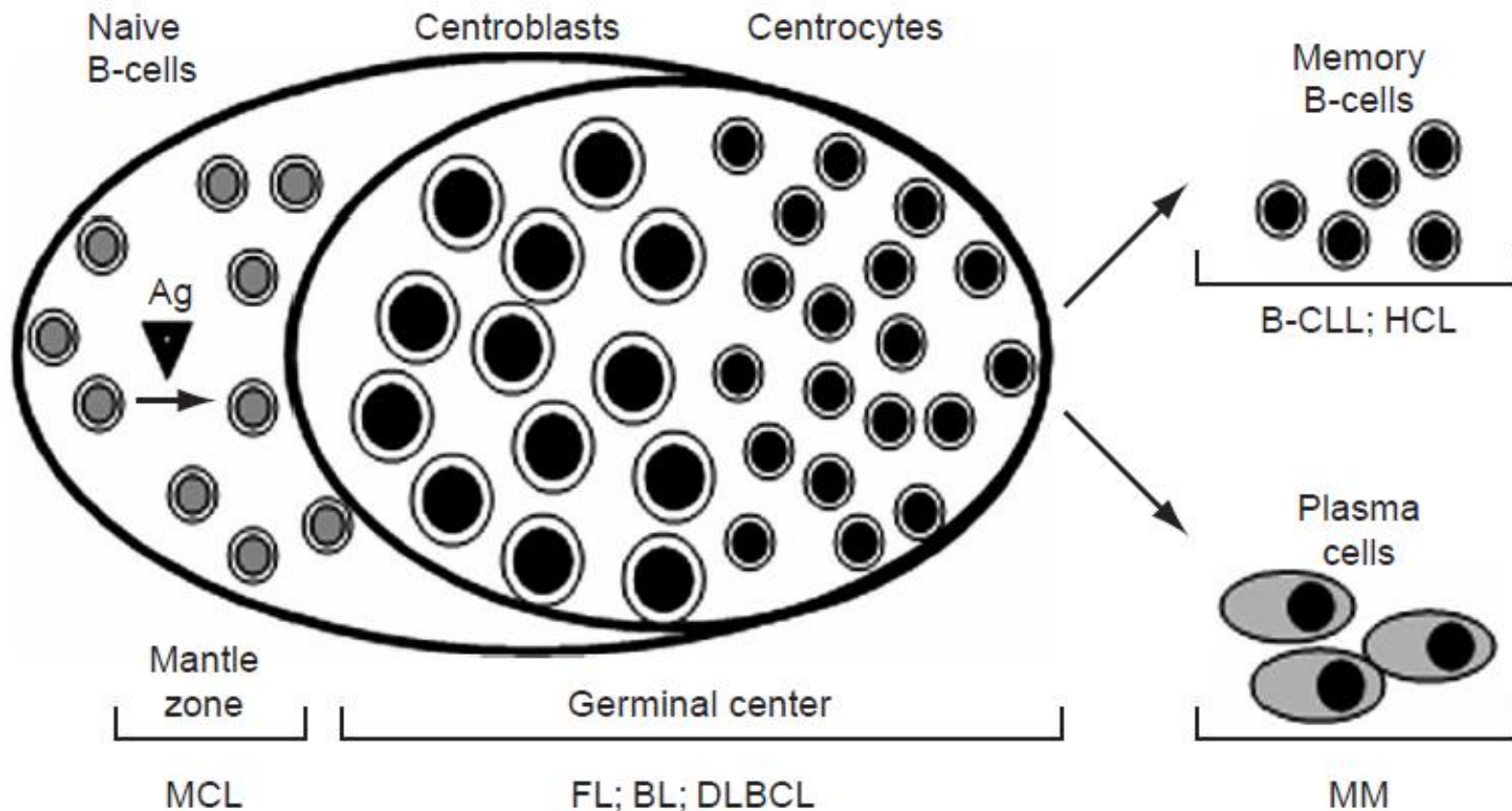
$$\Leftrightarrow I(X; Y) = 0$$

Ambiguity of X does not decrease even if we know Y

B-cell lymphoma (B細胞リンパ腫)

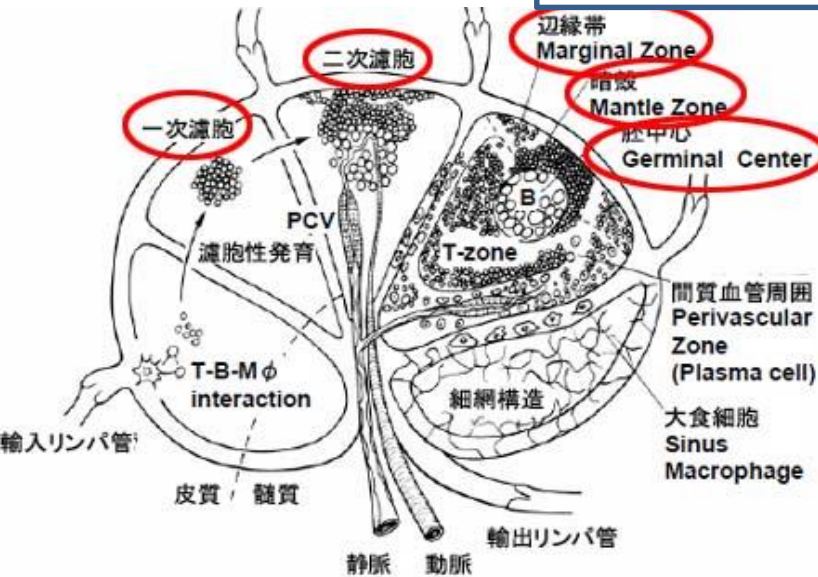
<https://www.youtube.com/watch?v=Y-HaTwHdSM4>

(1m 37s)



Data set

Normal B cell
正常B細胞



Cancer B cell
癌B細胞

Genome-wide
expression profiles
from a panel of 336 B
cell phenotypes to
produce antibodies
with high affinity

Supplementary Table 1. List of samples included in the gene expression profiles database

Sample code	Description
Normal B cells	
CB 2-23	purified CB
CB 3-10	purified CB
CB 3-30	purified CB
CB 3-7	purified CB
CB 6-8	purified CB
CC 3-28	purified CC
CC 4-14	purified CC
CC 4-6	purified CC
CC 7-25	purified CC
CC 7-7	purified CC
M 4-12	purified M
M 4-14	purified M
M 4-26	purified M
M 5-2	purified M
M 6-8	purified M
N 4-13	purified N
N 4-14	purified N
N 4-7	purified N
N1 4-21	purified N
N2 4-21	purified N
Cord 26-2	purified B cells from Cord blood
Cord 511	purified B cells from Cord blood
Cord 512	purified B cells from Cord blood
Cord 1171	purified B cells from Cord blood
Cord 1193	purified B cells from Cord blood
Primary tumor samples and cell lines	
CLL1269	B-CLL unmutated
CLL1275	B-CLL unmutated
CLL1282	B-CLL unmutated
CLL1318	B-CLL unmutated
CLL1310	B-CLL unmutated
CLL4	B-CLL unmutated
CLL14	B-CLL unmutated
CLL1304	B-CLL unmutated
P-CLL1268	purified B-CLL unmutated
P-CLL1285	purified B-CLL unmutated
P-CLL1302	purified B-CLL unmutated
P-CLLB3	purified B-CLL unmutated
P-CLLB4	purified B-CLL unmutated
P-CLLB7	purified B-CLL unmutated
P-CLLB14	purified B-CLL unmutated
P-CLLB15	purified B-CLL unmutated
CLL1208	B-CLL mutated
CLL1255	B-CLL mutated
CLL1259	B-CLL mutated
CLL1333	B-CLL mutated
CLL1270	B-CLL mutated

Centroblasts
胚中心細胞

Centrocites
中心細胞

Mantle zone
マントル細胞

Naive B cell
抗原未認識B細胞

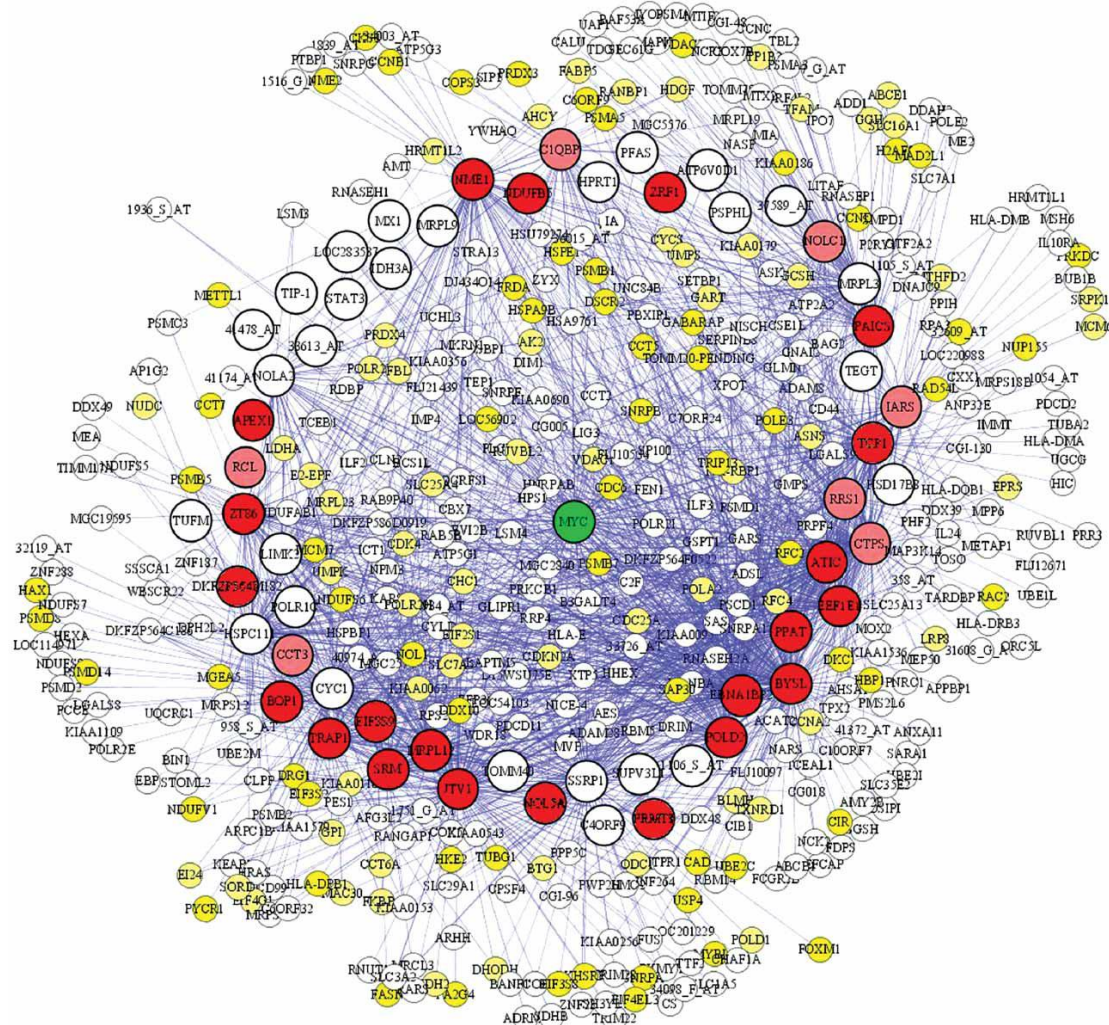
Cord blood
へその緒

B cell unmutated
非変異B細胞

Purified B cell
unmutated
精製非変異B細胞

B cell mutated
変異B細胞

A MYC-specific subnetwork was obtained by including all the genes that have $P < 10^{-7}$



56 genes directly connected
(first neighbors; red, pink, white)

444 most statistically
significant genes
(second neighbors; yellow)

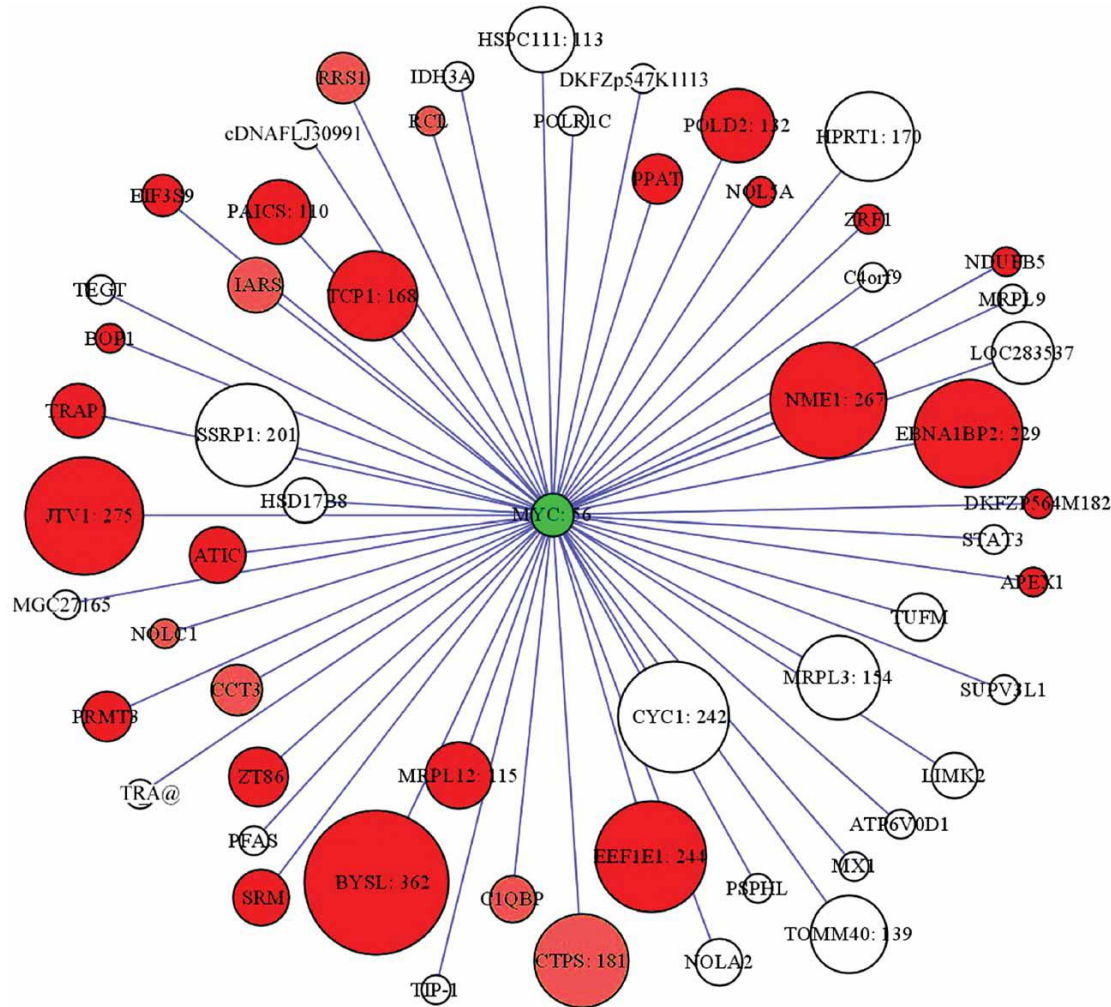
Green node: MYC

Red node: ChIP seq data available

Pink node: not available

White nodes: no information available about MYC relationship

Hub of hubs

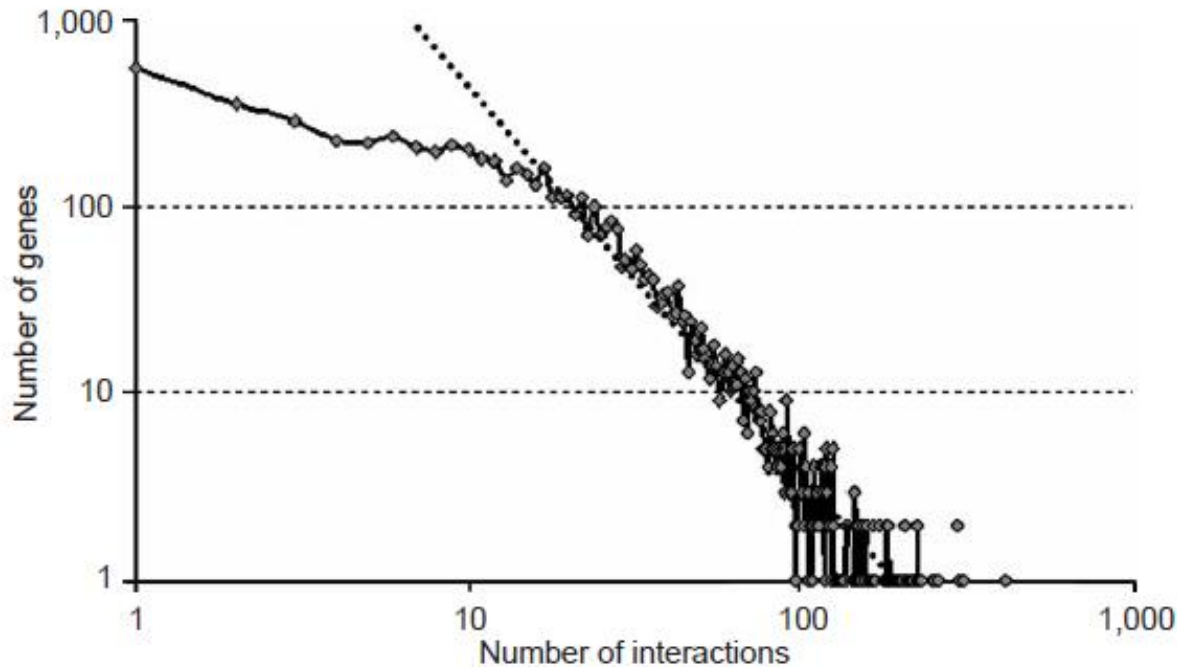


MYC plays a role of hub of hubs which regulates more than 2000 genes in total.

The sizes of circles are proportional to the number of connections.

Power law in gene expression regulatory network

Log-log scale plot with regards to the number of genes and interactions



Summary

- Gene expression is regulated by complex network.
- Estimation of causality is more important than finding co-expressions.
- Bayesian Network and mutual information are effective to extract gene expression regulatory networks.
- Complex network analysis is also applicable to gene expression regulatory networks.

Report Issues

- Due date Jan 25th (Thu) Noon
- Submission through OCW
- Subject Protein-protein interaction network
- Style a PDF file A4 1 or 2 pages

Issue

- Impression or investigation report on the issue of gene expression regulatory network held on Jan. 18th.