

functional genomics

Astrid Lægreid

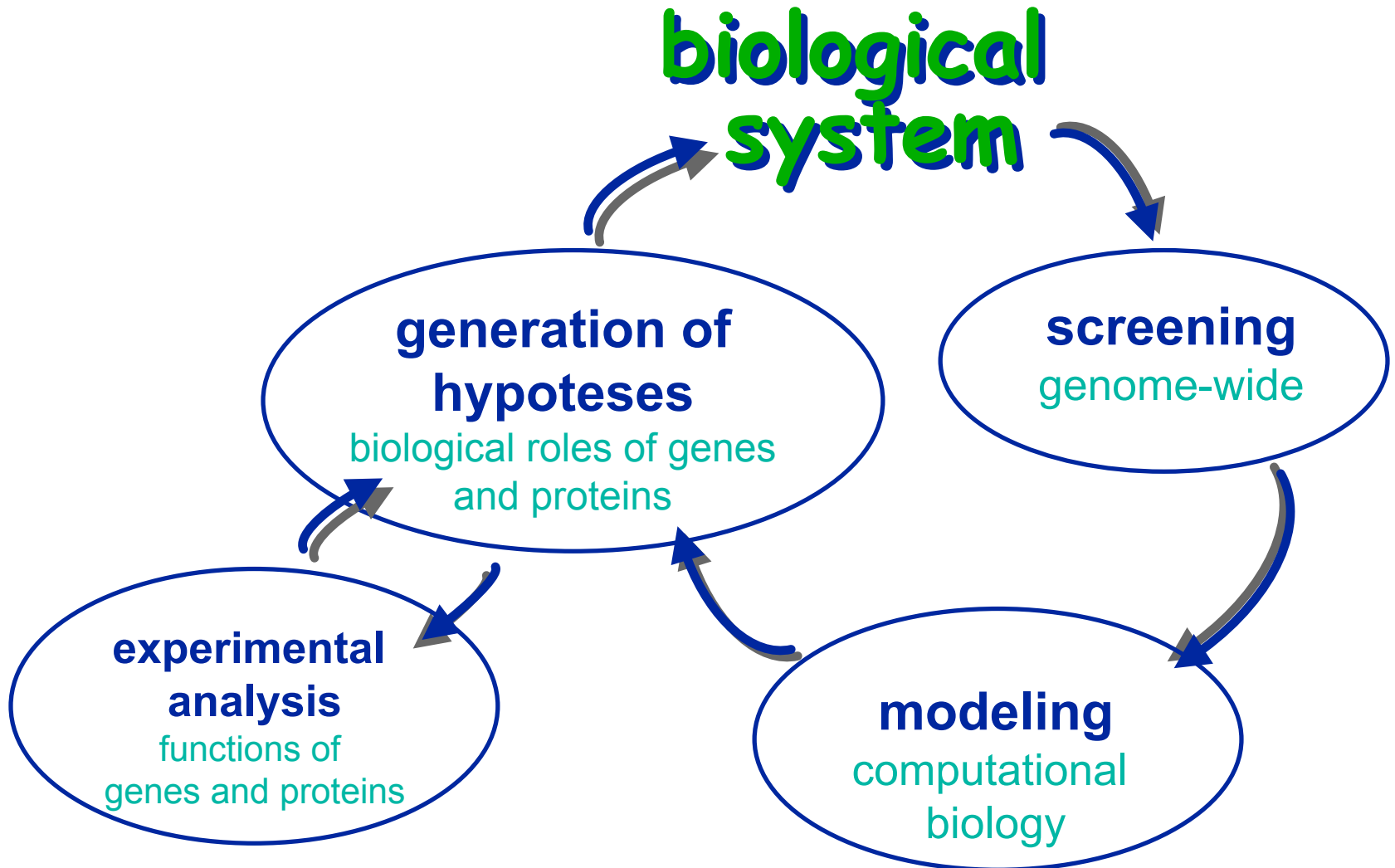
Department of Cancer Research and Molecular Medicine
Norwegian University of Science and Technology

Microarray Core Facility
Norwegian Microarray Consortium

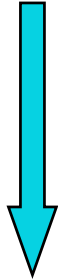
outline

- ▶ functional genomics
- ▶ gene expression
- ▶ predicting gene function
- ▶ challenges

functional genomics



genome sequencing



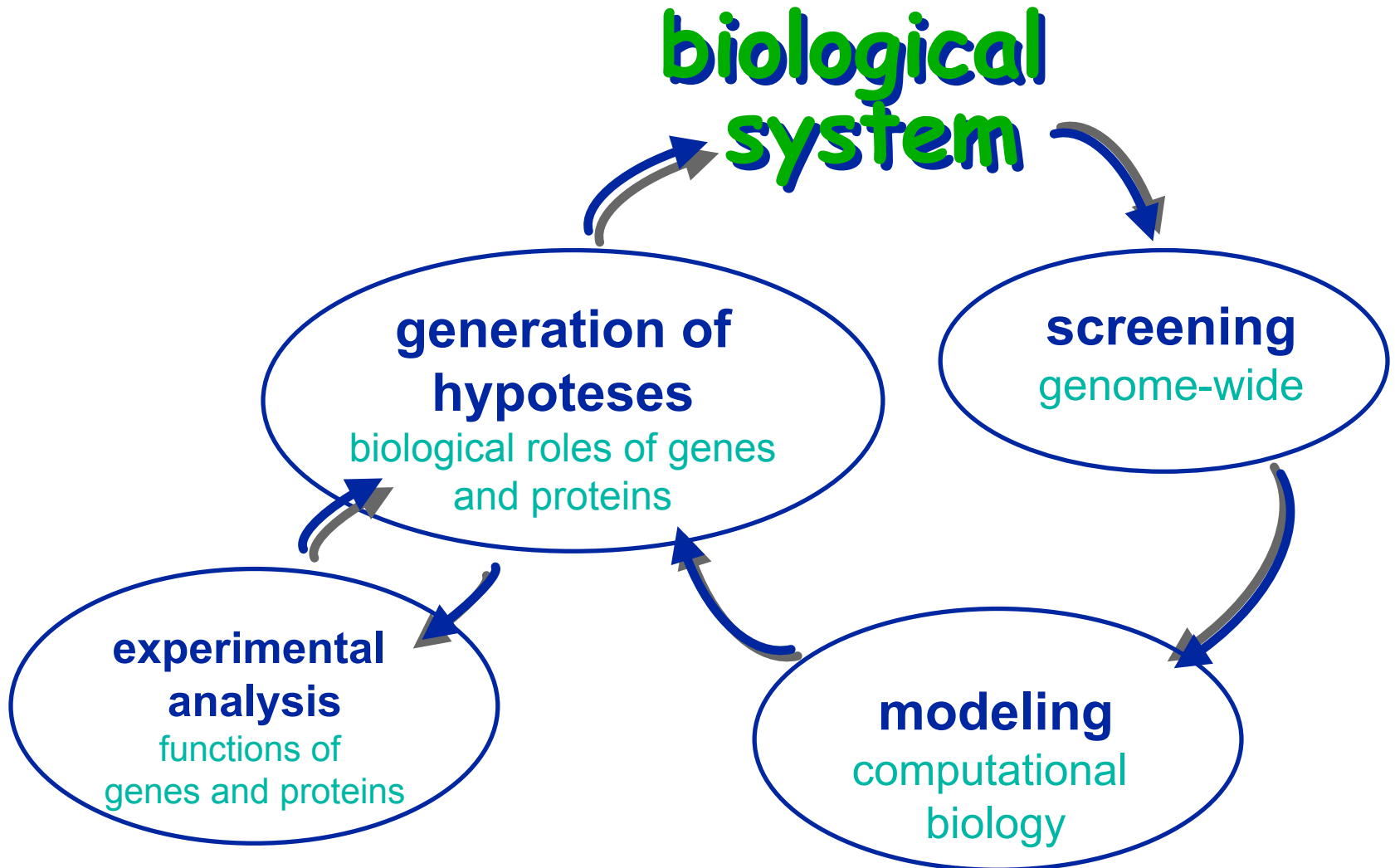
human genome

- ▶ 3×10^9 basepairs
- ▶ ~ 35.000 genes
- ▶ > 100.000 splice variants

genome-wide screening

- ▶ how? high-throughput - HTP
- ▶ what? gene expression, gene-dosage, gene-variation (SNP), protein
- ▶ with? microarray, mass spectrometry, 2D-gel electrophoresis









functional genomics



chromosome 21

- ▶ 127 known genes
- ▶ 98 unknown genes

50 million bases

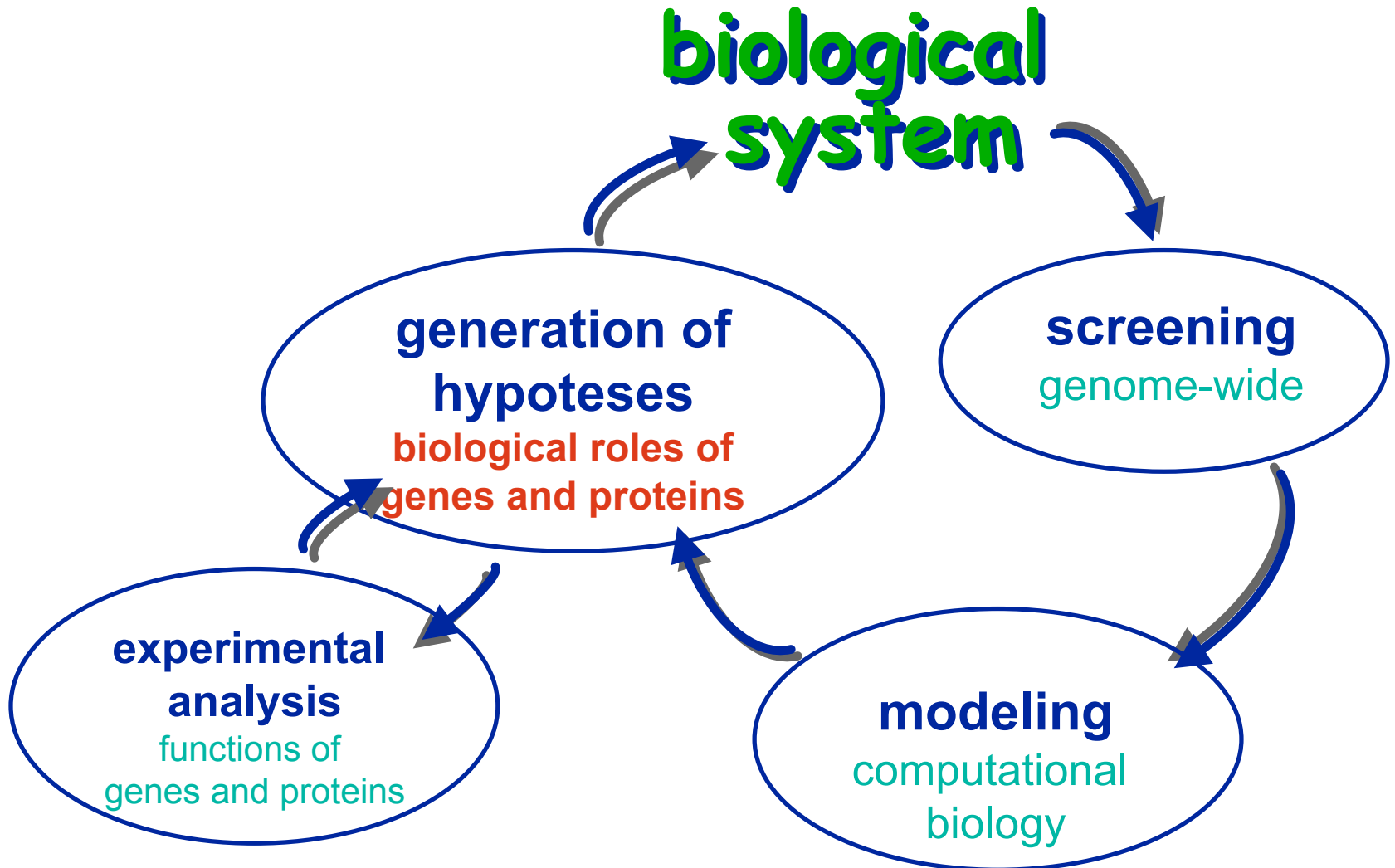
Coxsackie and adenovirus receptor		Myeloproliferative syndrome, transient
Amyloidosis, cerebroarterial, Dutch type		Leukemia, transient, of Down syndrome
Alzheimer disease, APP-related		Enterokinase deficiency
Schizophrenia, chronic		Multiple carboxylase deficiency
Usher syndrome, autosomal recessive		T-cell lymphoma invasion and metastasis
Amyotrophic lateral sclerosis		Mycobacterial infection, atypical
Oligomycin sensitivity		Down syndrome (critical region)
Jervell and Lange-Nielsen syndrome		Autoimmune polyglandular disease, type I
Long QT syndrome		Bethlem myopathy
Down syndrome cell adhesion molecule		Epilepsy, progressive myoclonic
Homocystinuria		Holoprosencephaly, alobar
Cataract, congenital, autosomal dominant		Knobloch syndrome
Deafness, autosomal recessive		Hemolytic anemia
Myxovirus (influenza) resistance		Breast cancer
Leukemia, acute myeloid		Platelet disorder, with myeloid malignancy

164 million bases

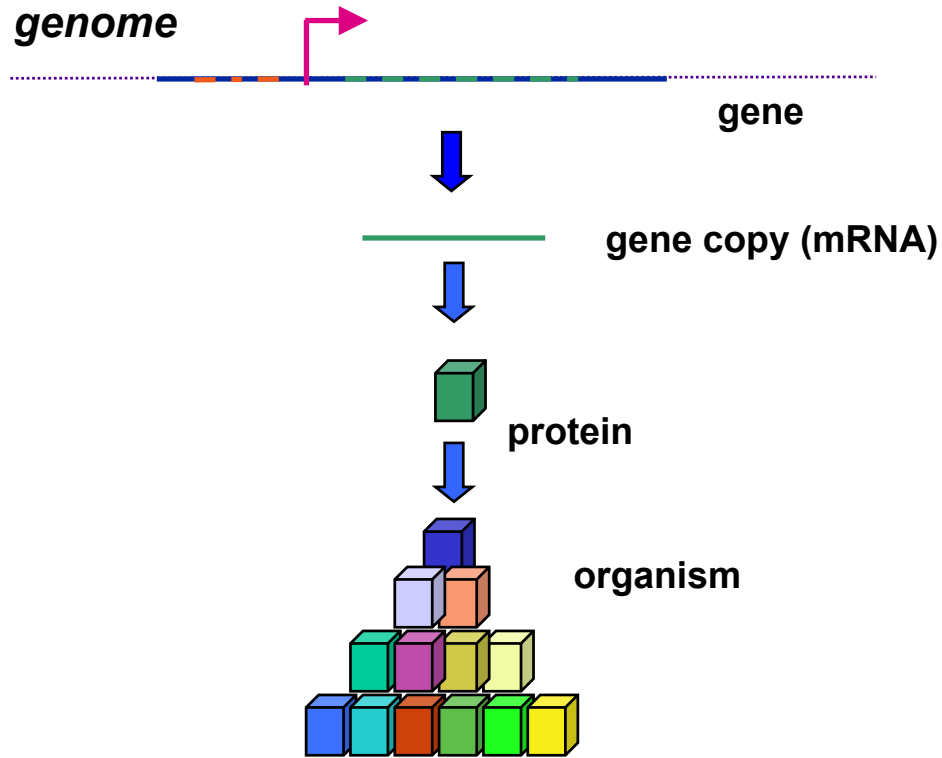


- | | |
|--|--|
| Short stature, idiopathic lamellar | Hodgkin disease susceptibility, pseudotubercular |
| Ure-Weil dysphosphatasia | leukemia |
| Large mesencephalic dysplasia | Mucopolysaccharidosis, dermal adasia, and sclerocornea |
| Leukemia, acute myeloid, M2 type | Basophil muscle weakness |
| Chondrodysplasia punctata | Mental retardation |
| Kallmann syndrome | Ocular albinism and sensorial deafness |
| Ocular albinism, Neri-Eshel-Falls type | Amelogenesis imperfecta |
| Oral-facial-digital syndrome | Charco-Marie-Tooth disease, axonal |
| Nance-Horan cleft-tooth syndrome | Chondro-sarcoma, spindle, scleroma |
| Hereditary persistence of fetal hemoglobin | Hypochosphatemia, hereditary |
| Pyruvate dehydrogenase deficiency | Perington syndrome |
| Glycogen storage disease | Retinoblastoma |
| Coffin-Lovsy syndrome | Gonadal dysgenesis, XY female type |
| Mental retardation | Mental retardation, non-syndromic |
| Spondyloepiphyseal dysplasia tarda | Agammaglobulinemia, type 2 |
| Perinatal nocturnal hemoglobinuria | Cardiofacioskeletal dysplasia |
| Infantile spasm syndrome | Opitz G syndrome, type 1 |
| Alcarrá syndrome | Pigment disorder, reticulate |
| Deafness, sensorineural | Melanoma |
| Simpson-Golabi-Beher syndrome, type 2 | Duchenne muscular dystrophy |
| Adrenal hypoplasia, congenital | Becker muscular dystrophy |
| Dosage-sensitive sex reversal | Cardiomyopathy, dilated |
| Deafness, congenital sensorineural | Chronic granulomatous disease |
| Retinitis pigmentosa | Snyder-Robinson mental retardation |
| Wilson-Kume syndrome | Narcolepsy |
| Cone dystrophy | Exudative vitreoretinopathy |
| Alard island eye disease (ocular albinism) | Cystic disease |
| Optic atrophy | Remington syndrome |
| Night blindness, congenital stationary, type 1 | Retinitis pigmentosa, recessive |
| Cytokine-inhibiting activity | Mental retardation, nonspecific and syndromic |
| Achrocytosis multiplex congenital | Dyserythropoietic anemia with thrombocytopenia |
| Night blindness, congenital stationary, type 2 | Chondrodysplasia punctata, dominant |
| Brunner syndrome | Autoimmunity-immunodeficiency syndrome |
| Wilson-Aldrich syndrome | Renal cell carcinoma, papillary |
| Thrombocytopenia | Facioscapular dysplasia (Aarskog Scott syndrome) |
| Dent disease | Chachaciney syndrome with mental retardation |
| Nephrolithiasis, type I | Sarcosis, synovial |
| Hypochosphatemia, type III | Pruvo syndrome |
| Proteinuria | Spinal muscular atrophy, lethal infantile |
| Anemia, sideroblastic/hypochromic | Migraine, familial typical |
| Cerebellar ataxia | Androgen insensitivity |
| Renal cell carcinoma, papillary | Spinal and tubular muscular atrophy |
| Diabetes mellitus, insulin-dependent | Prostate cancer |
| Schwartz-Haas syndrome | Perineal hypospadias |
| Cognitive function, social | Breast cancer, male, with Reiterstein syndrome |
| Mental retardation, nonspecific | Endometrial dysplasia, sericiferous |
| Menkes disease | Alpha-thalassemia/mental retardation |
| Occipital horn syndrome | Luberg-Marsh syndrome |
| Cleft lip, neonatal | Schwartz-Haas syndrome |
| FG syndrome | Smith-Fineman-Myers syndrome |
| Immunodeficiency, moderate and severe | Hemolytic anemia |
| Miles-Campaner syndrome | Myoglobinuria/hemolysis |
| Charco-Marie-Tooth neuropathy, dominant | Wiedemann-Rautavaara syndrome |
| Mental retardation | Torsion dystonia-parkinsonism, Filipino type |
| X-inactivation center | Leukemia, myeloid/lymphoid or mixed lineage |
| Premature ovarian failure | Anemia, sideroblastic, with ataxia |
| Arts syndrome | Albinism-Deafness syndrome |
| Cleft palate and/or arylglucosidase | Deafness |
| Megalocornea | Choroideremia |
| Epilepsy (Luberg-Hellman syndrome) | Agammaglobulinemia |
| Pellaeus-Merzbacher disease | Fabry disease |
| Spastic paraplegia | Mohr-Taraberg syndrome |
| Alport syndrome | Jensen syndrome |
| Cowchock syndrome | Lisencephaly |
| Hypertichosis, congenital generalized | Boxer syndrome |
| Prax, hereditary congenital | Mental retardation with growth hormone deficiency |
| Apoptosis inhibitor | Mental retardation, South African type |
| Perihypocytosis | Lymphocystin-like syndrome |
| Thoracoabdominal syndrome | X-inactivation, familial skewed |
| Simpson-Golabi-Beher syndrome, type 1 | Periglaug syndrome |
| Split hand/foot malformation, type 2 | Gustavson mental retardation syndrome |
| Hypoparathyroidism | Immunodeficiency, with hyper IgM |
| Mental retardation, Shafiq type | Retinitis pigmentosa |
| Lech-Nyman syndrome | SRI (on determining region 1) |
| HRP-related gauz | Wood neuroimmunologic syndrome |
| Loose syndrome | Hereditary, visceral |
| Bojerson-Fassman-Uehmann syndrome | Albinism-deafness syndrome |
| Testicular germ cell tumor | Cone dystrophy, progressive |
| Hemophilia B | Prostate cancer susceptibility |
| Wernerin sensitivity | Fogley X mental retardation |
| Osteous dysplasia (male lethal), digital | Epidemiology bulosa, muscular type |
| Adrenoleukodystrophy | Diabetes insipidus, neurogenic |
| Adrenomyeloneuropathy | Carcinoma antigen |
| Colabindres, blue monochromatic | Dyskeratosis |
| Cardiac valvular dysplasia | Hemochromatosis |
| Erey-DeRube muscular dystrophy | Hunter syndrome |
| Heterotopia, periventricular | Mucopolysaccharidosis |
| Fusion | Intestinal pseudoobstruction, neuronal |
| Hemolytic anemia | Melanoma antigens |
| Colabindres, green cone pigment | Mental retardation-deafness dysplasia |
| Incontinentia pigmenti, type II | Myotubular myopathy |
| Hydrocephalus | Optic atrophy (syndrome, type I) |
| MASA syndrome | Colabindres, red cone pigment |
| Spastic paraplegia | Gonorrhea DCR syndrome |
| Teti syndrome | Walden parkinsonism mental retardation |
| Mature T cell proliferation | Saich syndrome |
| Myopia (Borholm eye disease) | Cardiomyopathy, dilated |
| Mental retardation with psychosis | Noncompaction of left ventricular myocardium |
| Endocardial fibroelastosis | Van Hope-Lindau binding protein |

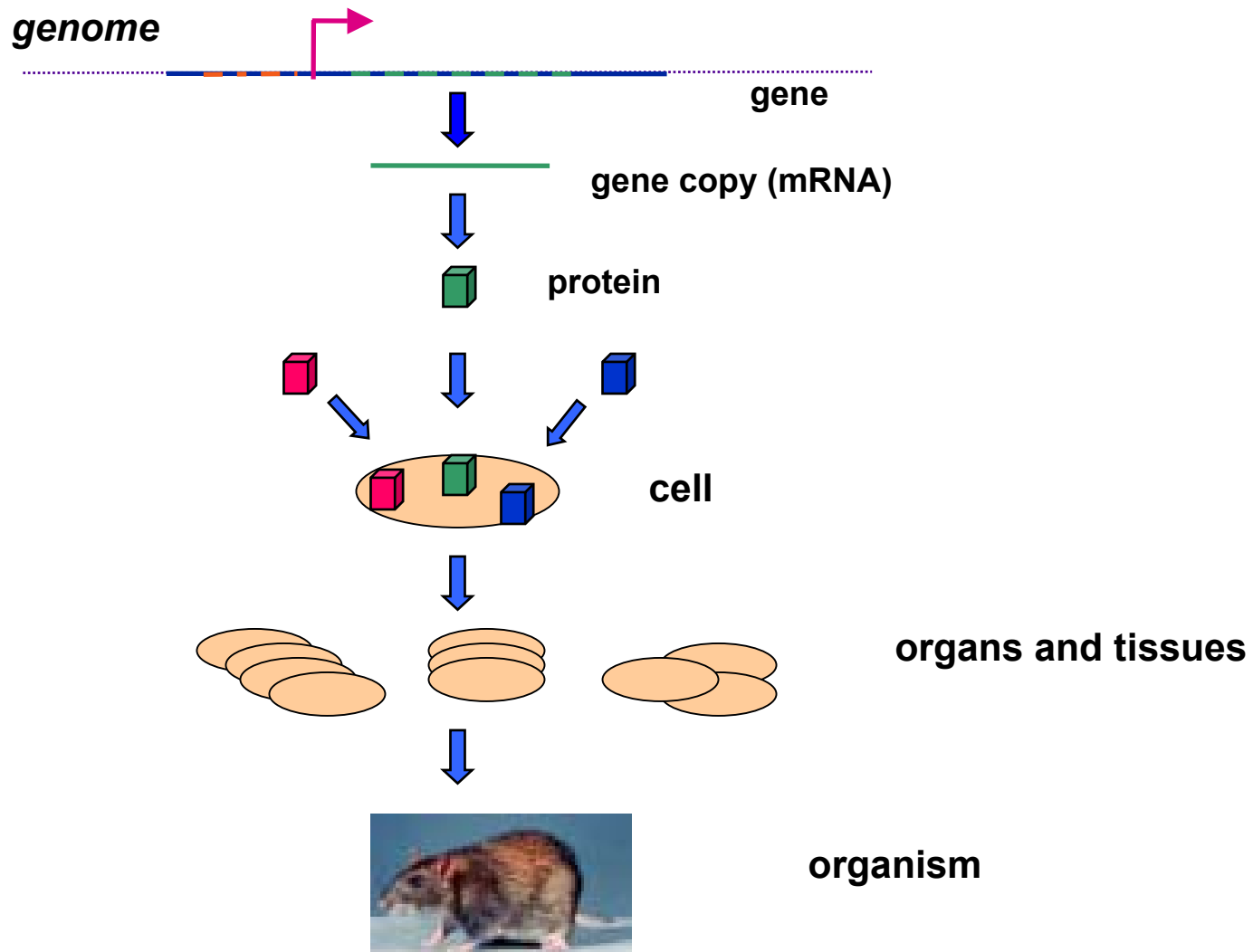
functional genomics



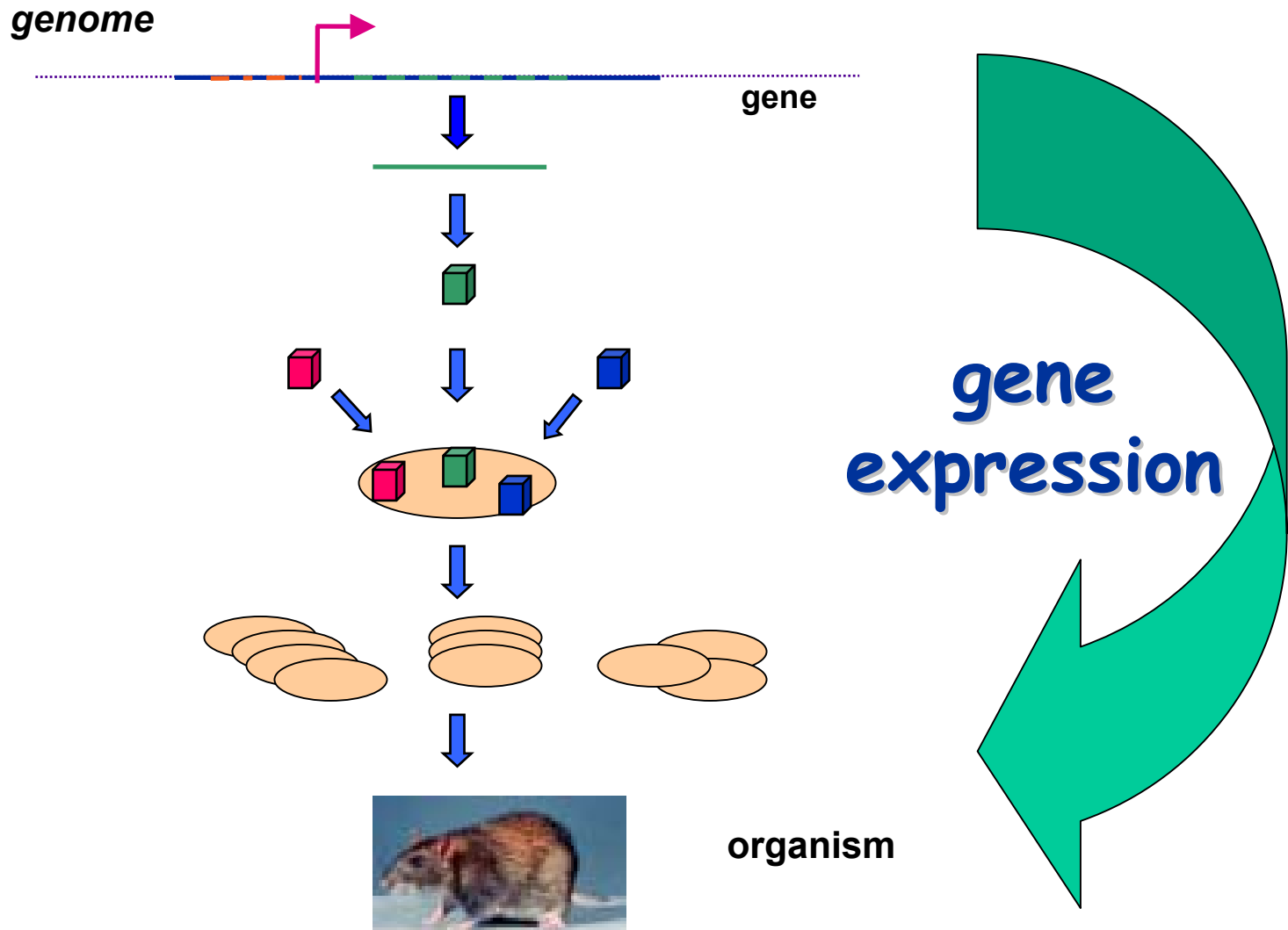
from genome to organism



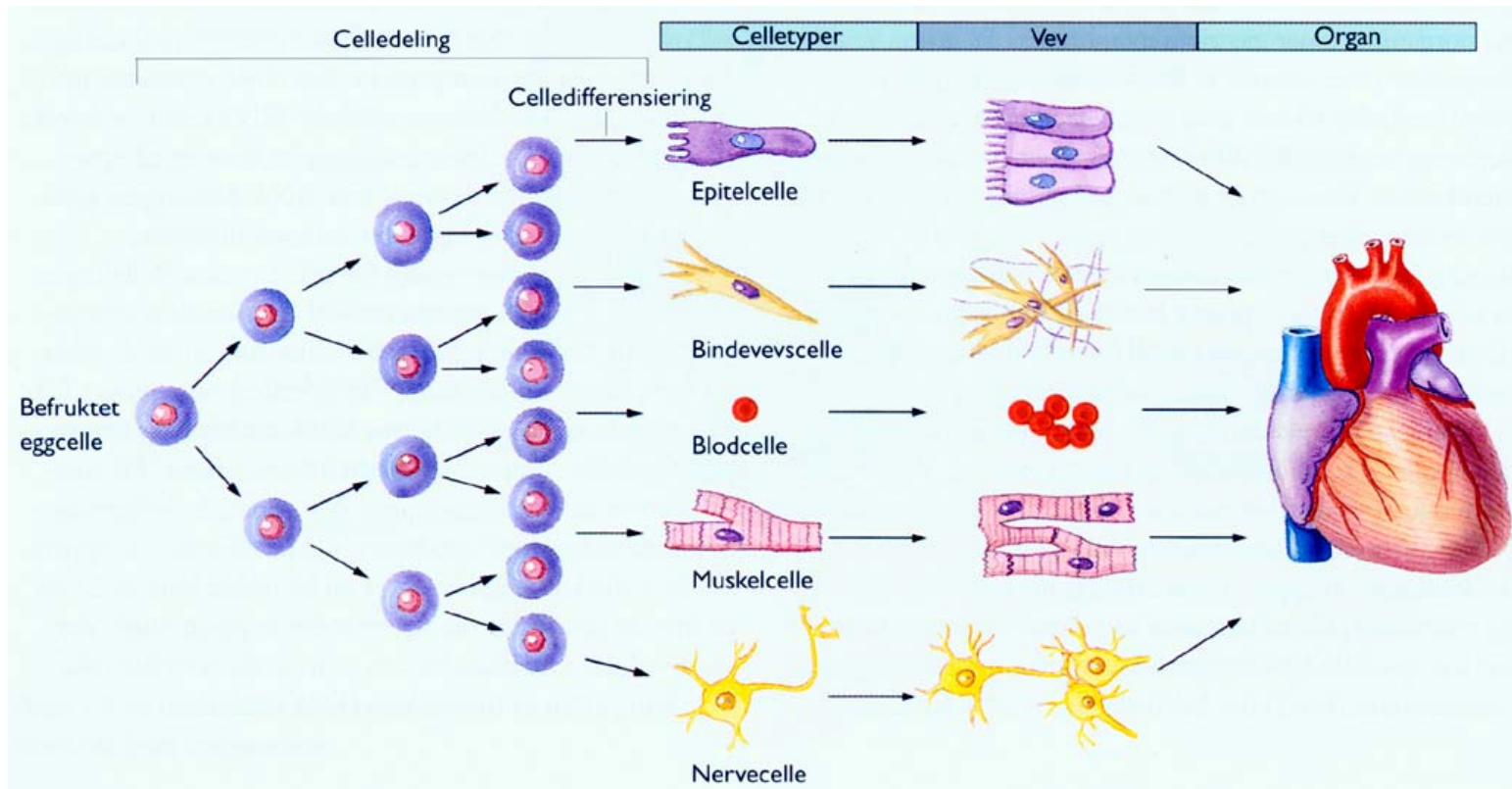
from genome to organism



from genome to organism

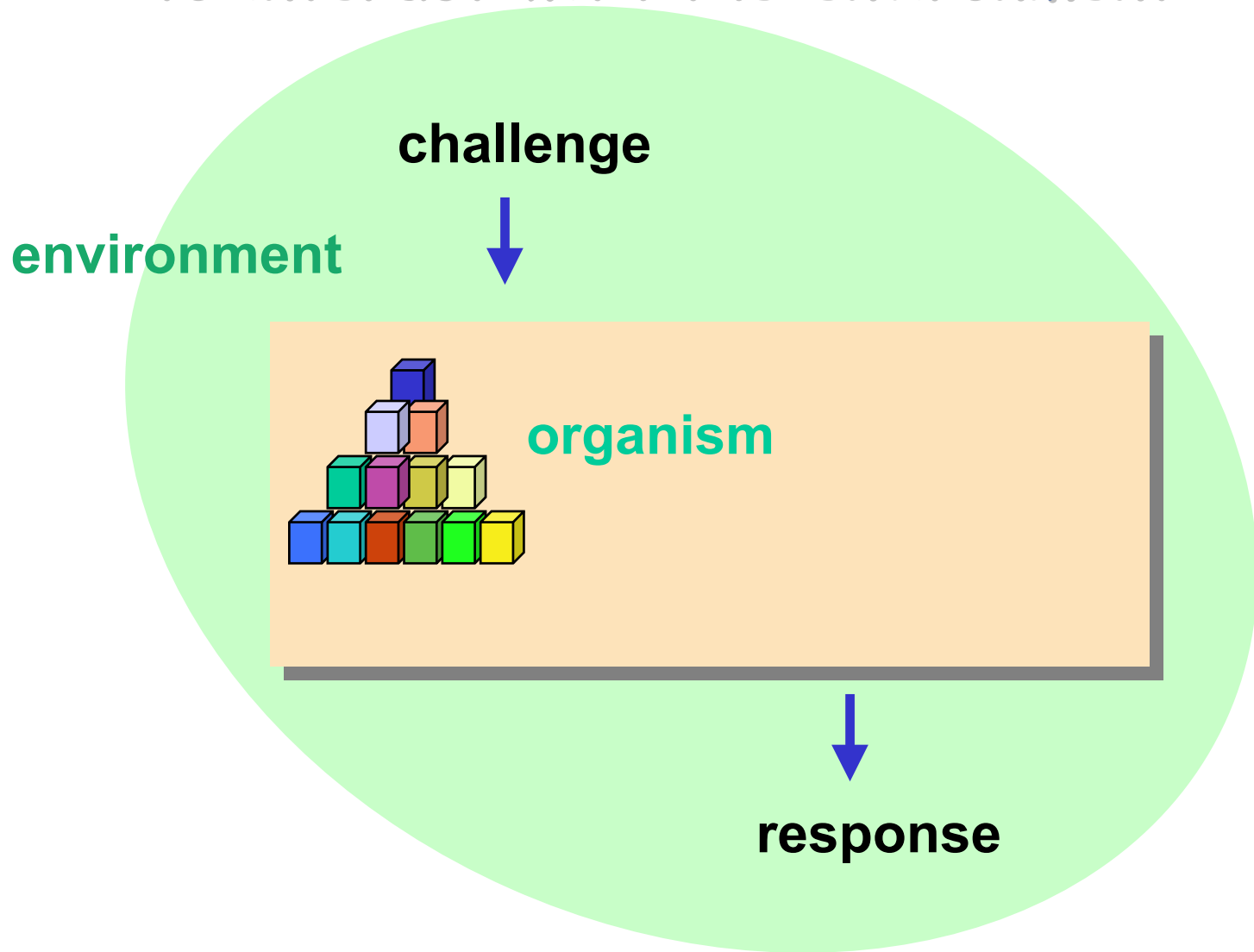


all cell types contain the same genome
 ..but differ in gene expression patterns....

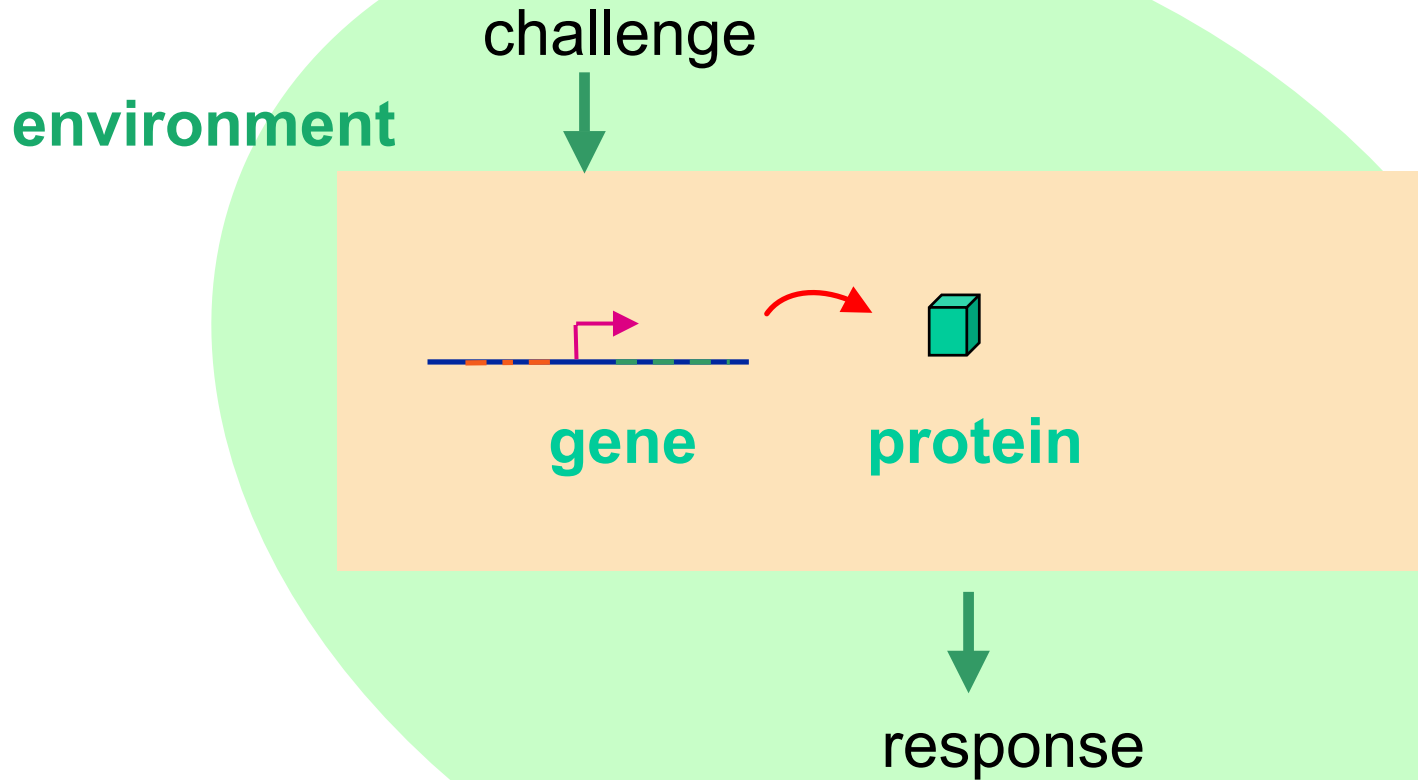


gene expression determines what you are....

to live is
to interact with the environment

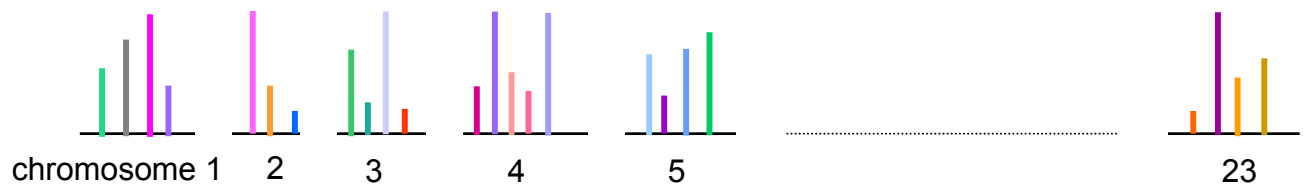


the organism responds
by making new proteins



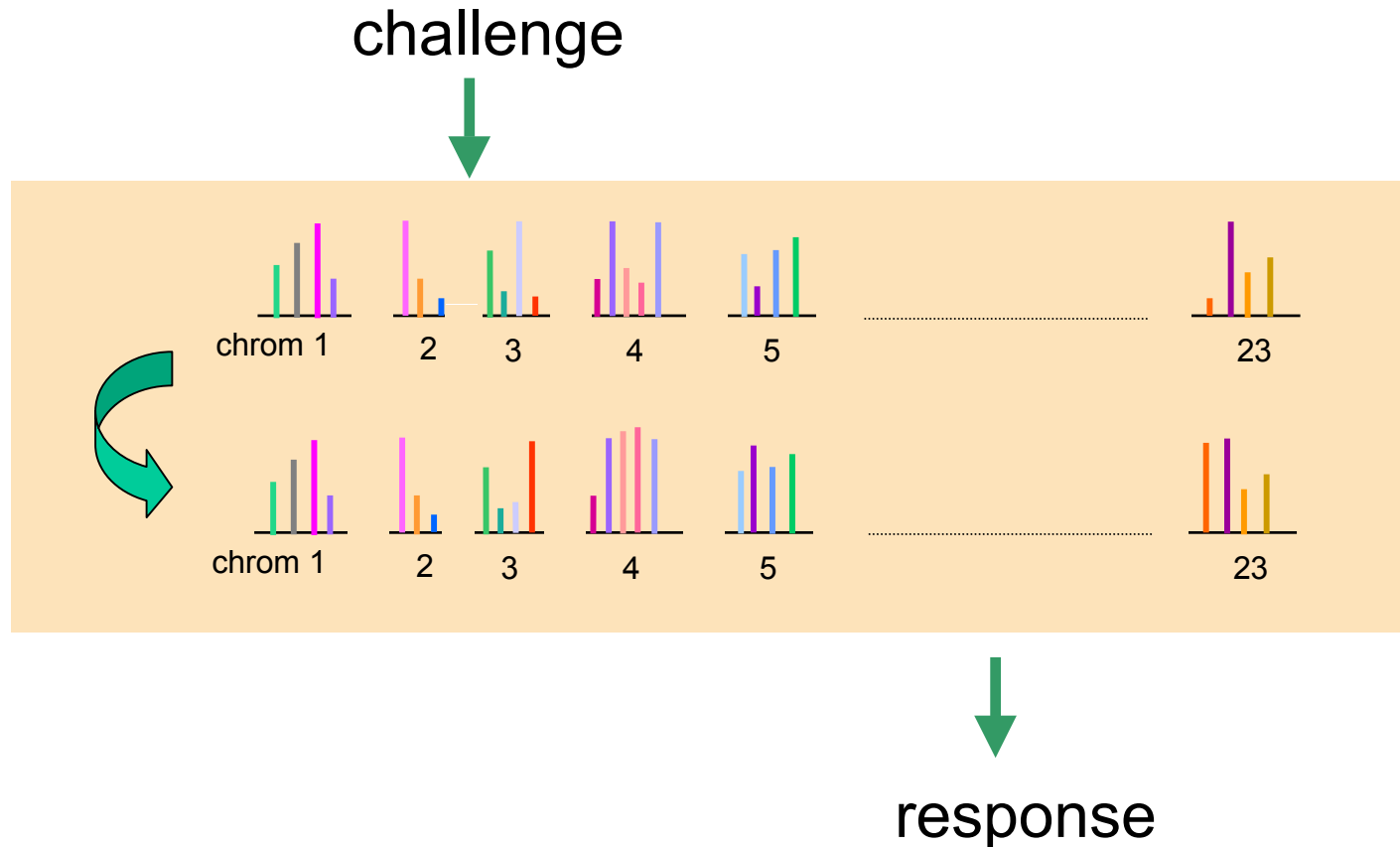
or by stop making
some of the old ones.....

we can learn more by measuring expression of **all** genes

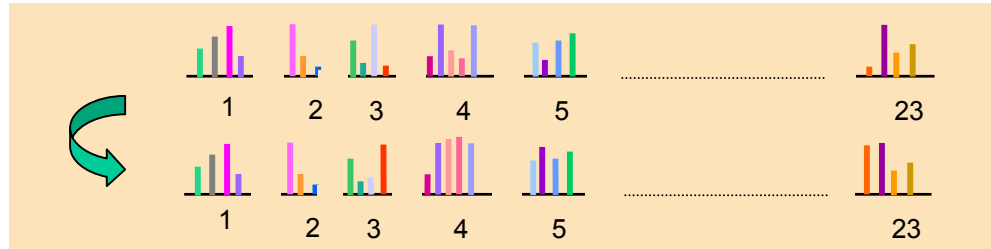


gene expression depends on cell type and cell state

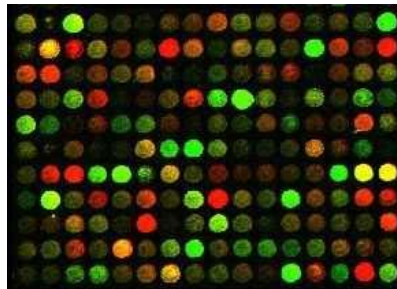
by measuring changes in gene expression we can discover genes participating in a given biological response



measure thousands of genes and proteins in high throughput analyses

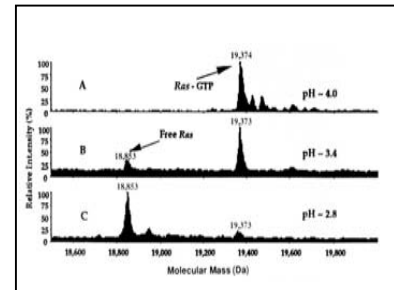


mRNA profiling



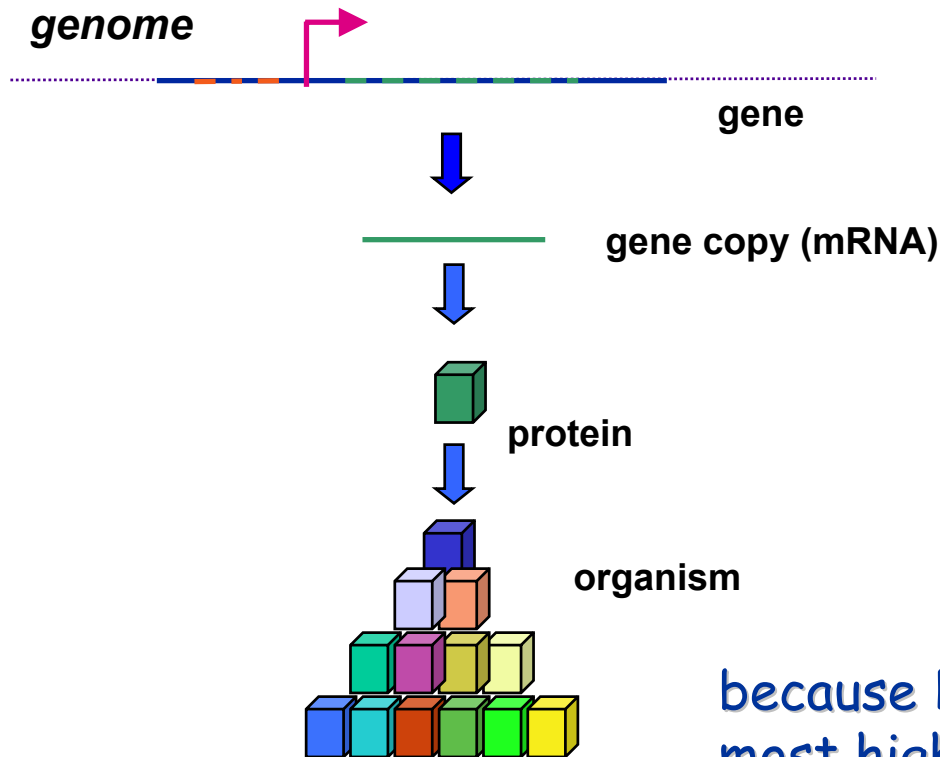
microarray

protein profiling



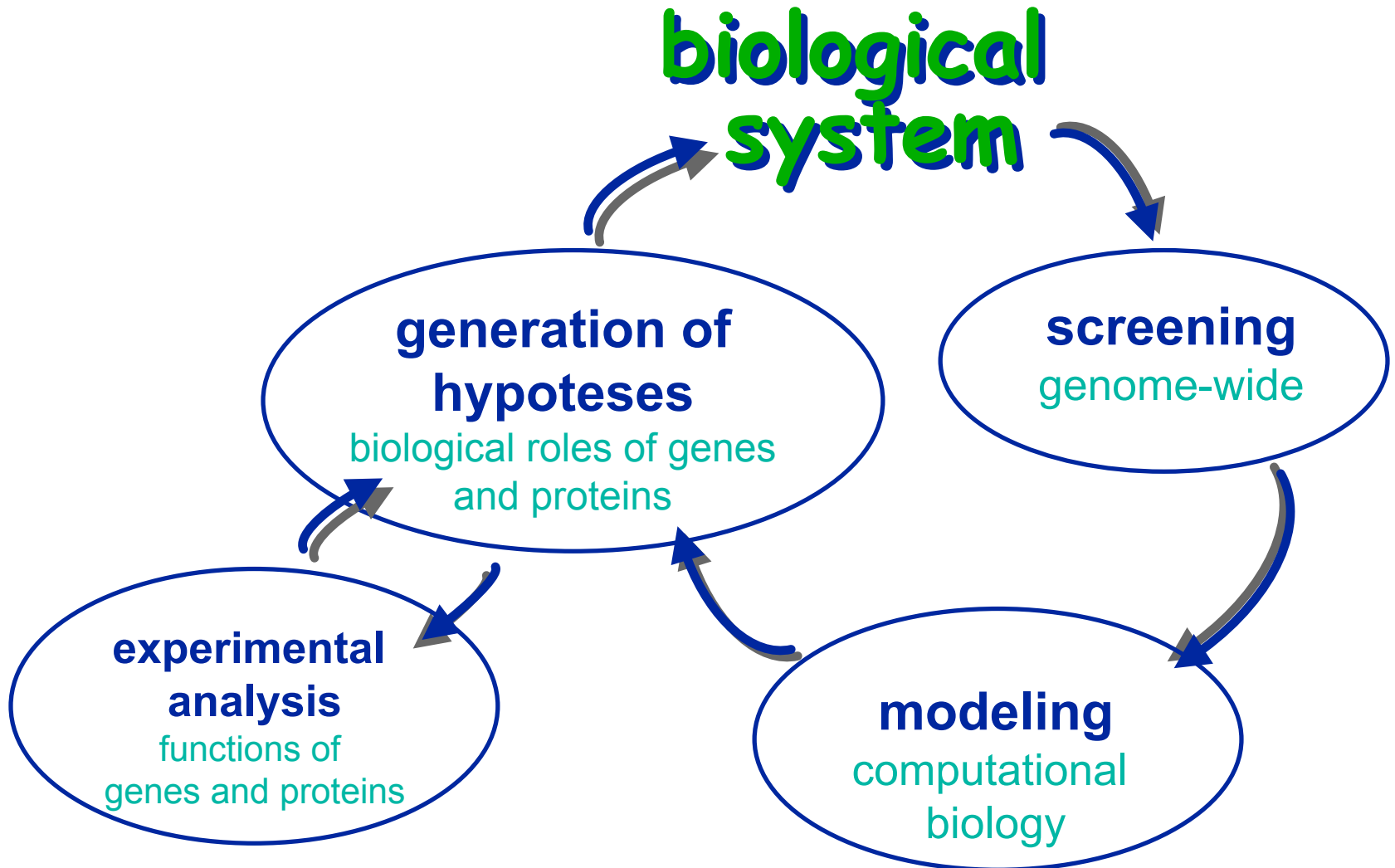
2D gel electrophoresis
mass spectrometry

why measure mRNA?

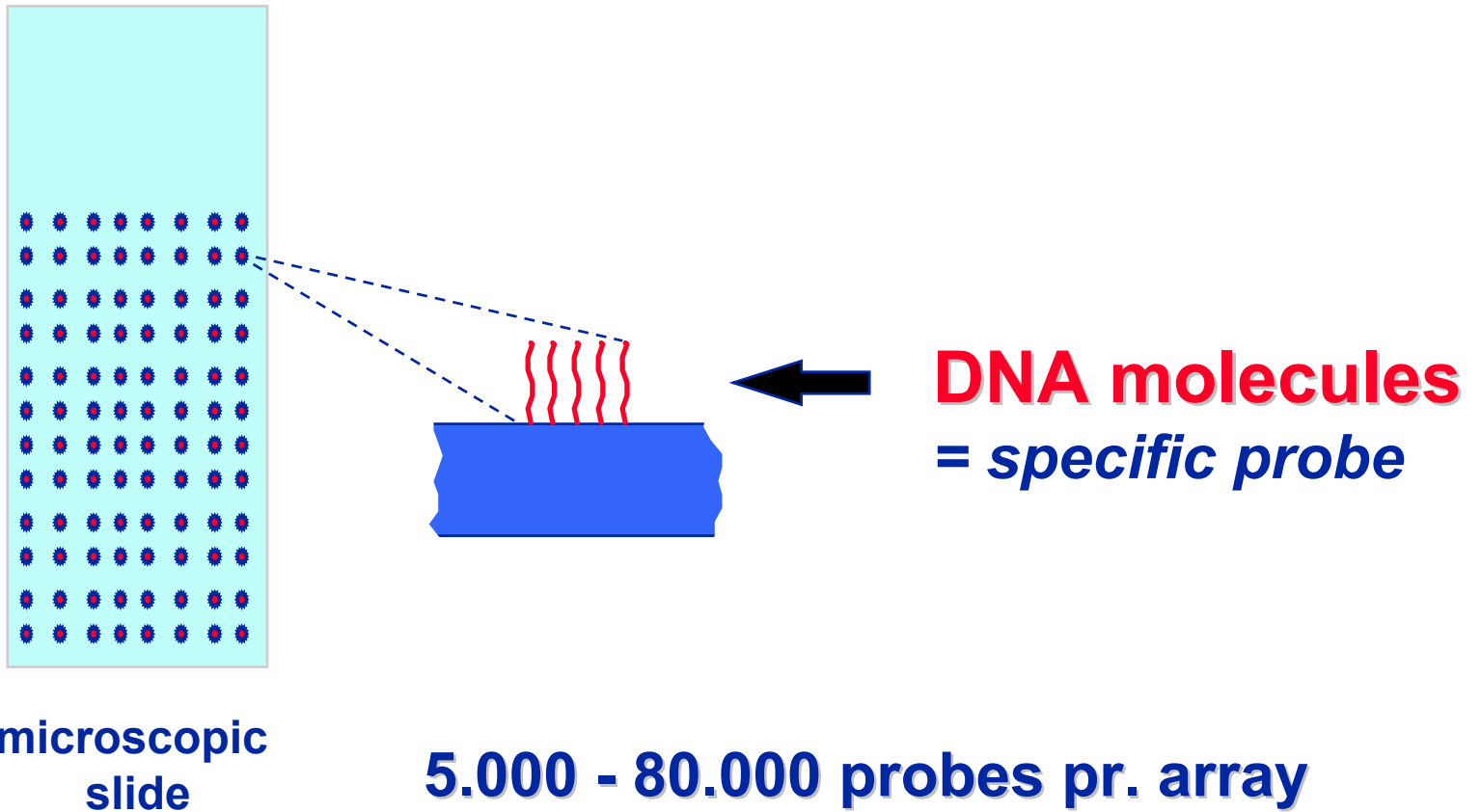


because DNA microarray is the most high throughput method that can measure gene expression with high sensitivity and specificity

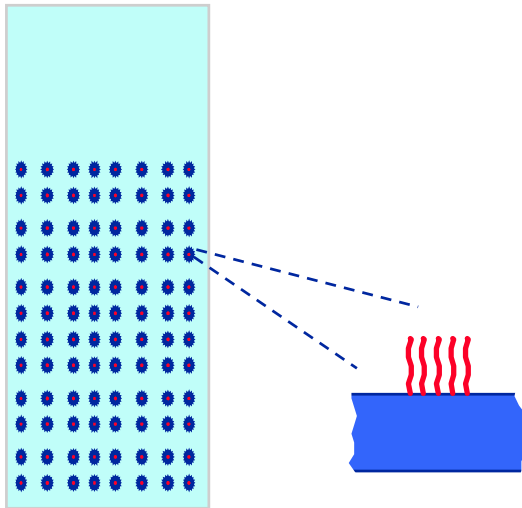
functional genomics



DNA microarray



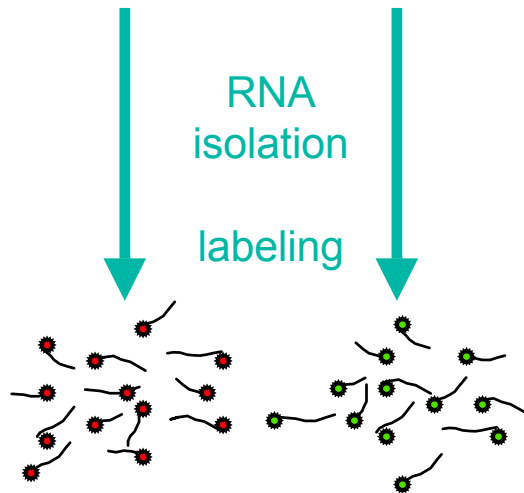
microarray formats



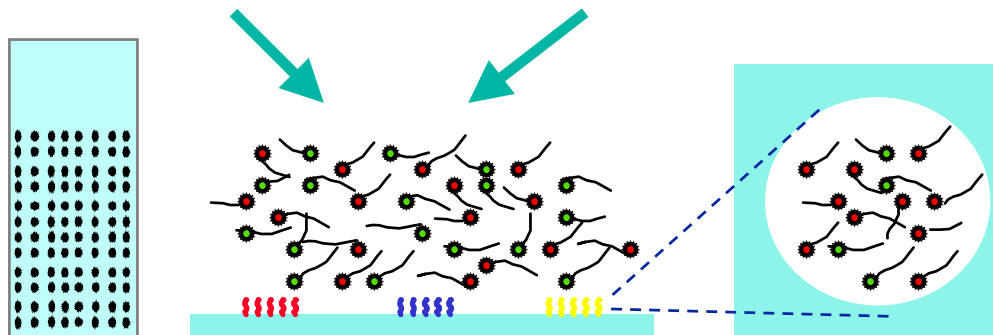
- ▶ cDNA (500-1500 bp)
- ▶ long oligonucleotides (40-70-mers)
- ▶ short oligonucleotides (20-25-mers)

microarray analysis

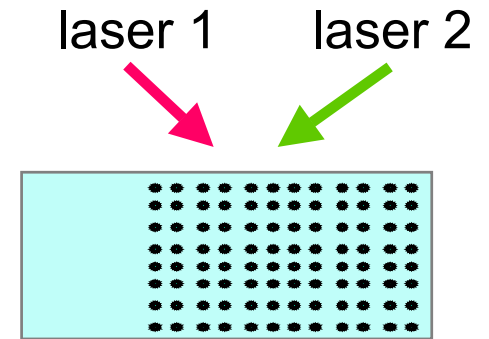
1 sample / control



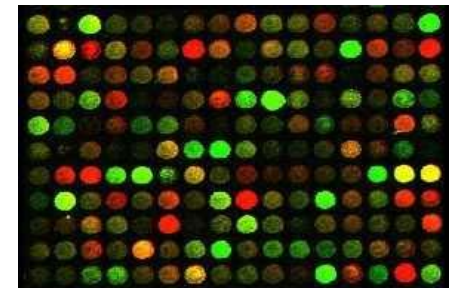
2 hybridization



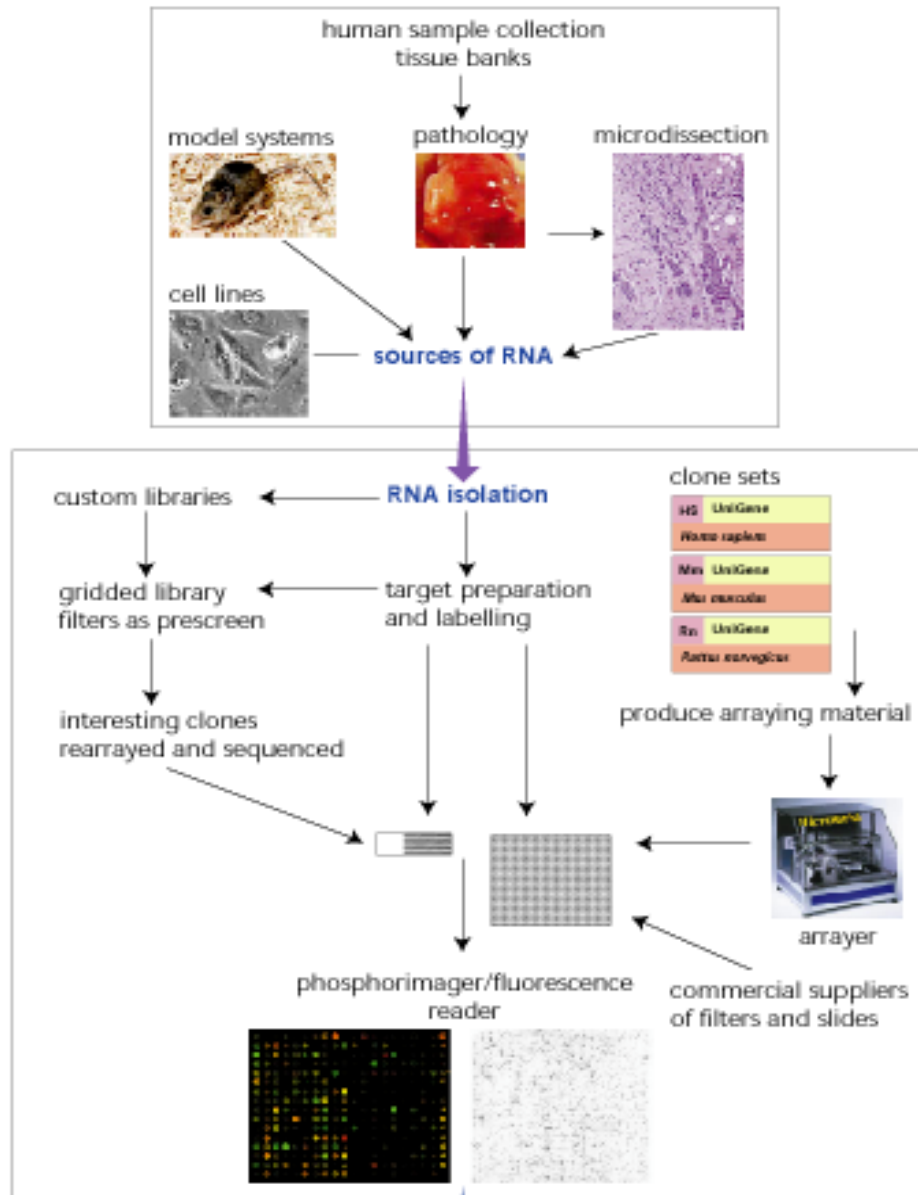
3 scanning



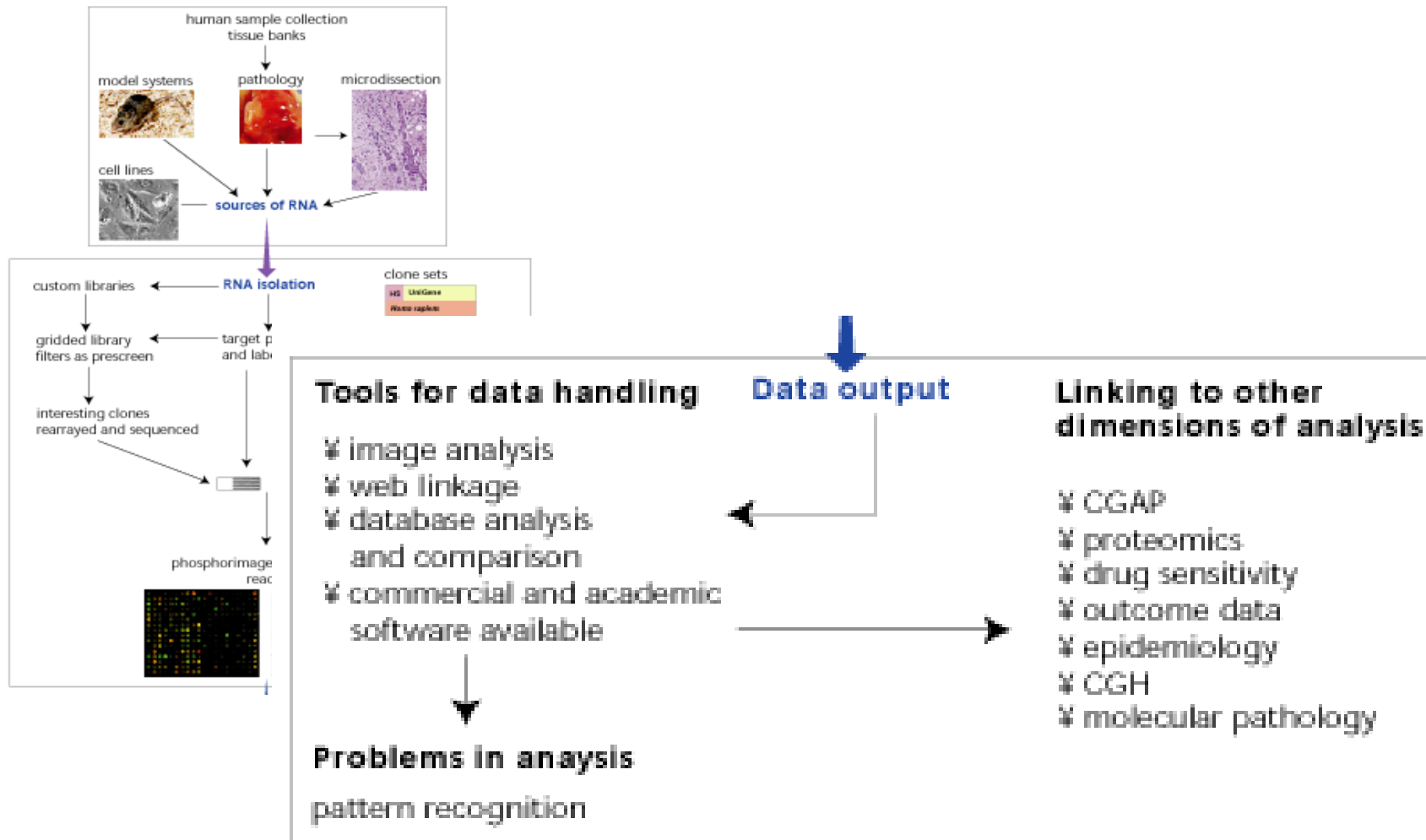
red = "up"
green = "down"



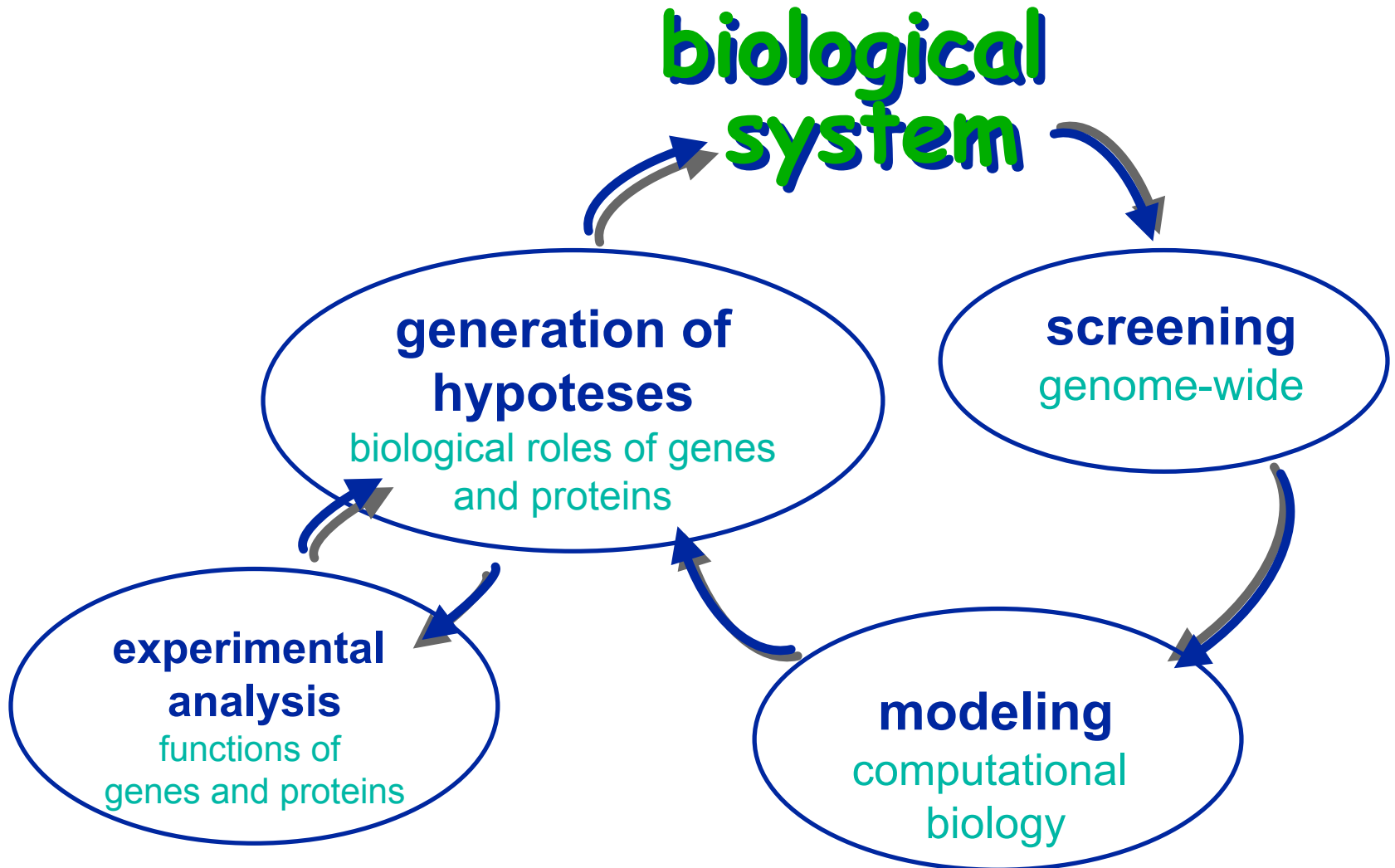
microarray analysis

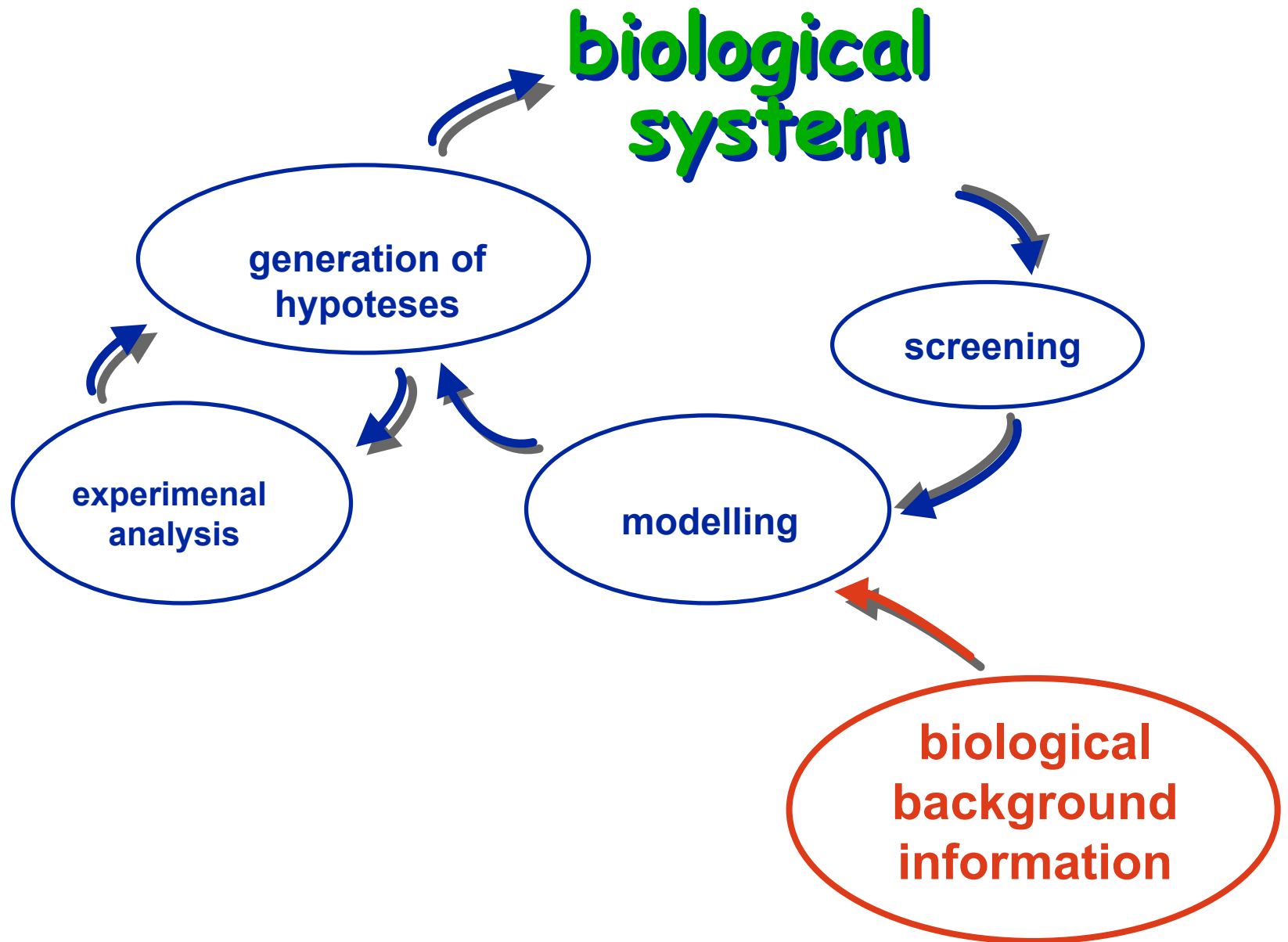


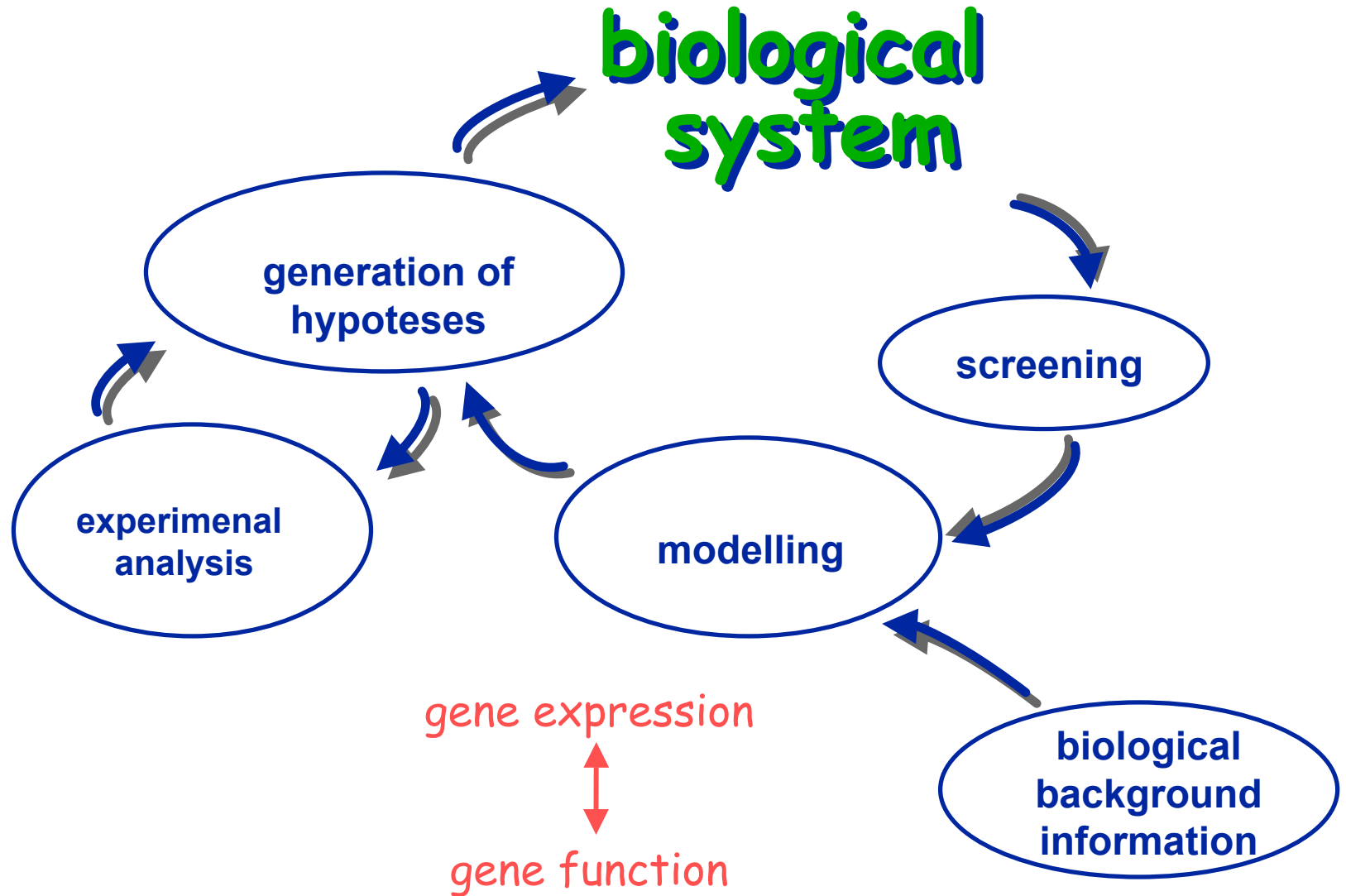
microarray analysis



functional genomics







functional classification of genes from time profiles

Astrid Lægreid¹ and Jan Komorowski²

¹Department of Cancer Research and Molecular Medicine

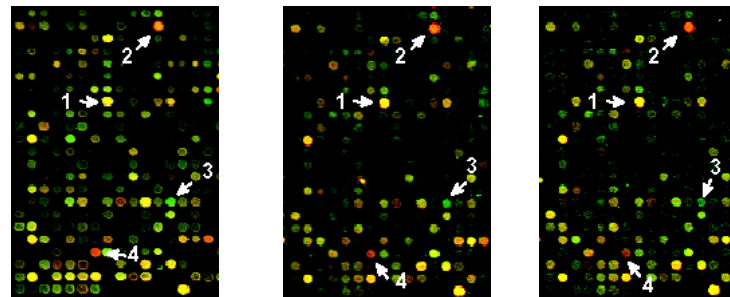
Norwegian University of Science and Technology

²The Linnaeus Centre for Bioinformatics,

Uppsala

The Transcriptional Program in the Response of Human Fibroblasts to Serum

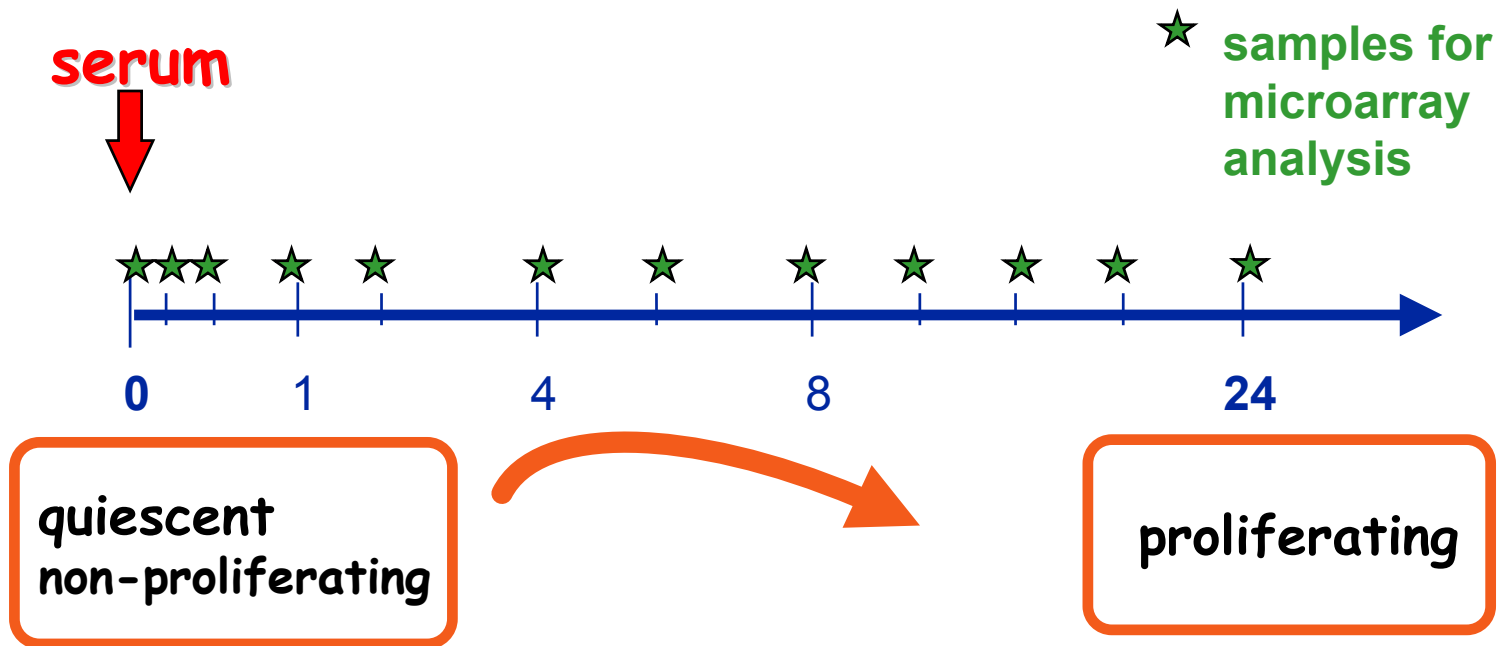
Iyer et al, *Science*, 283: 83, 1999



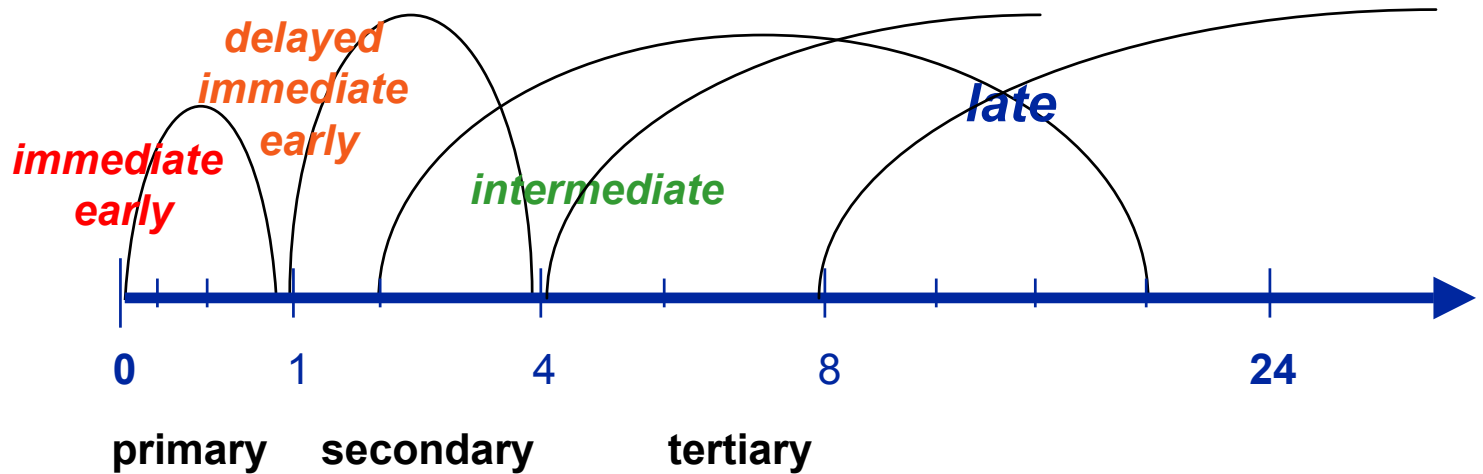
8 hours serum treatment

- 1, protein disulfide isomerase-related protein
- 2, IL-8 precursor
- 3, EST AA057170
- 4, vascular endothelial growth factor

fibroblast - 24 h serum response



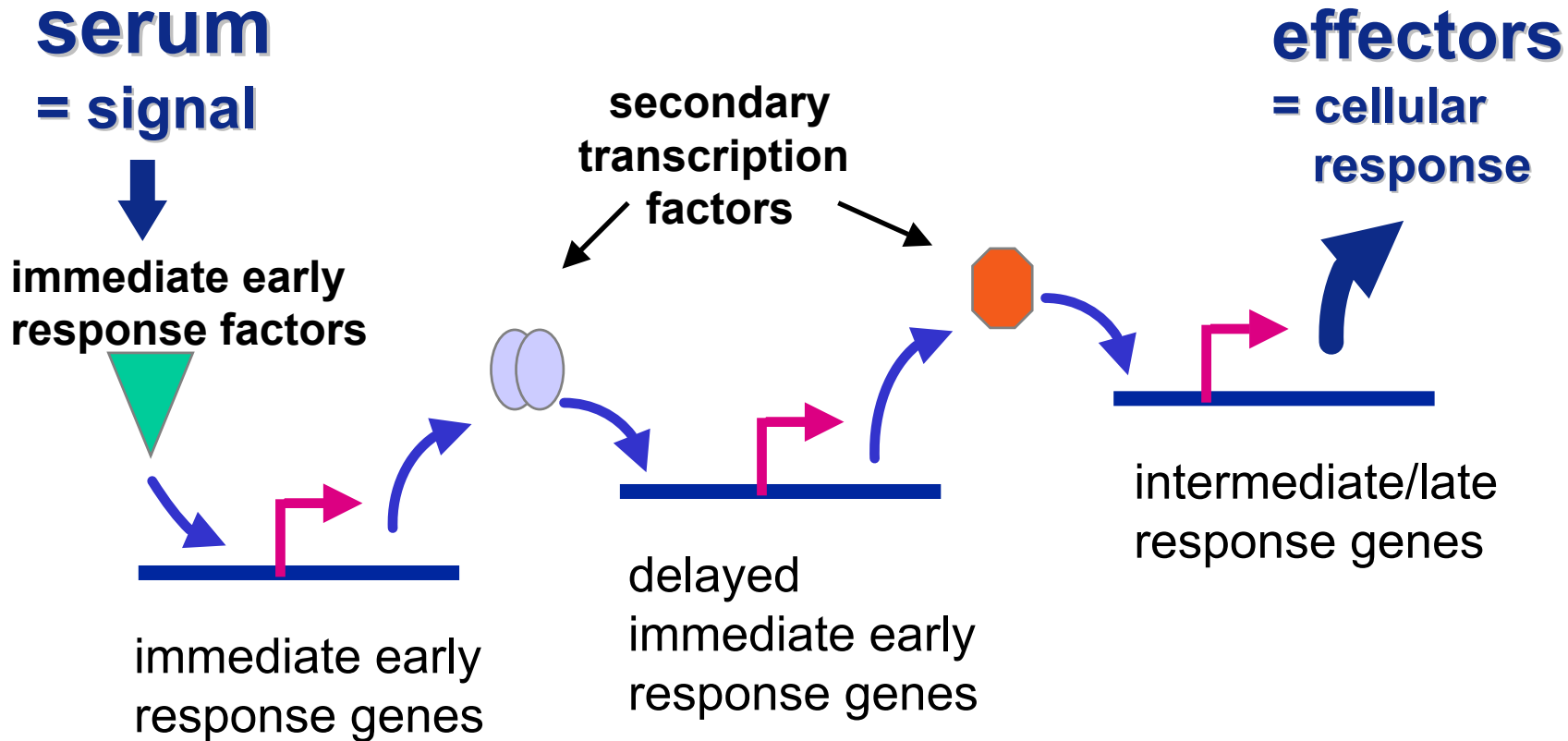
dynamic processes



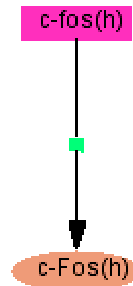
quiescent
non-proliferating

proliferating

molecular mechanisms of transcriptional response



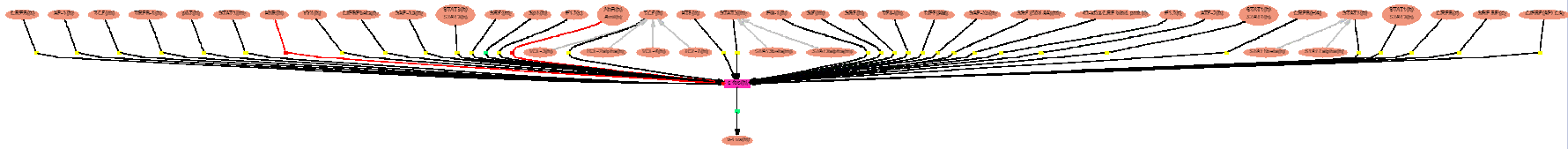
fos - immediate early transcription factor



Transpath; biobase.de

fos - immediate early transcription factor

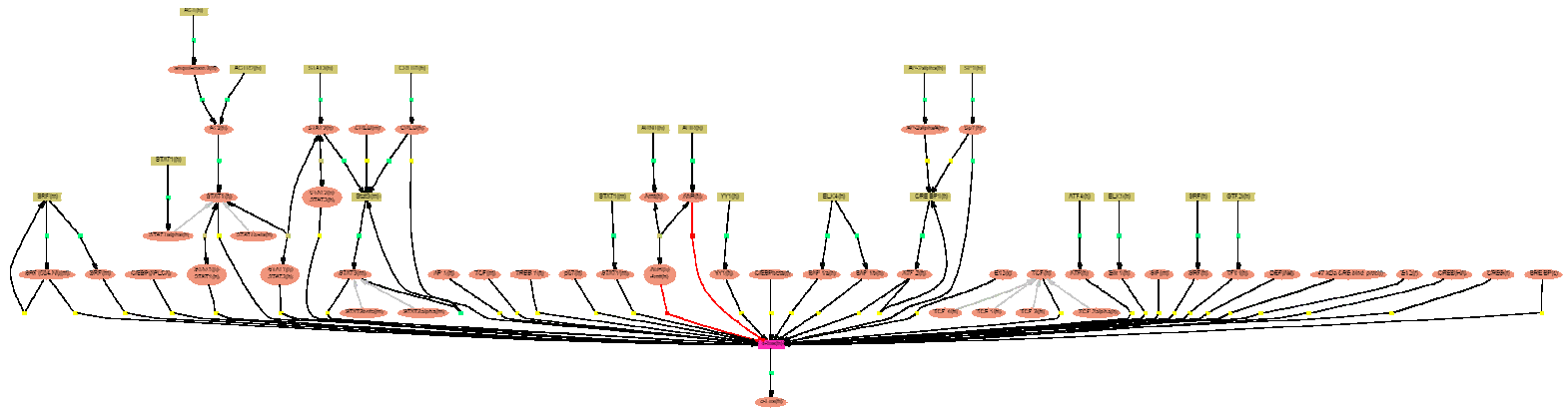
upstream factors



Transpath; biobase.de

fos - immediate early transcription factor

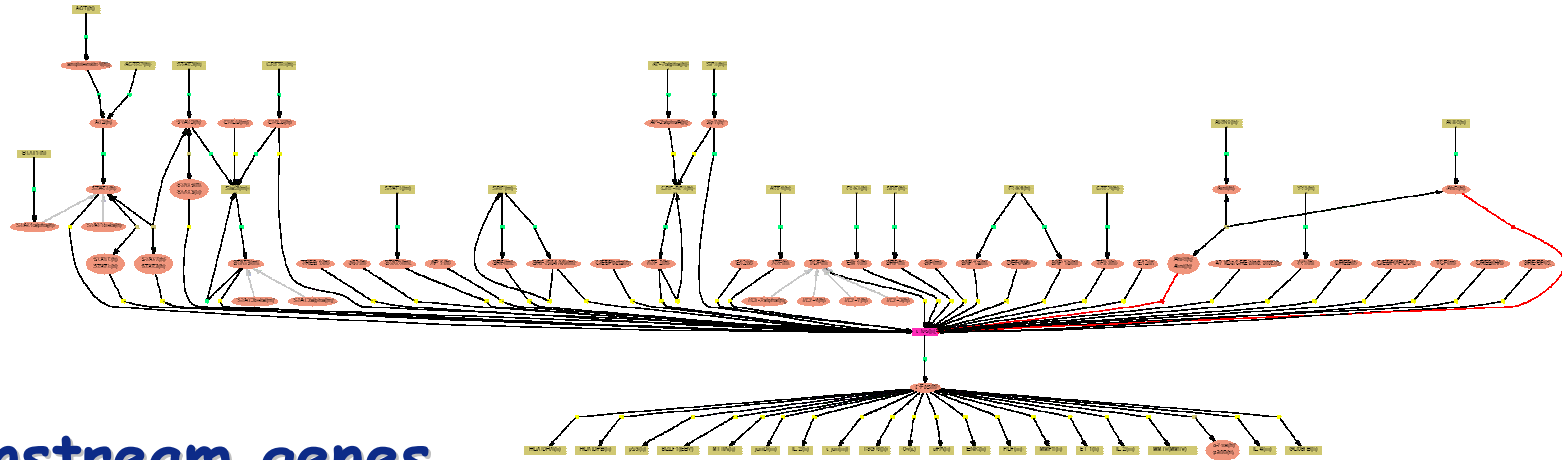
upstream factors



Transpath; biobase.de

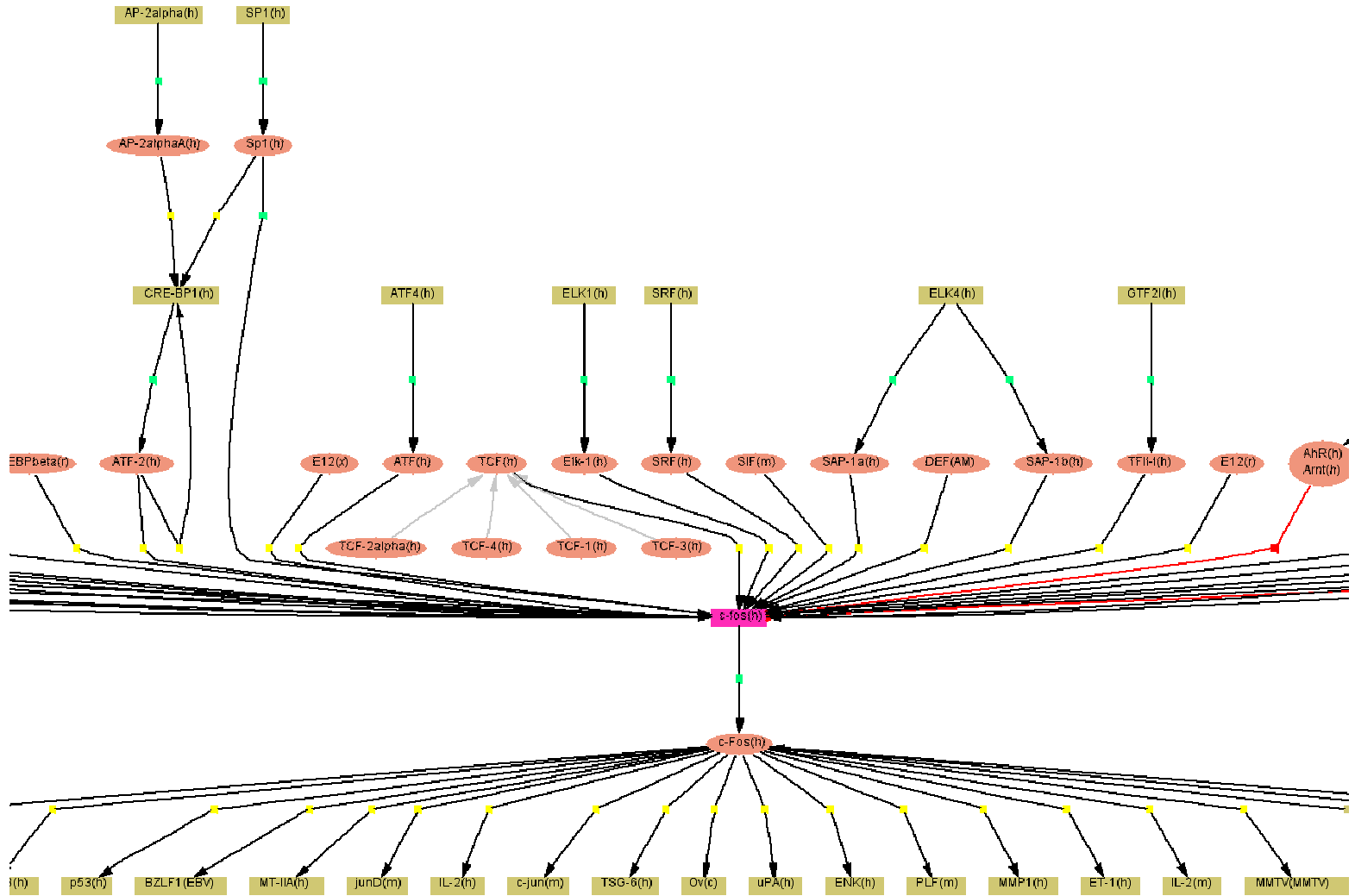
fos - immediate early transcription factor

upstream factors



+ downstream genes

Transpath; biobase.de



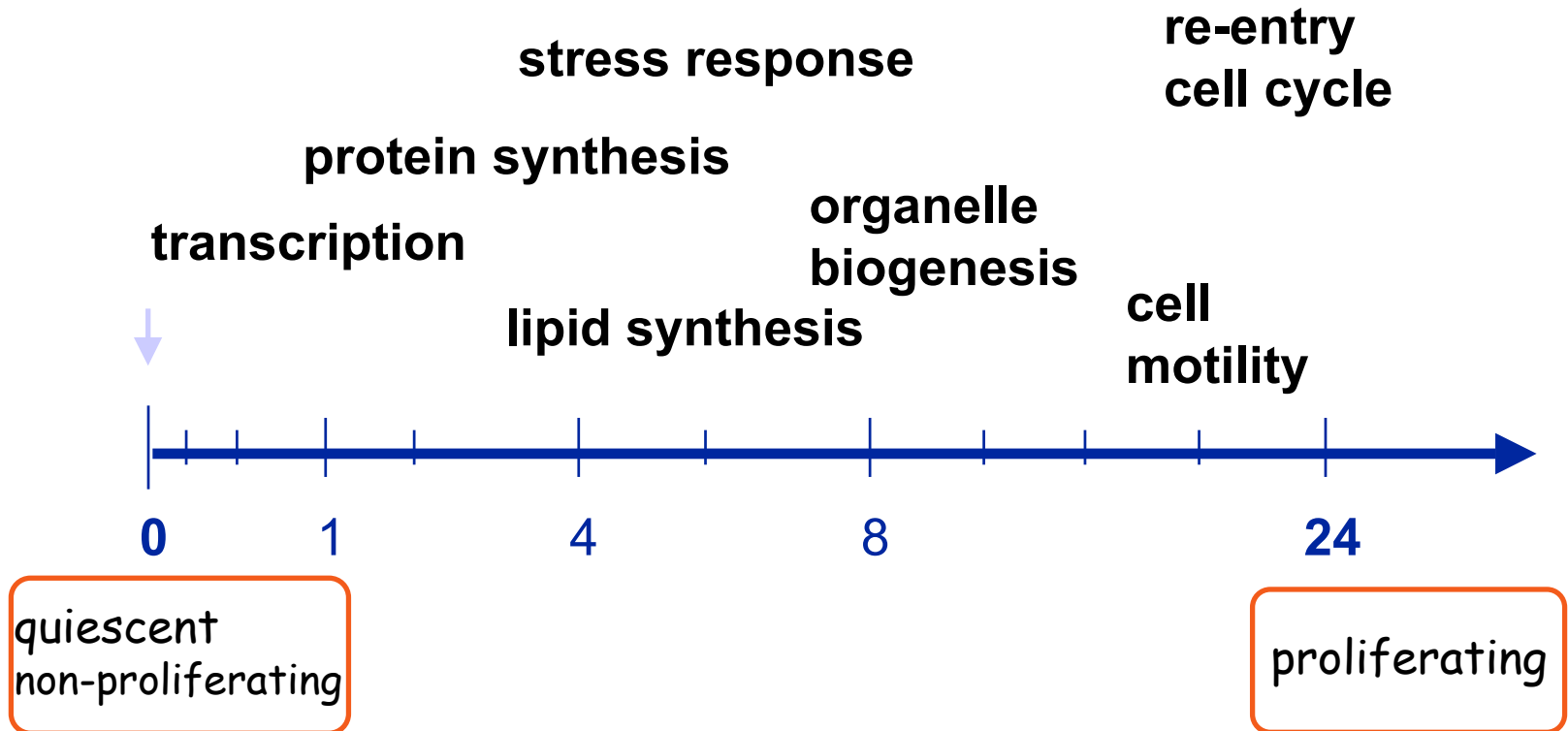
co-regulation of genes coding for proteins in a network in fibroblast serum-response

pro-endothelin → active endothelin → inactive endothelin

co-regulation of genes coding for proteins in a network in fibroblast serum-response



cellular processes



fibroblast serum-response transcriptional program

517 gen-probes differential gene expression

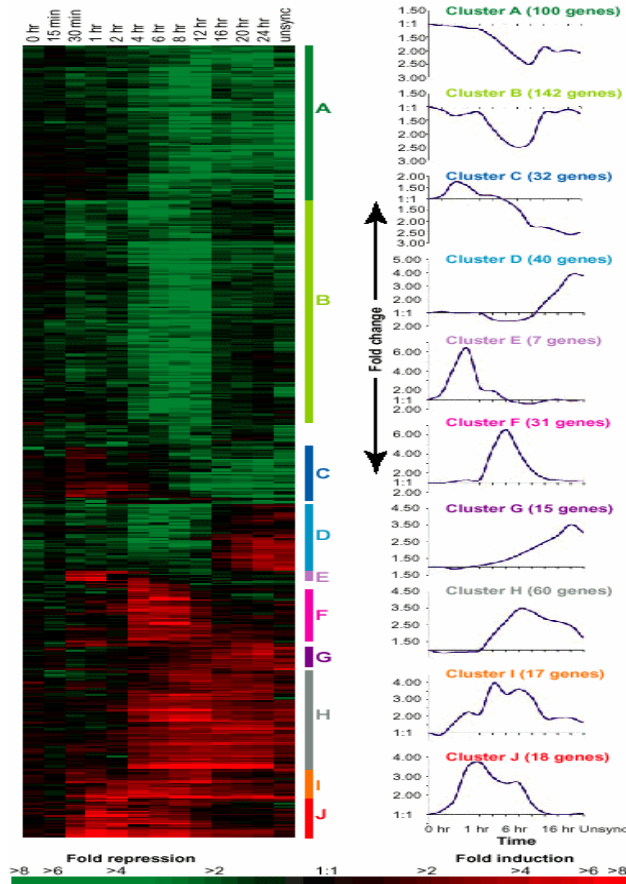
497 unique genes

284 known genes

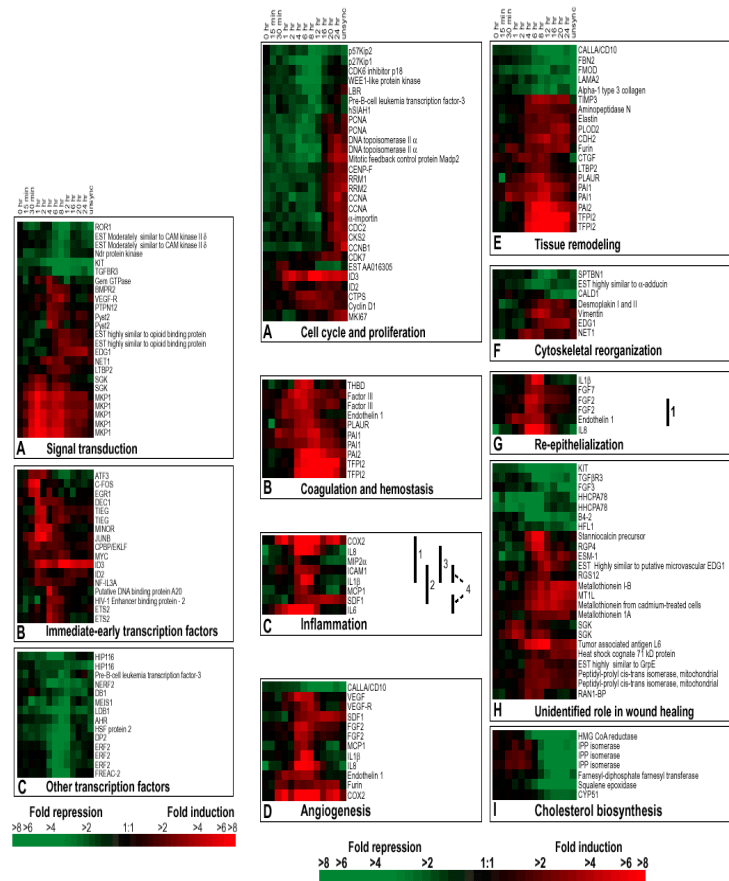
213 unknown genes

Iyer's analysis of transcriptional fibroblast serum response

Expression clusters



Functional clusters



functional classification
from time profiles

our aim

find relationship between

gene function - gene expression profile

selected challenges in gene-expression analysis

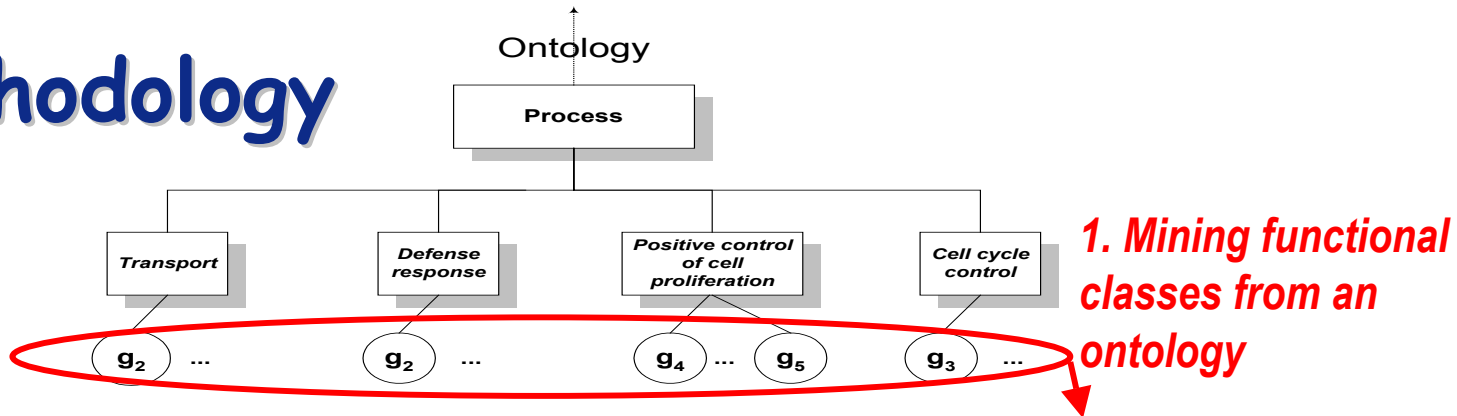
- ▶ function similarity corresponds to expression similarity but:
 - ▶ functionally correlated genes may be expression-wise dissimilar (e.g. anti-coregulated)
 - ▶ genes usually have multiple function
 - ▶ measurements may be *approximate* and *contradictory*

- ▶ can we obtain clusters of biologically related genes?

- ▶ can we build models that classify unknown genes to functional classes, that are human legible, and that handle *approximate* and often *contradictory* data?

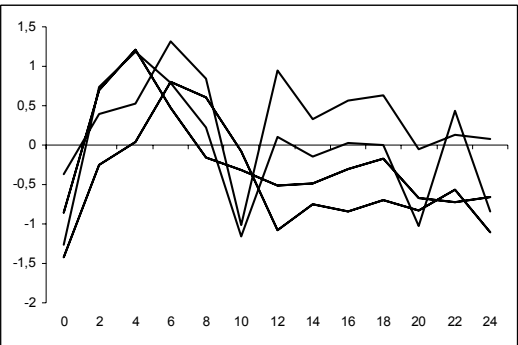
- ▶ how can we re-use biological knowledge?

methodology



Gene	0HR	15MIN	30MIN	1HR	2HR	4HR	6HR	8HR	12HR	16HR	20HR	24HR	Process
g_1	0.00	-0.47	-3.32	-0.81	0.11	-0.60	-1.36	-1.03	-1.84	-1.00	-0.60	-0.94	Unknown
g_2	0.00	0.66	0.07	0.20	0.29	-0.89	-0.45	-0.29	-0.29	-0.15	-0.45	-0.42	Transport and defense response
g_3	0.00	0.14	-0.04	0.00	-0.15	-0.58	-0.30	-0.18	-0.38	-0.49	-0.81	-1.12	Cell cycle control
g_4	0.00	-0.04	0.00	-0.23	-0.25	-0.47	-0.60	-0.56	-1.09	-0.71	-0.76	-0.62	Positive control of cell proliferation
g_5	0.00	0.28	0.37	0.11	-0.17	-0.18	-0.60	-0.23	-0.58	-0.79	-0.29	-0.74	Positive control of cell proliferation
...

2. Extracting features for learning



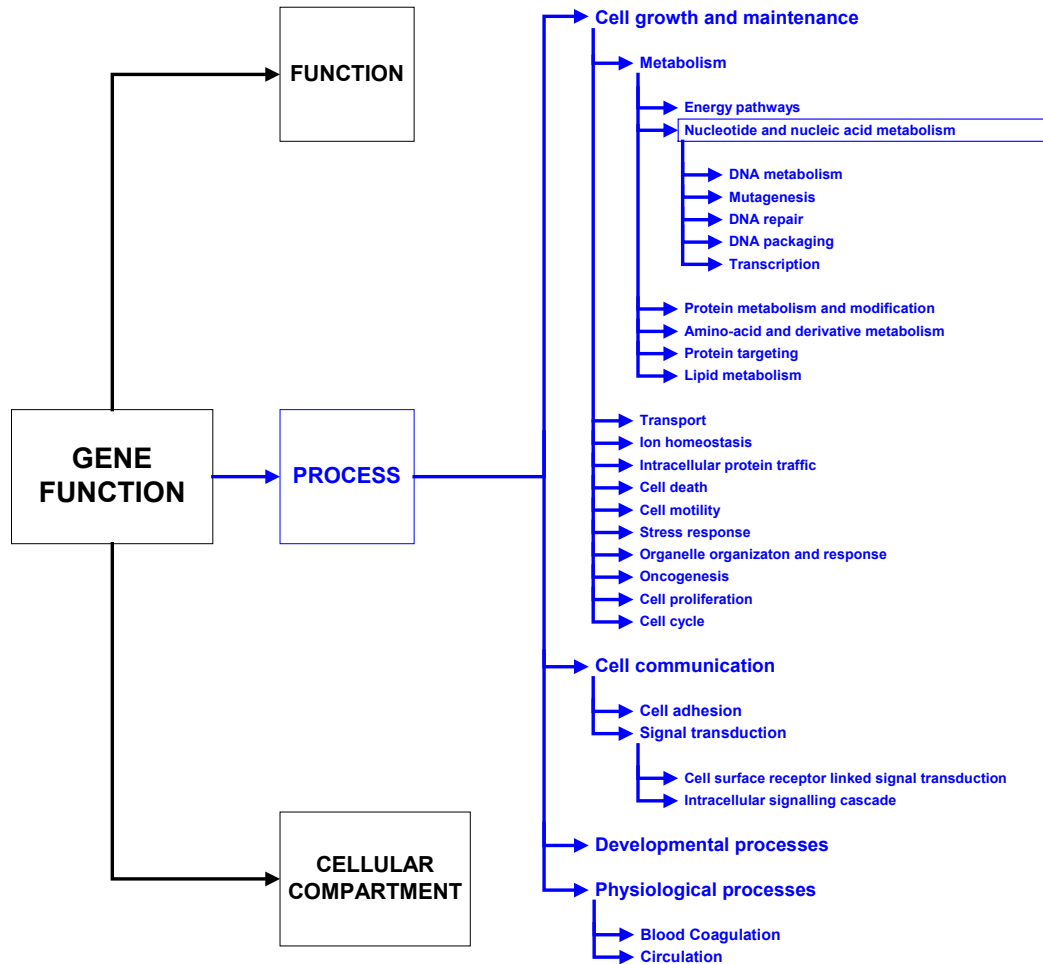
3. Inducing minimal decision rules using rough sets

0 - 4 (Increasing) AND 6 - 10 (Decreasing) AND 14 - 18 (Constant) => GO (cell proliferation)

4. The function of unknown genes is predicted using the rules



Gene Ontology

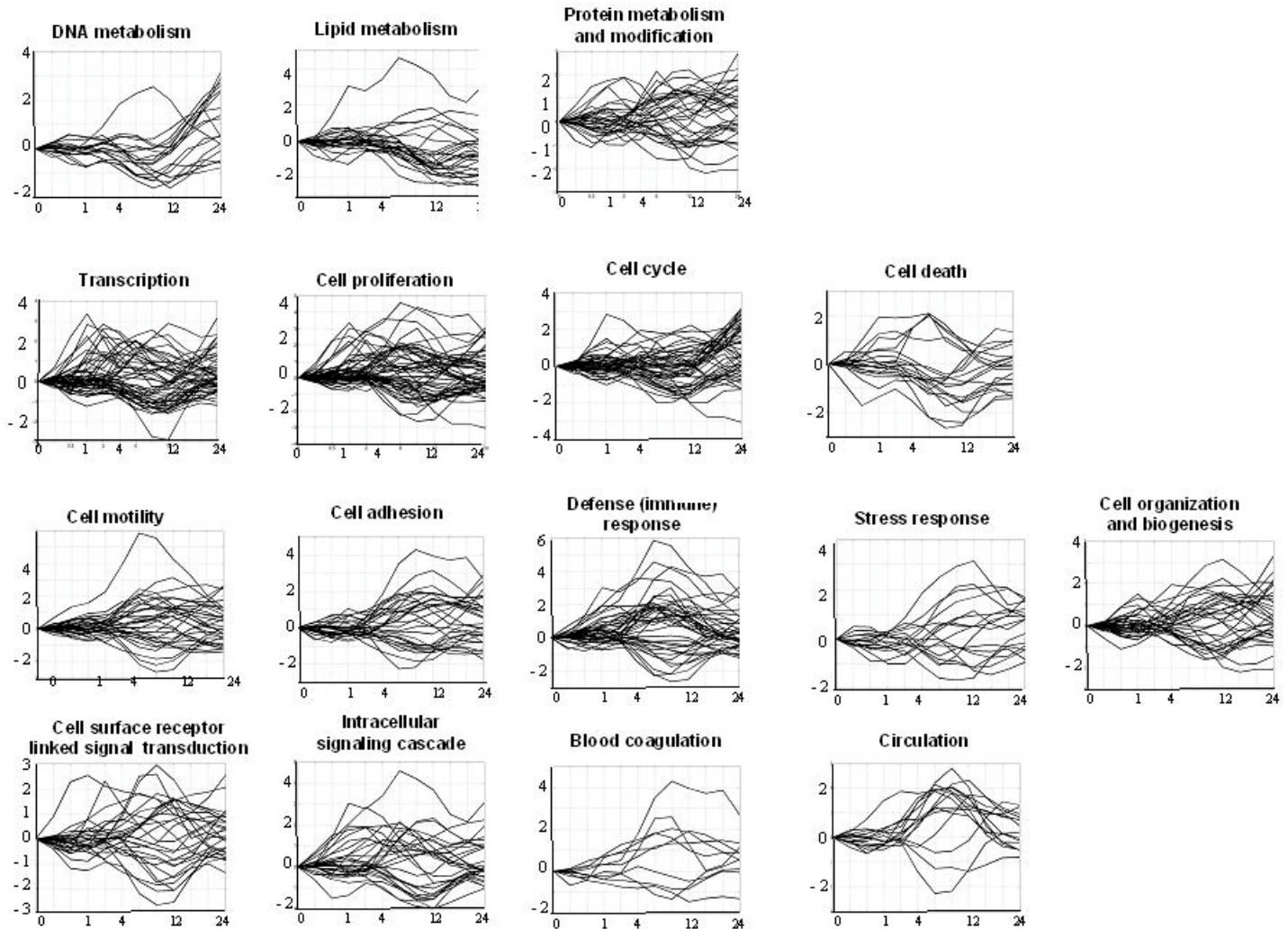


annotations

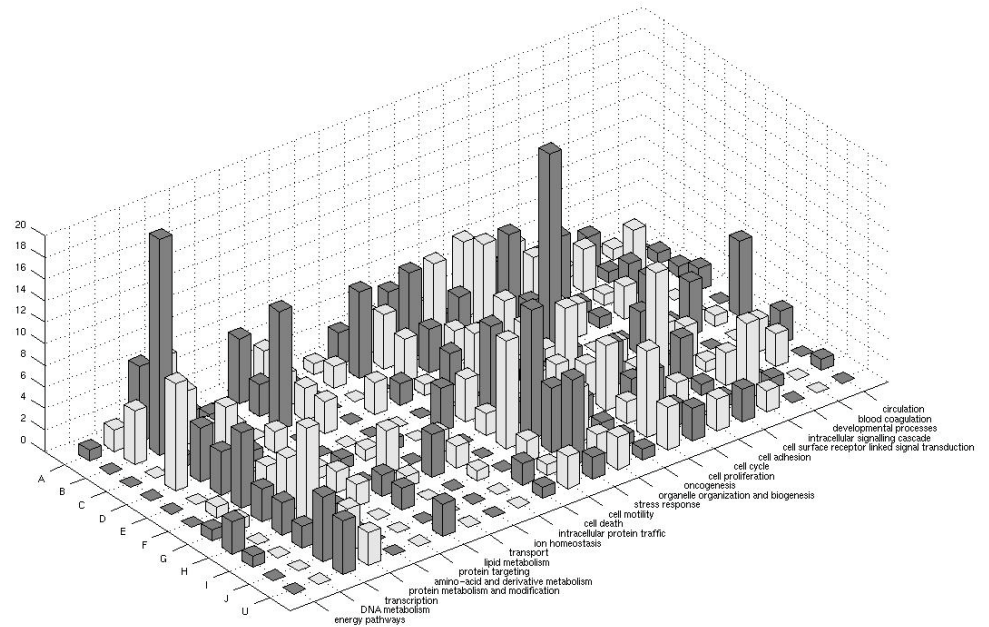
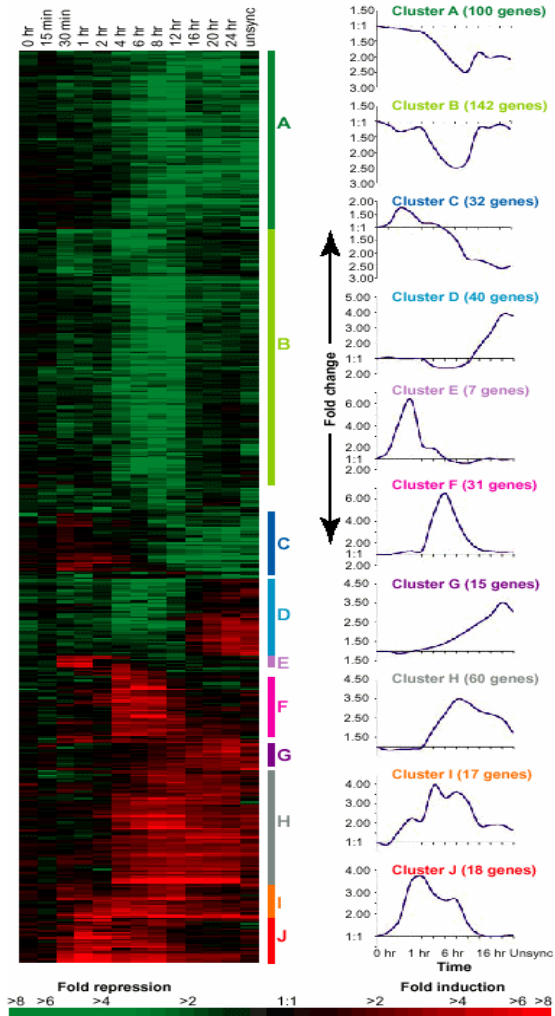
Annotation of Known Genes

GENE SYMBOL	GENE NAME	GENEBANK ACCESSION NUMBER	ANNOTATIONS AT THE MOST SPECIFIC LEVEL OF GO	ANNOTATIONS TO THE 23 BROAD CELLULAR PROCESSES USED FOR LEARNING
SEPP1	selenoprotein P, plasma, 1	AA045003	oxidative stress response(GO:0006979), metal ion transport(GO:0006823)	stress response(GO:0006950), transport(GO:0006810)
EPB41L2	erythrocyte membrane protein band 4.1-like 2	W88572	positive control of cell proliferation(GO:0008284)	cell proliferation(GO:0008283)
OA48-18	acid-inducible phosphoprotein	AA029909	cell proliferation(GO:0008283)	cell proliferation(GO:0008283)
CTSK	cathepsin K (pseudodeficiency)	AA044619	proteolysis and peptidolysis(GO:0006508)	protein metabolism and modification(GO:0006411)
CPT1B	camitine palmitoyltransferase I, muscle	W89012	fatty acid beta-oxidation(GO:0006635)	lipid metabolism(GO:0006629)
CLDN11	claudin 11 (oligodendrocyte transmembrane protein)	N22392	cell adhesion(GO:0007155), substrate-bound cell migration(GO:0006929), cell proliferation(GO:0008283), developmental processes(GO:0007275)	cell adhesion(GO:0007155), cell motility(GO:0006928), cell proliferation(GO:0008283), developmental processes(GO:0007275)
RPL5	ribosomal protein L5	AA027277	protein biosynthesis(GO:0006412), ribosomal large subunit assembly and maintenance(GO:0000027)	protein metabolism and modification(GO:0006411), cell organization and biogenesis(GO:0006996)
	Homo sapiens clone 23785 mRNA sequence	N32247	calcium-independent cell-cell matrix adhesion(GO:0007161)	cell adhesion(GO:0007155)

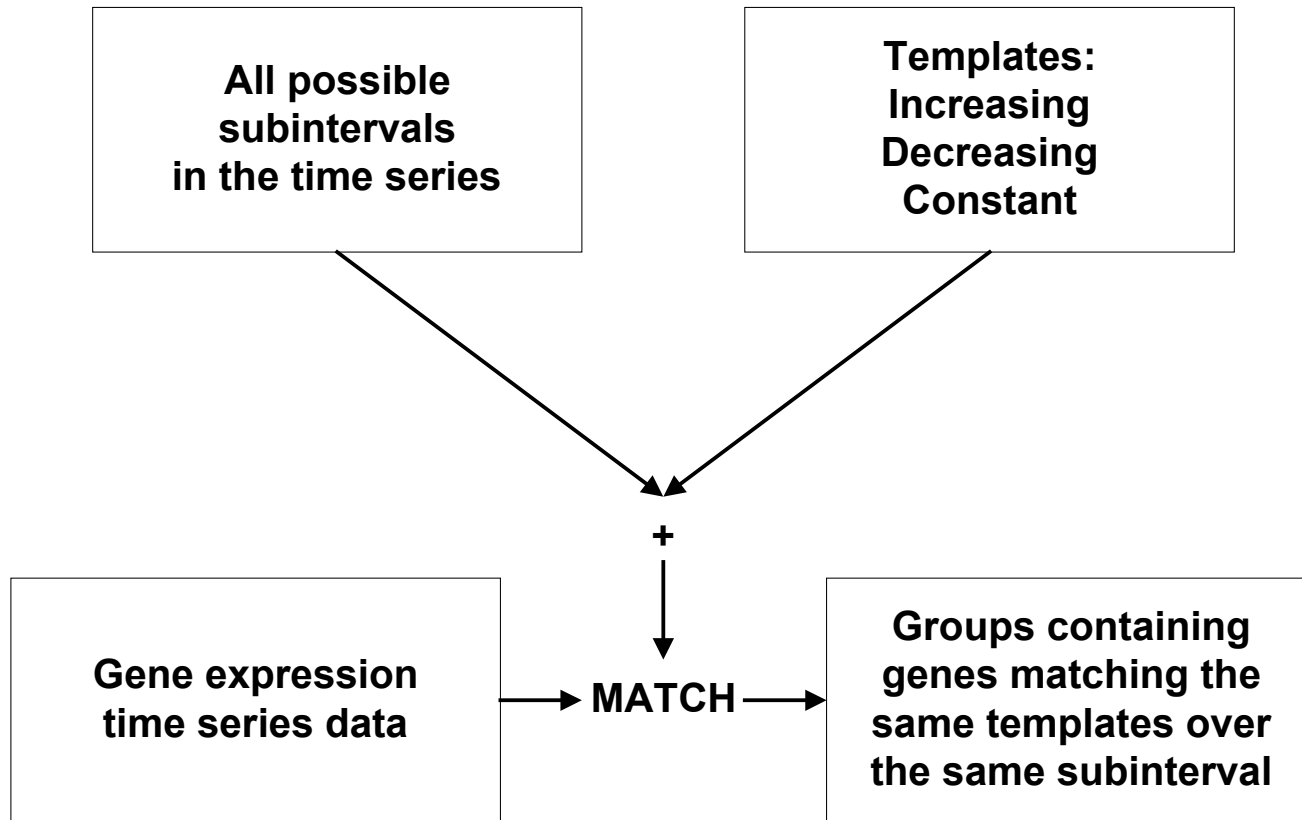
time profiles of selected processes



Gene Ontology vs. clusters



template-based feature synthesis



12 measurement points, 55 possible intervals of length >2

cross validation estimates

PROCESS	AUC	SE
Ion homeostasis	1.00	0.00
Protein targeting	0.99	0.03
Blood coagulation	0.96	0.08
DNA metabolism	0.94	0.09
Intracellular signaling cascade	0.94	0.06
Energy pathways	0.93	0.12
Cell cycle	0.93	0.04
Oncogenesis	0.92	0.11
Circulation	0.91	0.11
Cell death	0.90	0.10
Developmental processes	0.90	0.07
Transcription	0.88	0.11
Defense (immune) response	0.88	0.05
Cell adhesion	0.87	0.09
Stress response	0.86	0.15
Protein metabolism and modification	0.85	0.10
Cell motility	0.84	0.11
Cell surface rec linked signal transd	0.82	0.15
Lipid metabolism	0.81	0.14
Transport	0.79	0.17
Cell organization and biogenesis	0.79	0.11
Cell proliferation	0.79	0.06
Amino acid and derivative metabolism	0.69	0.06
AVERAGE	0.88	0.09

A:

Coverage: 84%

Precision: 50%

B:

Coverage: 71%

Precision: 60%

C:

Coverage: 39%

Precision: 90%

Coverage = $TP/(TP+FN)$

Precision = $TP/(TP+FP)$

the model

Annotations, Rules and Classifications

Annotated genes		273
within the 23 broad classes of GO biological process		
Gene probes		284
associated with the 273 genes within the 23 broad biological process classes		
Training examples		
annotations associated with the genes in the 23 broad biological process classes	549	
co-annotations associated with the genes in the 23 broad biological process classes	444	
Rules generated from the training examples		18064
Estimated quality of classifications of unknown genes (cross-validation estimates)		
Sensitivity		84%
Specificity		91%
Fraction of classifications that are correct		49%
Classifications for unknown (uncharacterized) genes		548
classifications were obtained for 211 of the 213 unknown genes		
(Re-)Classifications for training examples		728
True positive classifications	519	
True positive co-classifications	356	
False positive classifications	219	
False negative (missing) classifications	30	



For 272 of the 273 training examples at least one correct (re-)classification was obtained

conclusions

- ▶ our methodology
 - ▶ incorporates background biological knowledge
 - ▶ handles well the noise and incompleteness in the microarray data
 - ▶ can be objectively evaluated
 - ▶ predicts multiple functions per gene
 - ▶ can re-classify known genes and provide possible new functions of the known genes
 - ▶ can provide hypotheses about the function of unknown genes

- ▶ experimental work needs to be done to confirm our predictions

Lægreid A, Hvidsten T, Midelfart H, Komorowski J, Sandvik AK.
 Predicting Gene Ontology Biological Process from Temporal Gene Expression Patterns.
Genome Research. 13: 965-979, 2003

Hvidsten TR, Lægreid A, Komorowski J.
 Learning rule-based models from gene expression time profiles annotated using Gene Ontology.
Bioinformatics, 19:1116-23, 2003

Genomic ROSETTA:

<http://www.idi.ntnu.no/~aleks/rosetta>

The screenshot displays the Rosetta software interface with several windows open. The main window shows a file explorer with a tree structure under 'fibroblast serum response data' containing 'cv-estimates.txt', 'rules.txt', and 'classification.txt'. Two smaller windows show the contents of 'rules.txt' and 'classification.txt'. The 'rules.txt' window contains the following text:

```

7 OMIN - 1H(Constant) AND OMIN - 4H(Constant) => Annotation(transport)
8 Supp. (LHS) = [4 object (<=) 1]
9 Supp. (RHS) = [3 object (<=) 1]
10 Acc. (RHS) = [0.75, 0.75]
11 Cov. (LHS) = [0.054794]
12 Cov. (RHS) = [0.166667]
13 Stab. (LHS) = [0]
14 Stab. (RHS) = [1, 1]
15
16 OMIN - 1H(Constant) AND OMIN - 4H(Constant) => Annotation(transport)
17 Supp. (LHS) = [4 object (<=) 1]
18 Supp. (RHS) = [3 object (<=) 1]
19 Acc. (RHS) = [0.75, 0.75]
20 Cov. (LHS) = [0.054794]
21 Cov. (RHS) = [0.166667]
22 Stab. (LHS) = [0]
23 Stab. (RHS) = [1, 1]

```

The 'classification.txt' window shows a list of gene annotations:

```

1 PREDICTIONS OF LABELLED DATA
2
3 GENE ANNOTATION
4
5 RANBP1 transport
6 FKBP1A transport
7 XP01 transport
8 NR2F2 transport
9 ABCA5 transport
10 ICFBP3 transport
11 AP0D transport
12 KIAA0789 transport
13 CLCN3 transport
14 STC1 transport
15 KPNA2 transport
16 SEPP1 transport

```

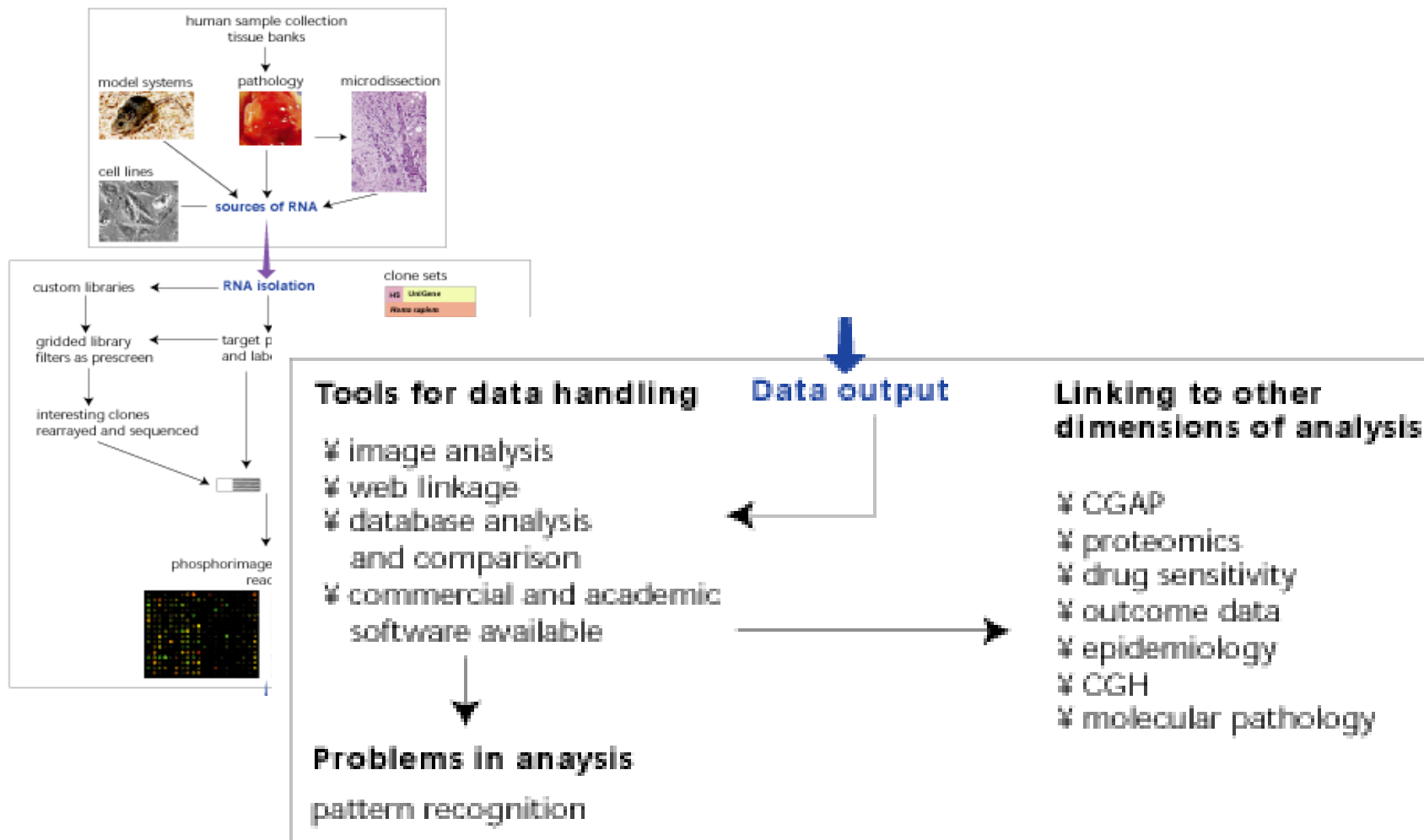
The 'Algorithm for Functional Genomics' dialog box is open, showing the following settings:

- Classification:**
 - Weight on false positives: 4
 - File with classifications: classifications.txt
 - File with rules: rules.txt
- Cross validation:**
 - Cross validation evaluation
 - File with cross validation: cv-estimates.txt
 - Number of cross validation iterations: 10
 - Seed to RNG: 5
- Template definitions:**
 - INCREASE/DECREASE TEMPLATES:
 - The increase/decrease required: 0.6
 - Required increase/decrease during the first and last sub-intervals: 0.1
 - Length of templates in terms of sub-intervals: 2
 - Maximum change in the "wrong" direction allowed: 0.2
 - CONSTANT TEMPLATE:
 - Maximum allowed variation from average: 0.2
 - Length of templates in terms of sub-intervals: 3
 - Supervised
- Genetic algorithm for reduct computation:**
 - IDG file
 - File with rules: D:\Rosetta\idg-not-symmetric.txt
 - Approximate reducts: 0.9

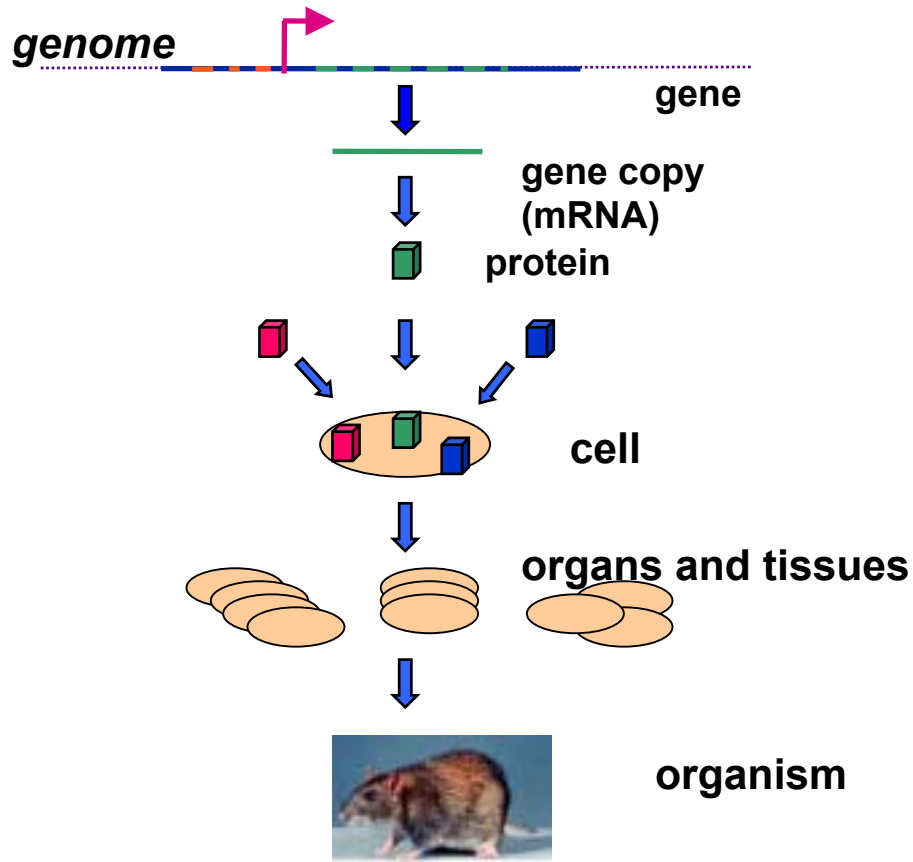
how to improve models for prediction of biological roles of genes/proteins?

- ▶ improved computational methods
- ▶ more training examples
 - ▶ more genes/proteins
 - ▶ more measurements per gene/protein (time points, cell types, tissues, states,...)
 - ▶ more annotations
(GO, sequence, protein structure, cell biology, physiology, pathology,...)

many levels of information



high complexity



~35.000 genes

> 100.000 gene (splice) products

> 100.000 proteins

> 200.000 protein states

each cell expresses

5.- 15.000 genes

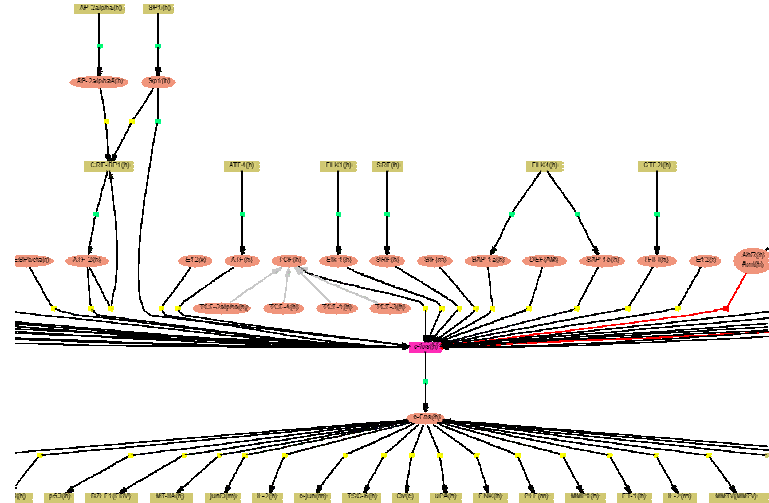
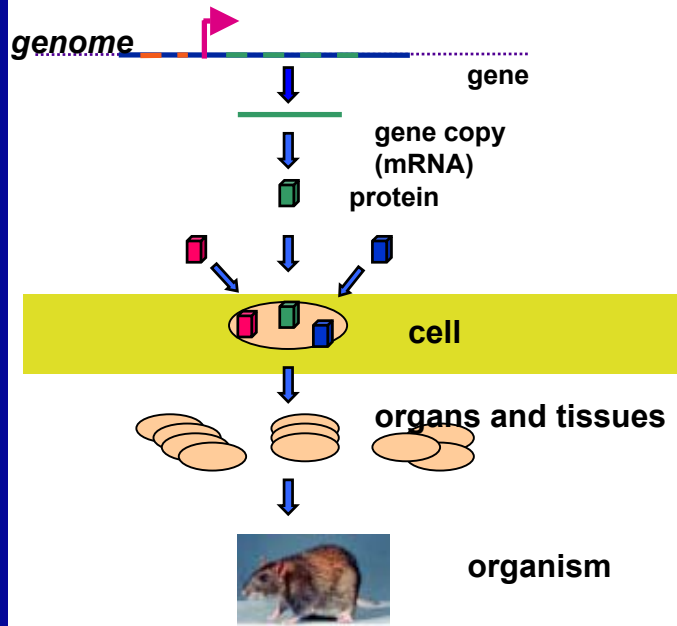
40.-60.000 proteins

several hundred cell types

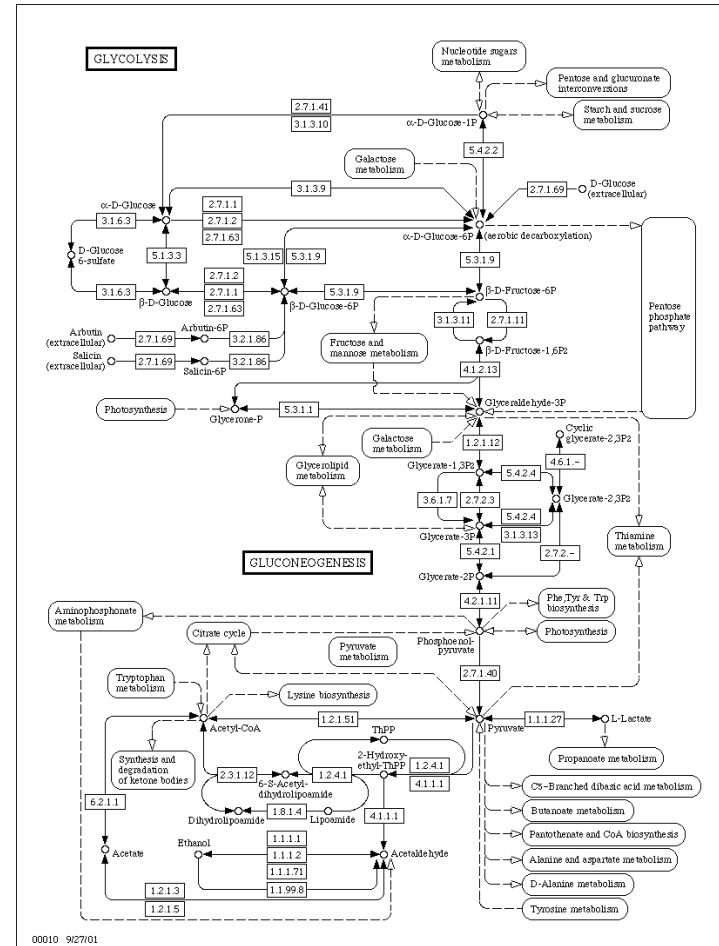
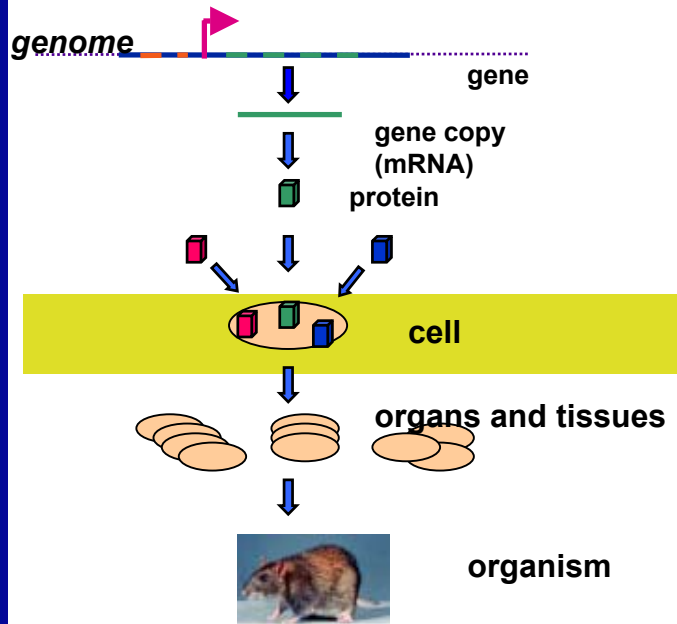
many different states per cell

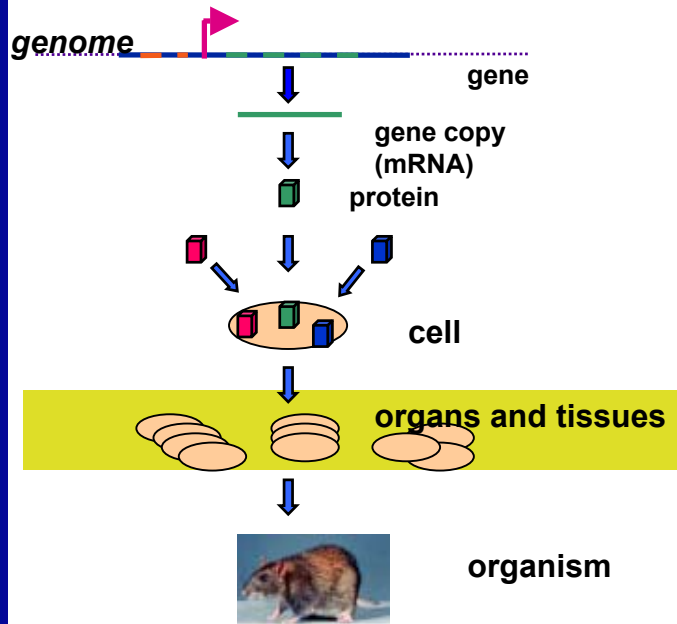
tissues and organs are
composed of many different cell
types

molecular networks within cells



molecular networks within cells

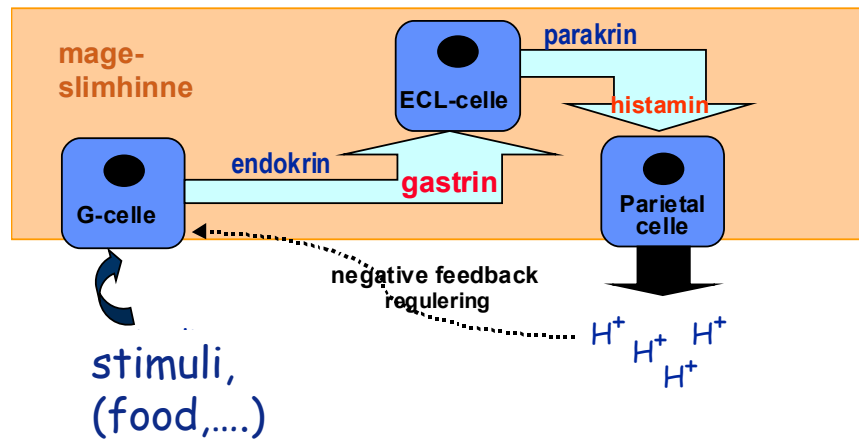


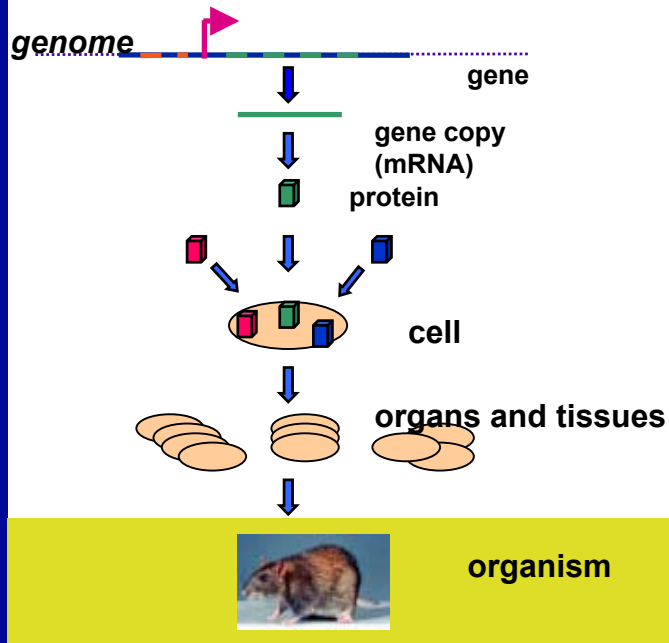


different cell types
interact within organs
and tissues

different cell types interact during gastric acid secretion

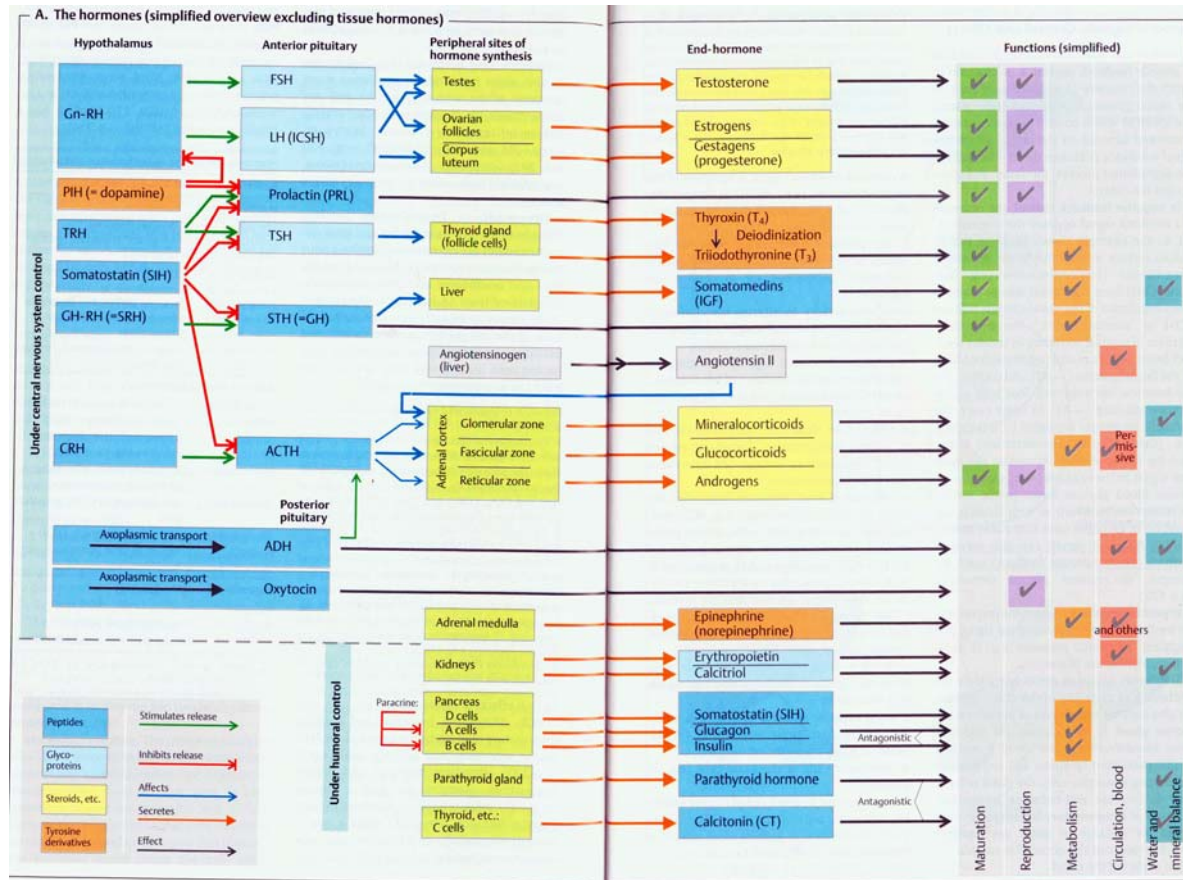
stomach mucosa





interconnection
within organism

hormones regulate interactions between organs and tissues



expression profiling in biology

determine molecular mechanisms underlying

- ▶ cell function related to cell type and state
- ▶ physiological functions of organisms

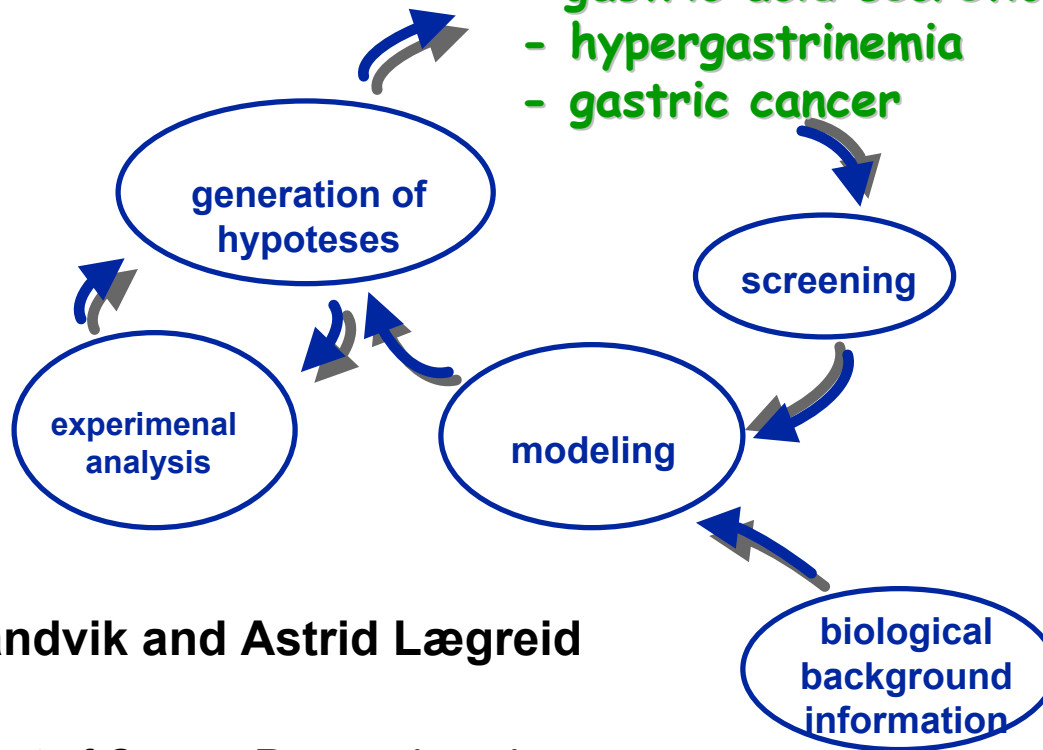
expression profiling in disease management

- ▶ discover disease subtypes
- ▶ improve disease diagnostics
- ▶ improve prognostics/choice of treatment
- ▶ discover new drug targets

our focus:

Molecular Mechanisms of the Normal and Diseased Gastrointestinal System

- gastric acid secretion
- hypergastrinemia
- gastric cancer



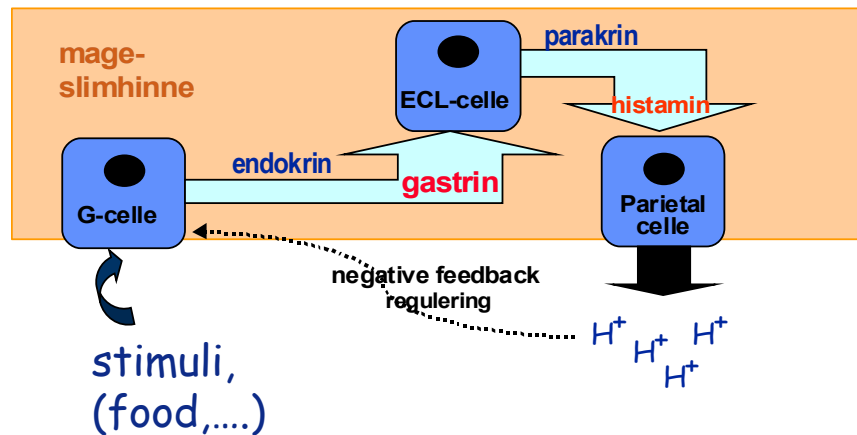
Arne Sandvik and Astrid Lægreid

Department of Cancer Research and Molecular Medicine, NTNU

gastrointestinal physiology and pathophysiology

- gastric acid secretion

stomach
mucosa



- molecular mechanisms?
- regulators, effectors?

gastrointestinal physiology and pathophysiology

- hypergastrinemia

- ↑ gastrin
- ↑ proliferation gastric mucosa
- ↑ ECL-cells
- ↑ cancer

- molecular mechanisms?
- regulators, effectors?

gastrointestinal physiology and pathophysiology

- gastric cancer

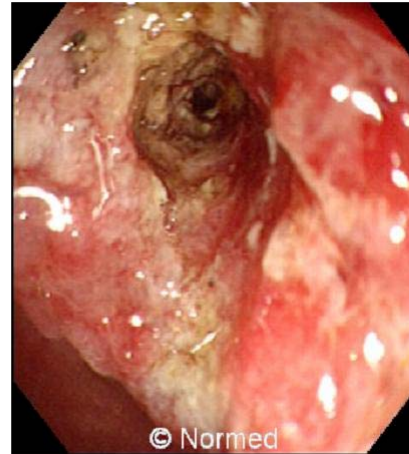
- classification & prediction

subtype diagnostics

prognostics

optimal treatment

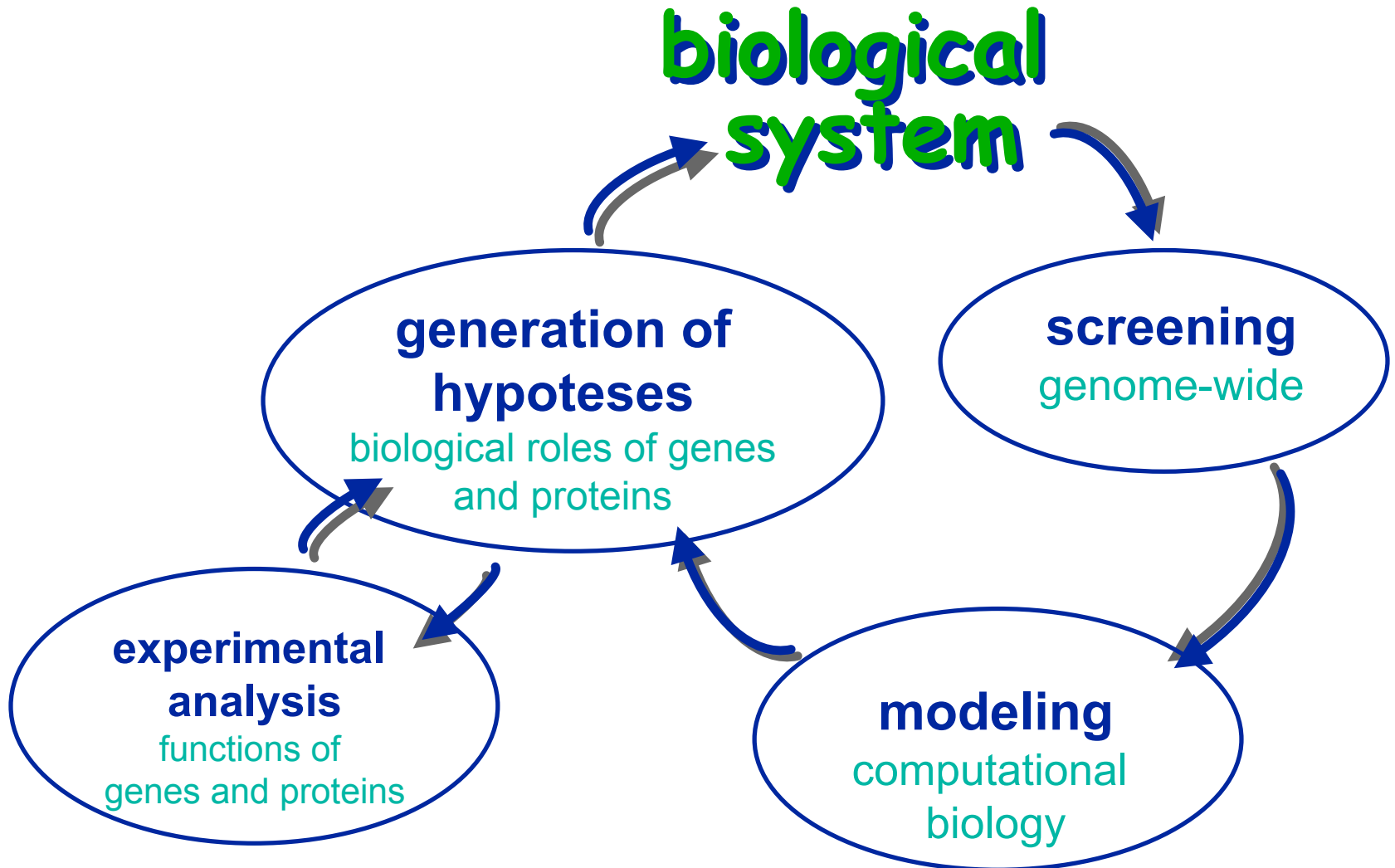
early diagnostics

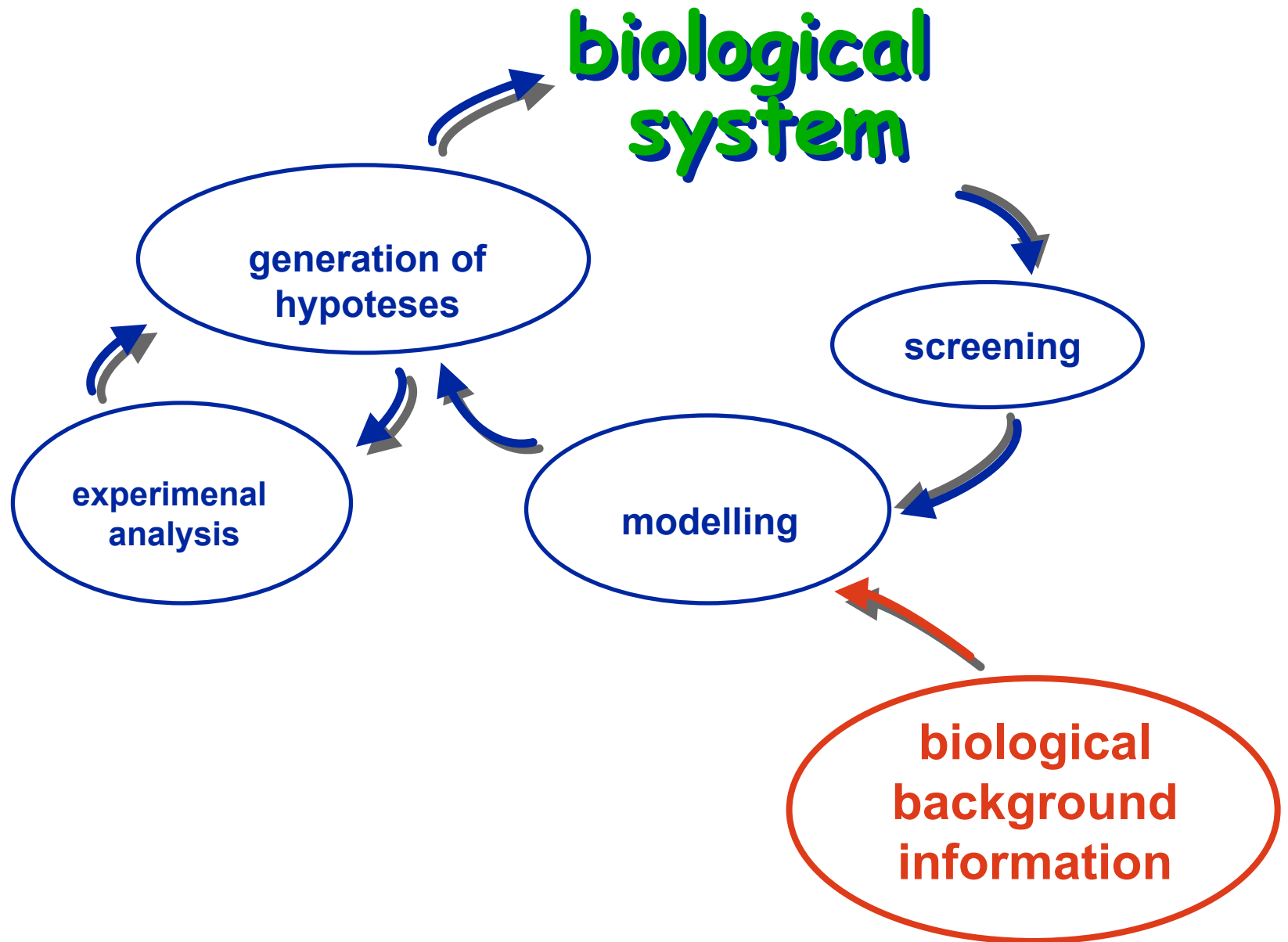


- molecular mechanisms?

- regulators, effectors?

functional genomics





challenges....

Information Bases/Derived-Data Databases

Experimental/Clinical Data

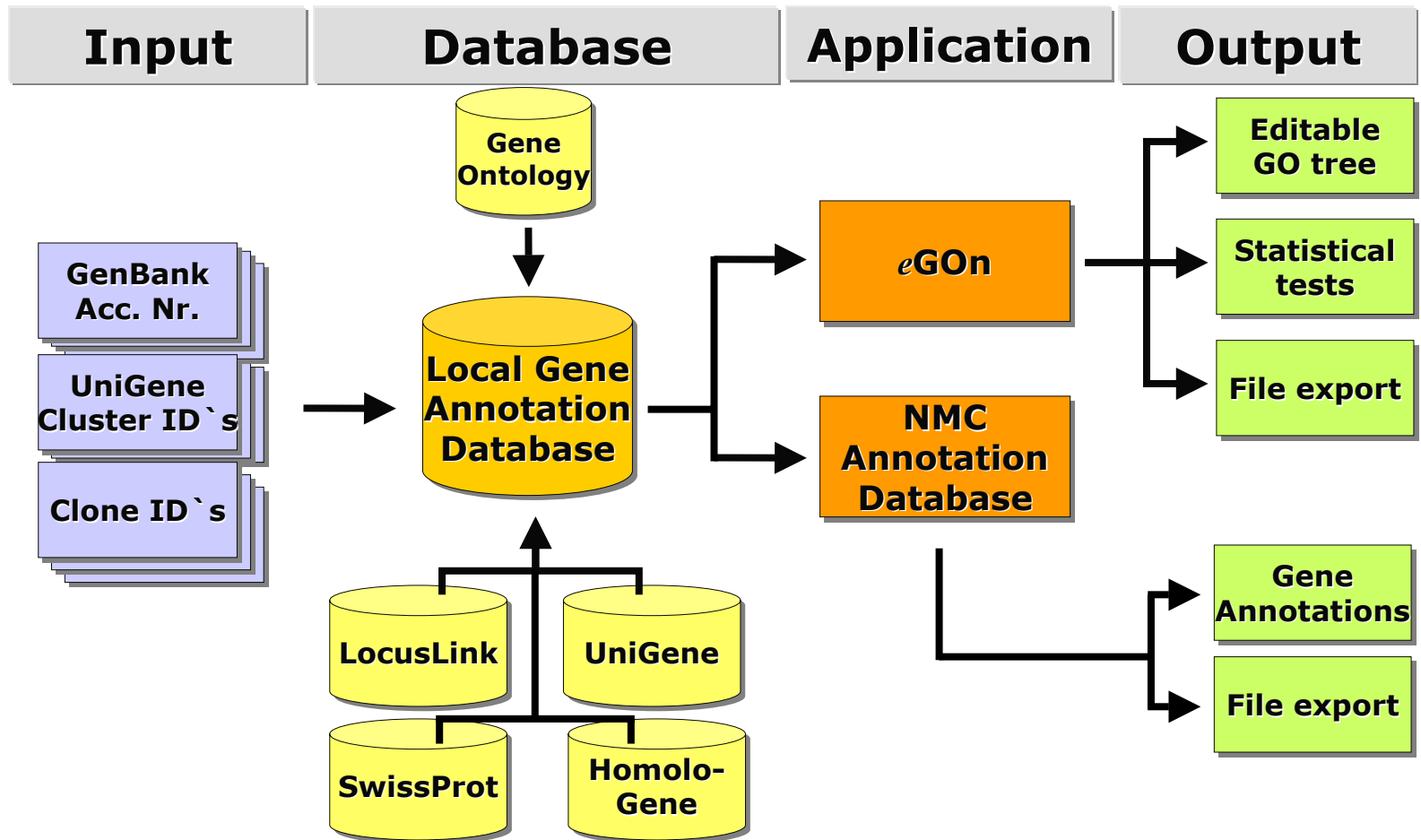
challenges....

Information Bases/Derived-Data Databases

Experimental/Clinical Data

**link information from various sources
in a relevant way**

relational database & tools at NTNU



challenges....

Information Bases/Derived-Data Databases

Experimental/Clinical Data

**mine information from unstructured
information sources**

mining the literature

PubGene Gene Database and Tools - Mozilla

File Edit View Go Bookmarks Tools Window Help

http://www.pubgene.org/ Search

Home Bookmarks The Mozilla Cr... Latest Builds

UNIVERSITAS OSLOENSIS
1811

The PubGene™ Web Tools are developed by PubGene Inc.
This installation of PubGene™ is hosted by University of Oslo
License for PubGene™ 2.2

PubGene

Expression Analysis & Text Association	Text & Sequence Homology Networks	Literature Search	Pathway Mapping	Supplementary
DNA Microarray Data	Network Browser Subset Network	Boolean Search Set Cover Search	Mesh Associations Ontology Associations	Gene Name Search Clone Mapping

PubGene™ Gene Database and Tools

Expression Analysis and Text Association

DNA Microarray Data Analyze gene expression data with literature network information

Text and Sequence Homology Networks

Network Browser Browse literature or sequence neighbors of a given gene

Subset Network Browse literature or sequence networks for a given set of genes

Literature Search

Boolean Search Search (Boolean) literature articles for a set of genes

Set Cover Search Search (Set Cover) literature articles for a set of genes

Pathway Mapping

Mesh Associations Search MeSH term associations for a set of genes

Ontology Associations Search ontology term associations for a set of genes

Supplementary

Gene Name Search Find PubGene gene symbols in the PubGene nomenclature database

Clone Mapping Look up gene symbols in batch from file by clone (sequence) ID

PubGene Gene Database and Tools, Copyright © 2000-2003 PubGene Inc.
www.pubgene.com
To report problems with this mirror: hpc-core@usit.uio.no

Tor-Kristian Jenssen, Astrid Lægreid, Jan Komorowski, Eivind Hovig.
A literature network of human genes for high throughput gene-expression analysis.
Nature Genetics, 28: 21-28

mining the literature

(at NTNU)

statistical methods

machine learning

natural language processing

challenges....

Information Bases/Derived-Data Databases

Experimental/Clinical Data

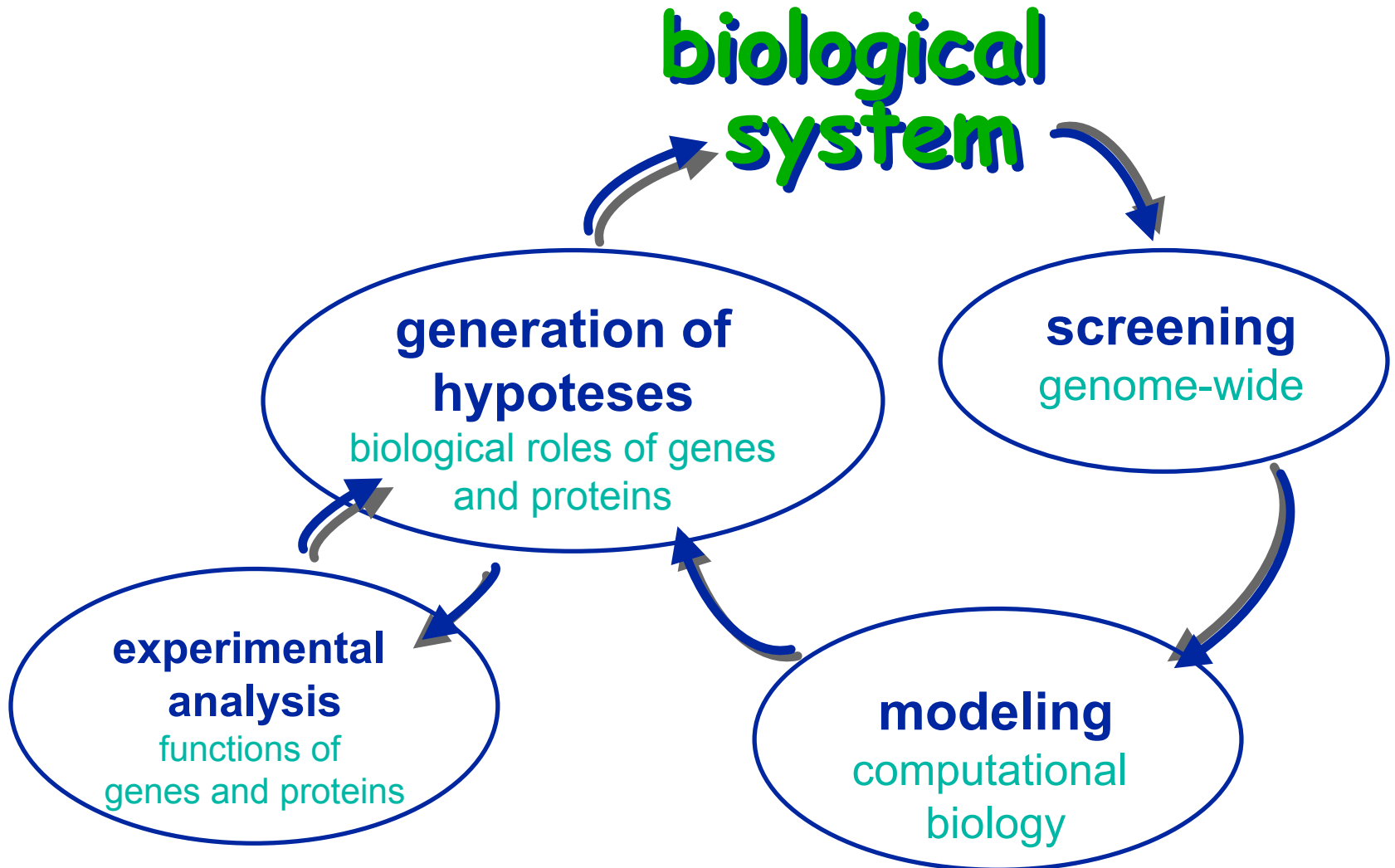
develop improved methods for modeling

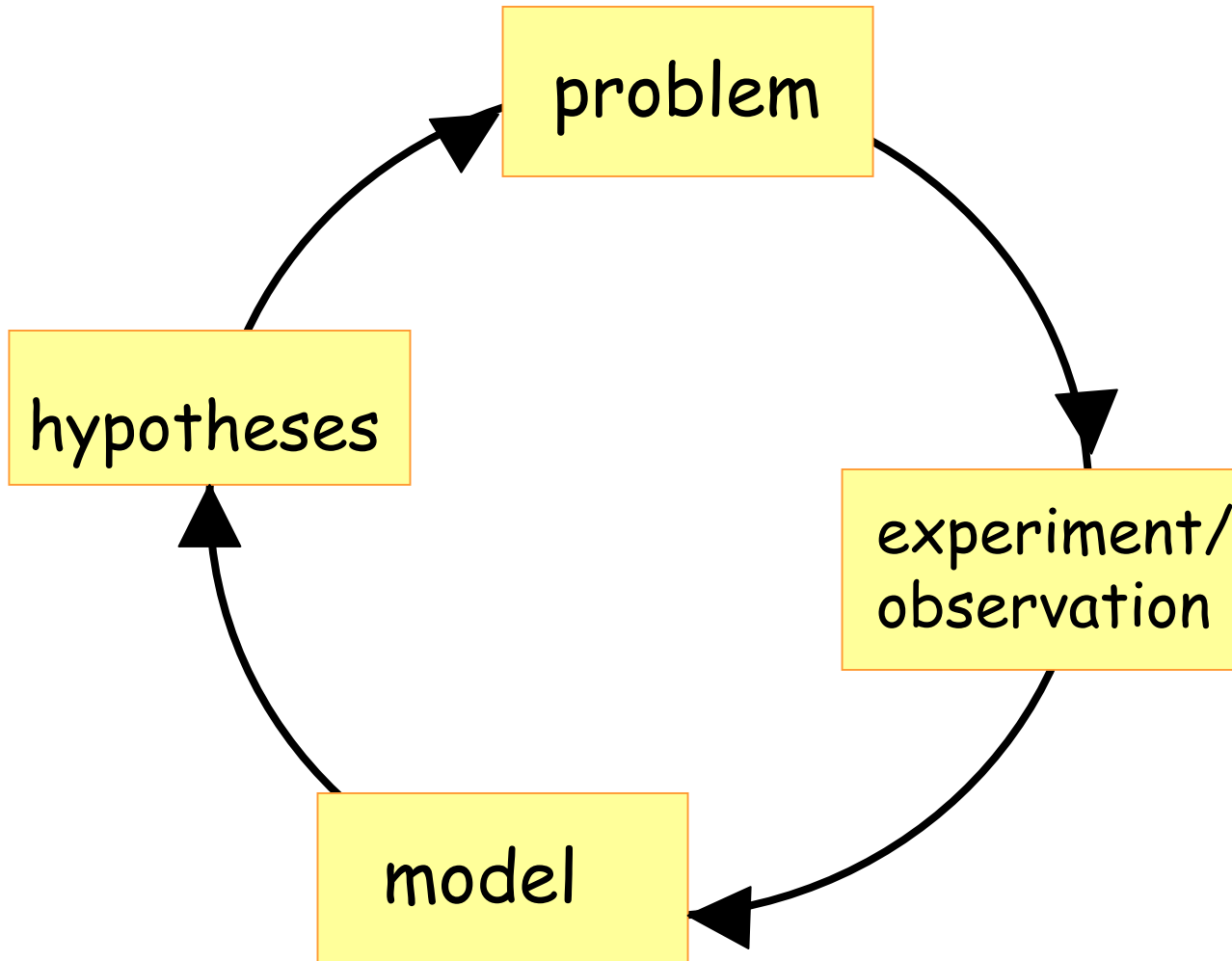
modeling

data driven

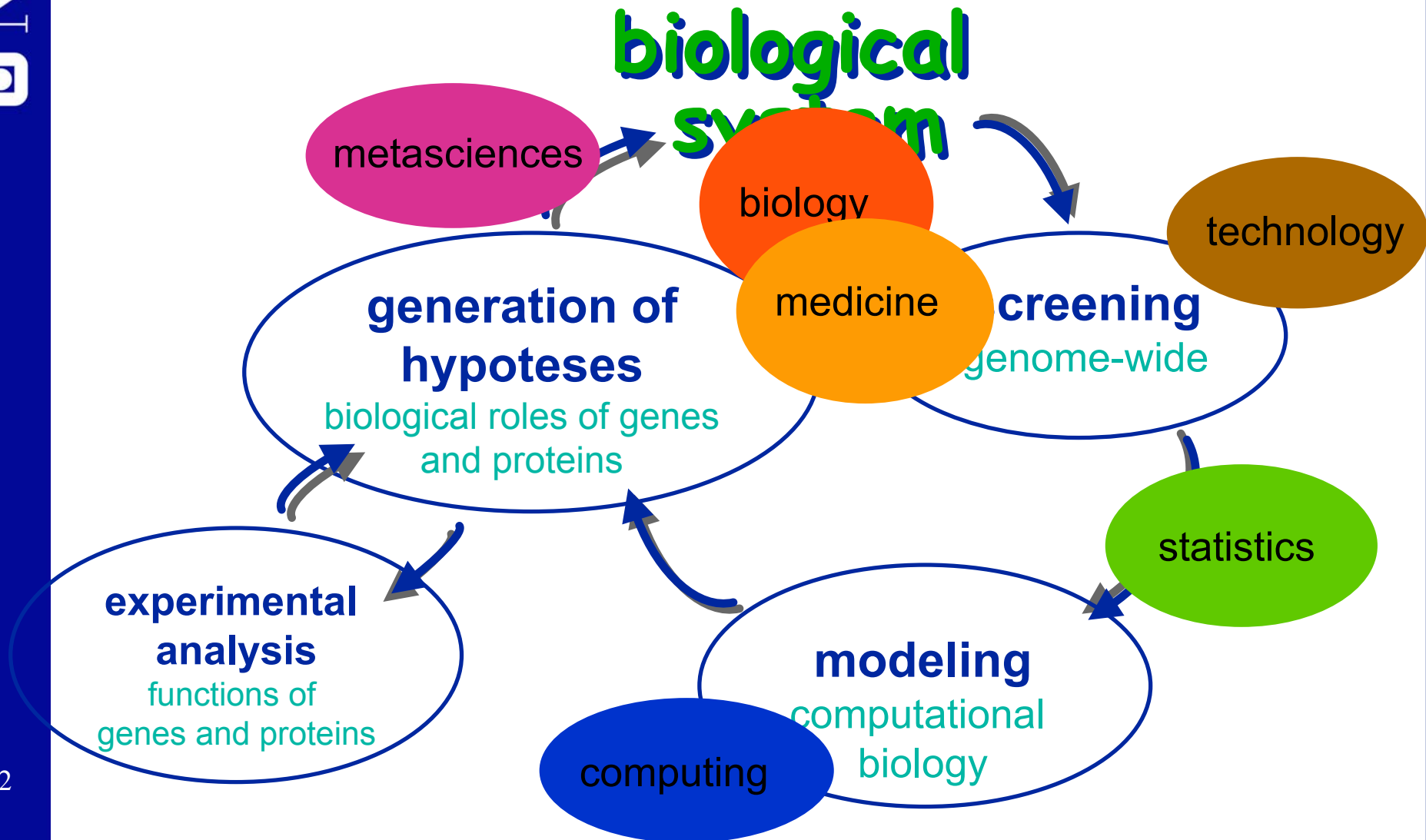
first principles

functional genomics





multi-disciplinary effort



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Kristin Nørsett
Vidar Beisvåg
Berit Doseith Eitrem
Hallgeir Bergum
Frode Jünge
Torunn Bruland
Ola Ween
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Pauline Haddow
Gunnar Tufte
Kjell Bratbergsengen
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Tore Amble
Rune Sætre
Bjørn Alsberg
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Lars Giskehaug
Torulf Mollestad
Henrik Tveit

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Torgeir Hvidsten
Herman Midelfart
Vladimir Yankovski