

# Bioinformatics in the Health Sciences:

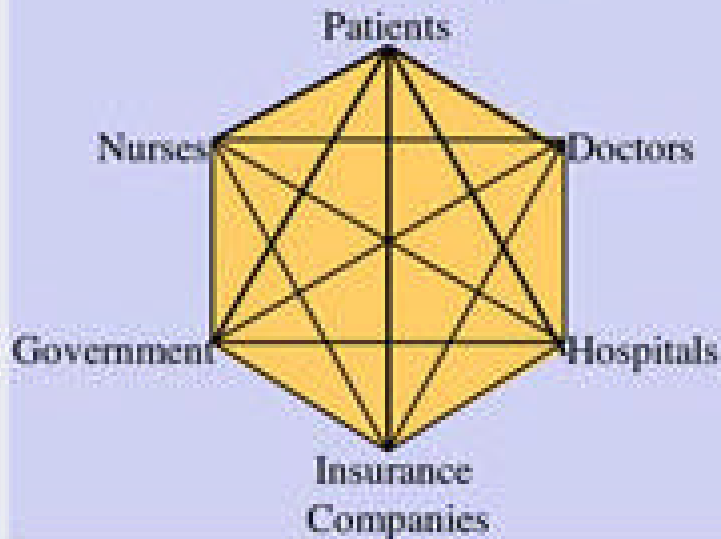
Towards tailored medicine?

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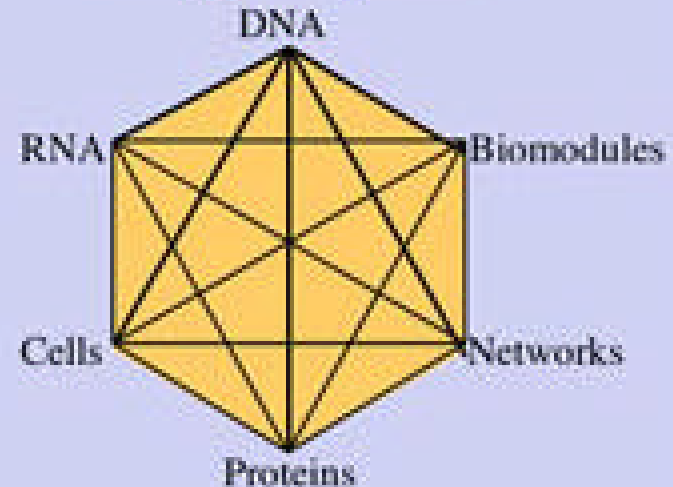
## Outline:

- ▶ health informatics and bioinformatics
- ▶ systems biology approaches
- ▶ bioinformatics technologies in health sciences
- ▶ biomarkers and diagnosis
- ▶ combinatoric therapeutics

### Healthcare System

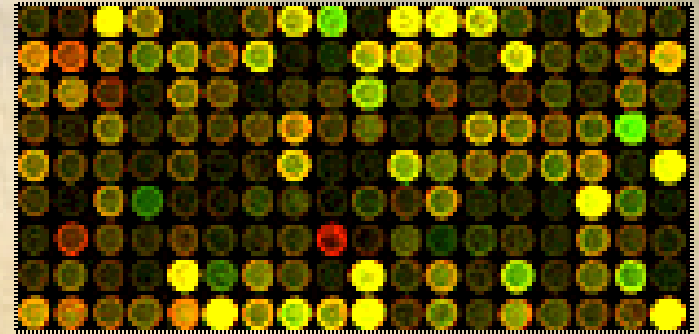
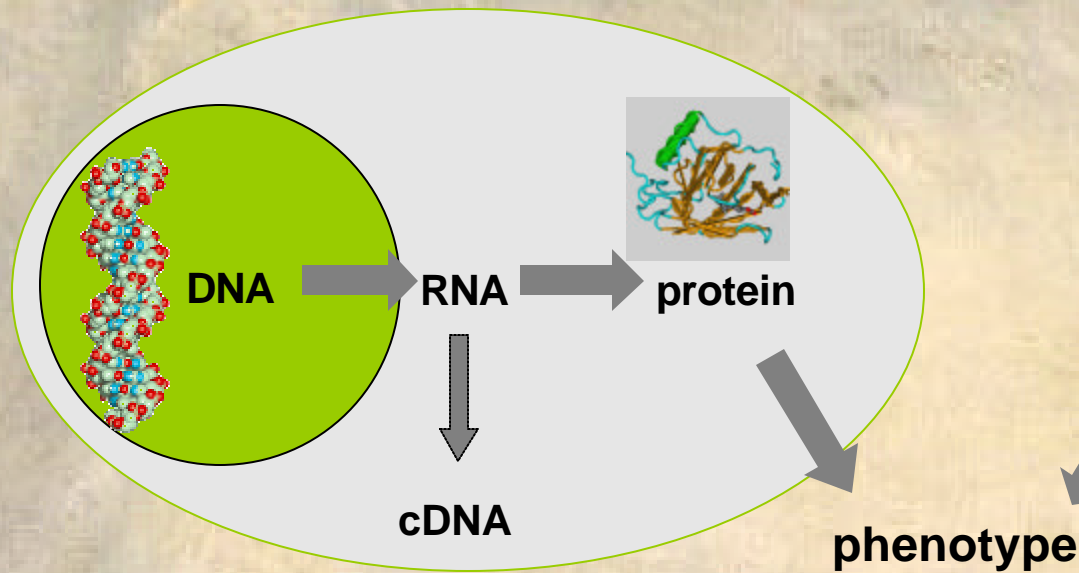


### Biological System



[www.systemsbiology.org](http://www.systemsbiology.org)

Bioinformatics and Health Informatics are both concerned with data management, and interactions between components of the system



### hypothesis driven research

focus on one gene  
or one protein

determine relations  
between genes/proteins

integrate components,  
describe effect on phenotype

### discovery based research

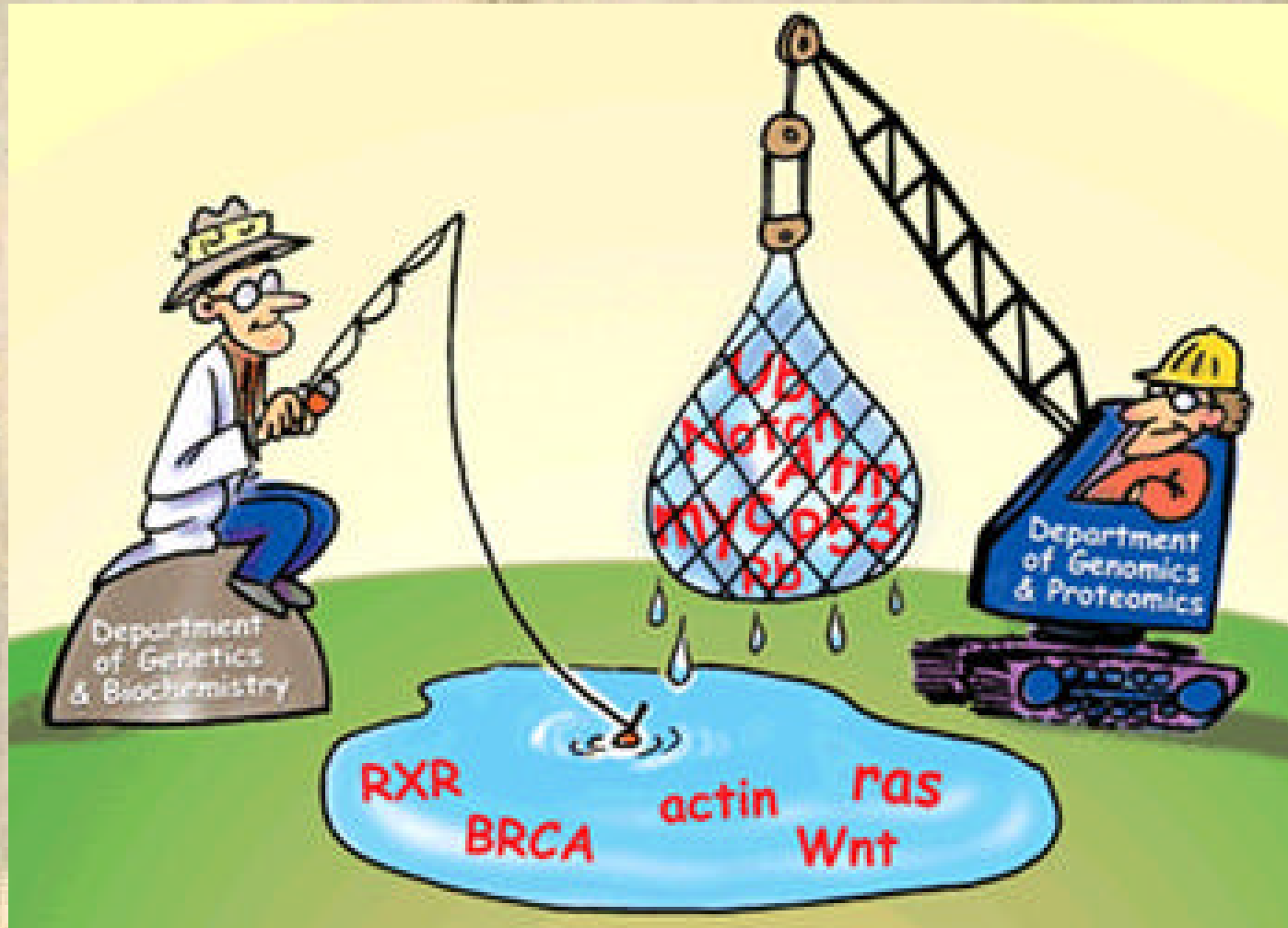
enumerate parts of a system  
(gene or protein expression)

identify patterns in data,  
relations between components

relate patterns to phenotype



## The optimistic view of genomics and proteomics:



*Science* (2001) **291** :1221

... but can you tell a fish from a rubber boot?

# The promise of bioinformatics in the health sciences

- ▶ genetic profiling
  - ▶ identification of genetic predispositions
- ▶ prognosis and treatment
  - ▶ prediction of response to treatment
  - ▶ tailoring treatment by tissue subtype
- ▶ diagnostics - early detection of disease
  - ▶ serum protein biomarkers
- ▶ identification of novel drug targets
- ▶ application to multi-factor disease



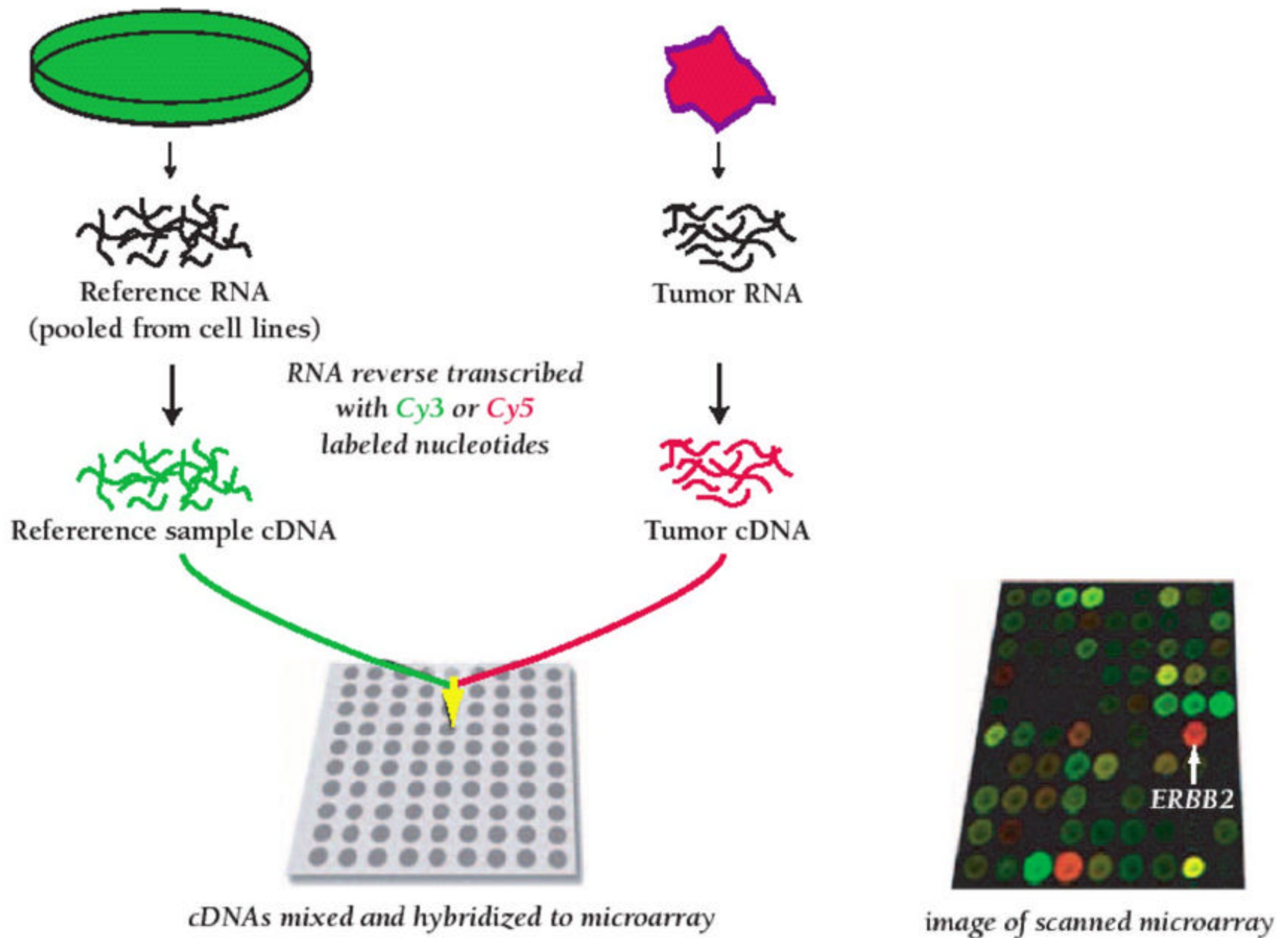
## The down side

- ▶ highly dimensional data sets
  - ▶ e.g. two treatments,  $>10,000$  genes
- ▶ data analysis - what does it all mean?
- ▶ 'black-box' approaches
- ▶ isolating cause and effect
- ▶ data management
- ▶ currently, cost is often high
  - ▶ expensive equipment and/or consumables

## Examples:

- ▶ microarrays in the diagnosis and treatment of breast cancer
- ▶ biomarkers of disease
  - ▶ serum analysis
- ▶ identification of drug targets
- ▶ cell cycle modeling - controls on cell division





Jeffrey SS, Fero MJ, Borresen-Dale AL, Botstein D. Expression array technology in the diagnosis and treatment of breast cancer. Mol Interv. 2002 Apr;2(2):101-9.

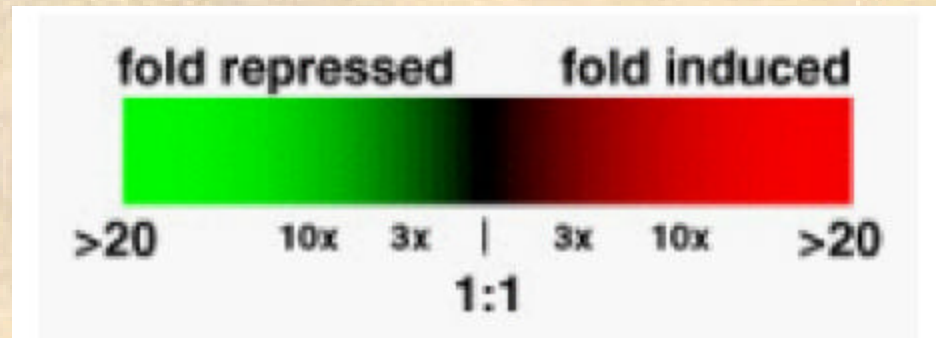
# Microarray setup

Stanford (Brown lab) microarrays

- unique Expressed Sequence Tag clusters for human cDNA
- >20000 ESTs represented
- typical distribution:
  - 40% annotated genes (UniGene)
  - 10% partly annotated
  - 50% little or no annotation
- requires 2-4 ug mRNA
- RNA sample often amplified
- standard reference RNAs available

# Microarray data handling

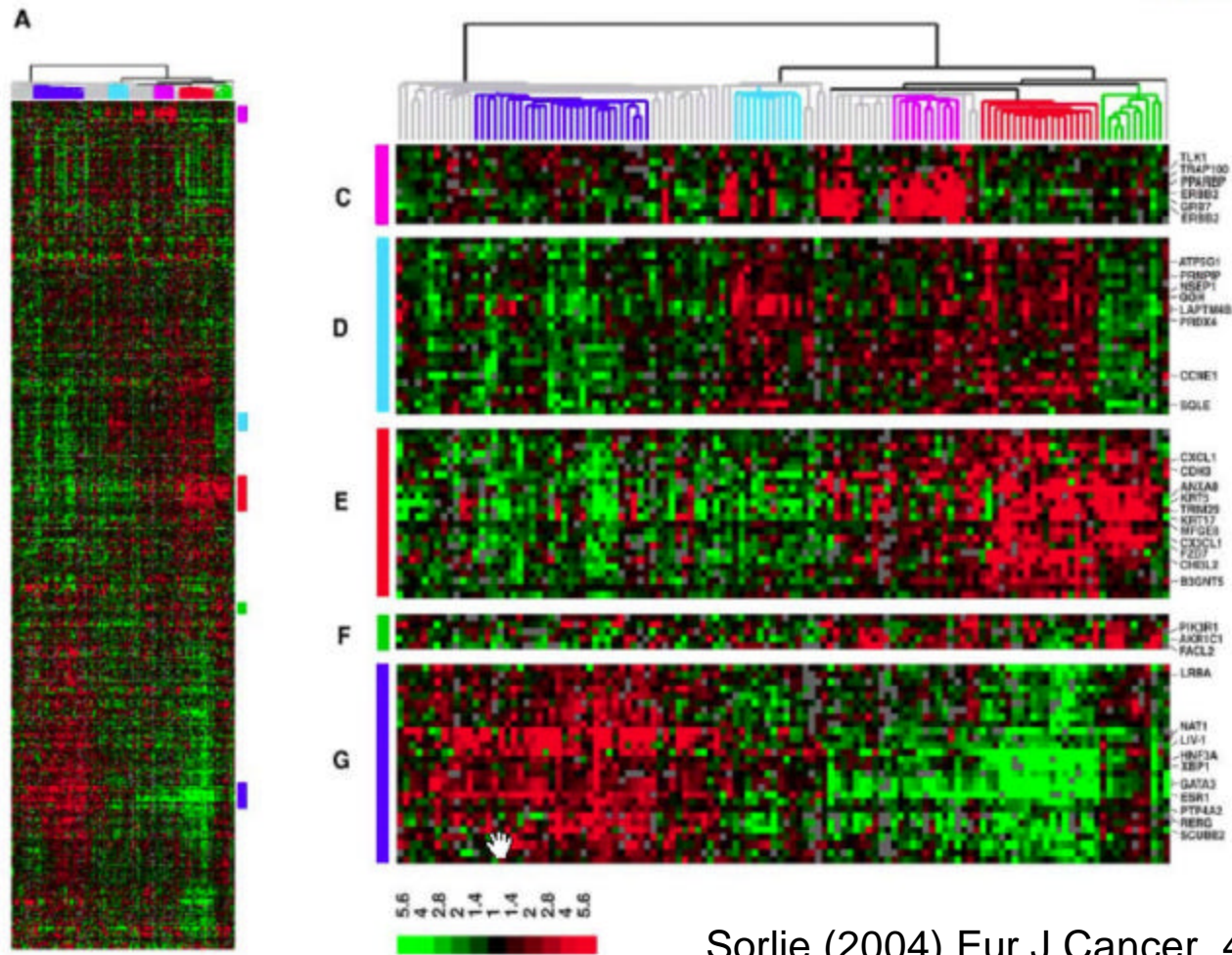
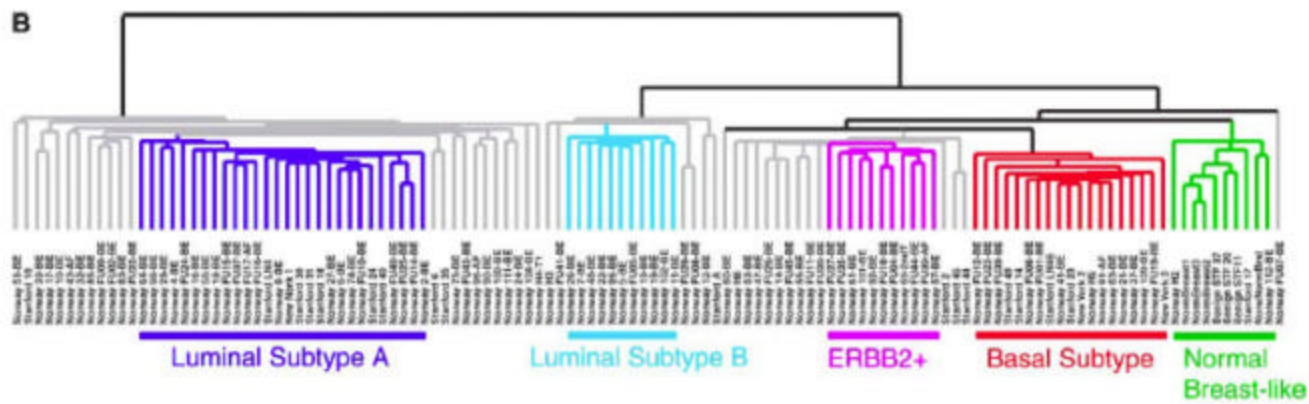
- each data point (spot on microarray) represents a change in expression level versus a reference sample  
(cDNA, sample#, ? expression)
- changes in expression ratios can be represented on a colour scale, to enable visualization of large data sets



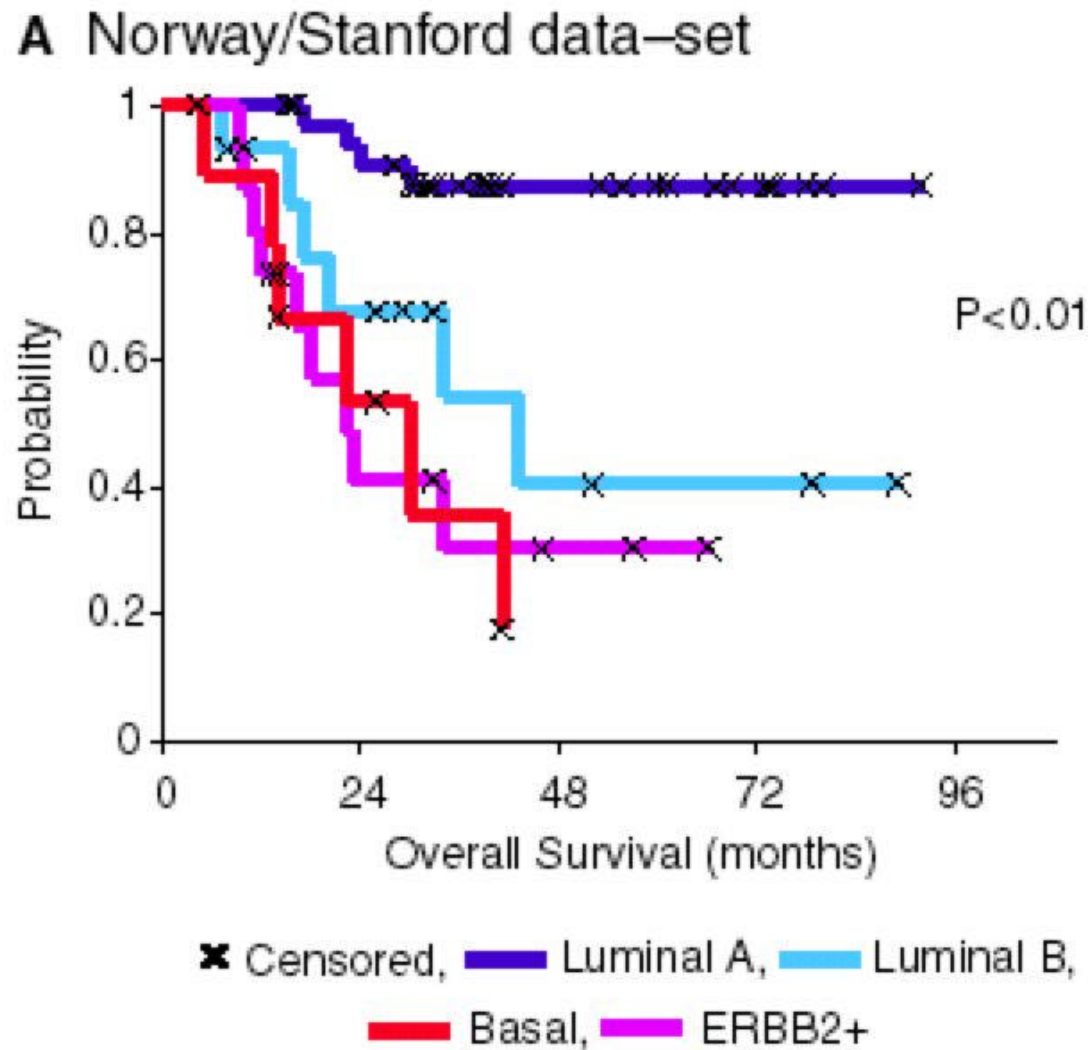
after Campbell and Heyer, 2002

- data points may be clustered by sample similarity and by expression similarity
- sample data set: 115 breast tumor tissues + 7 non-malignant tissues



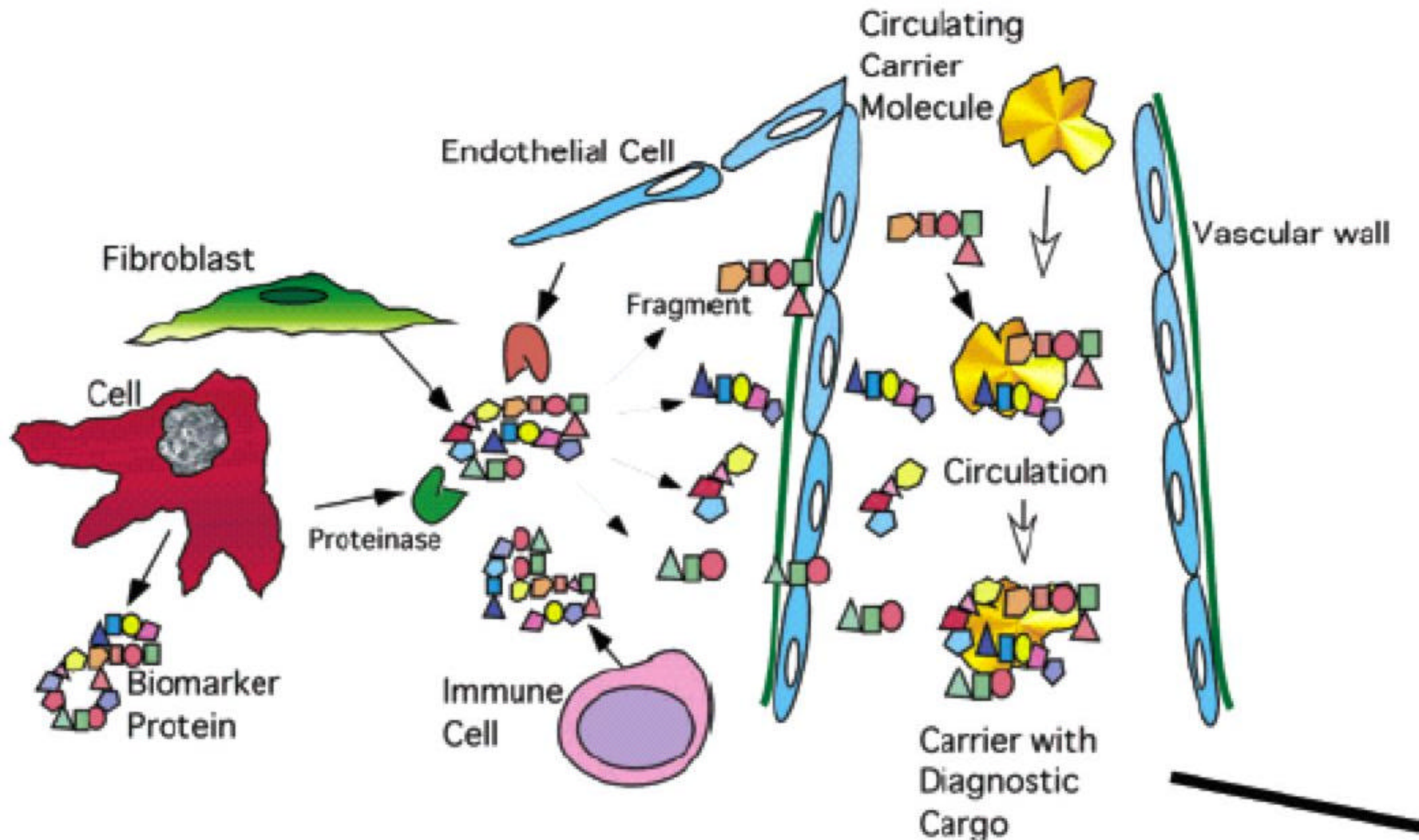


# Tumor subtype is highly correlated to survival

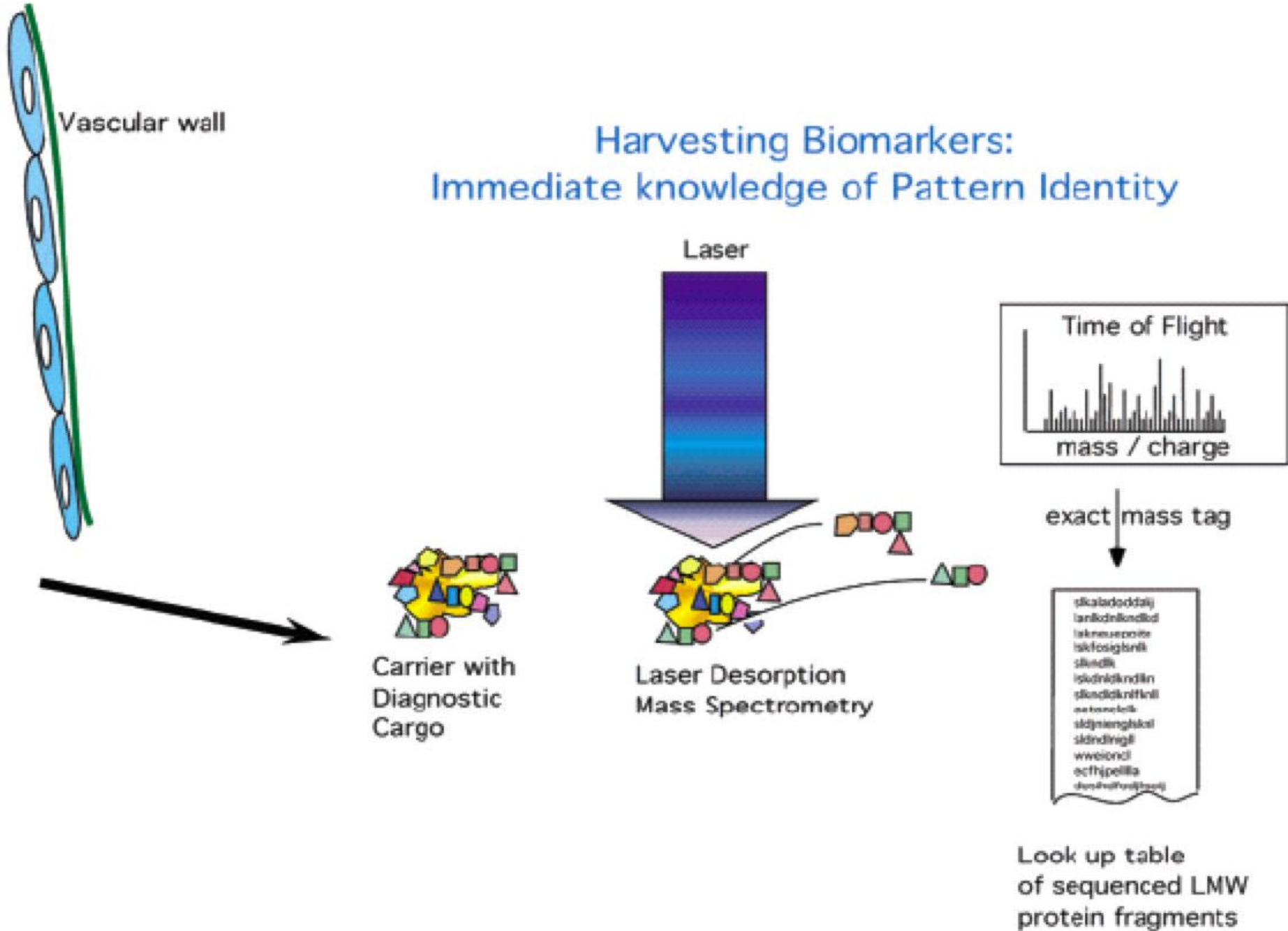


Sorlie (2004) Eur J Cancer. 40(18):2667-75.

# Biomarker Amplification and Harvesting by Carrier Molecules

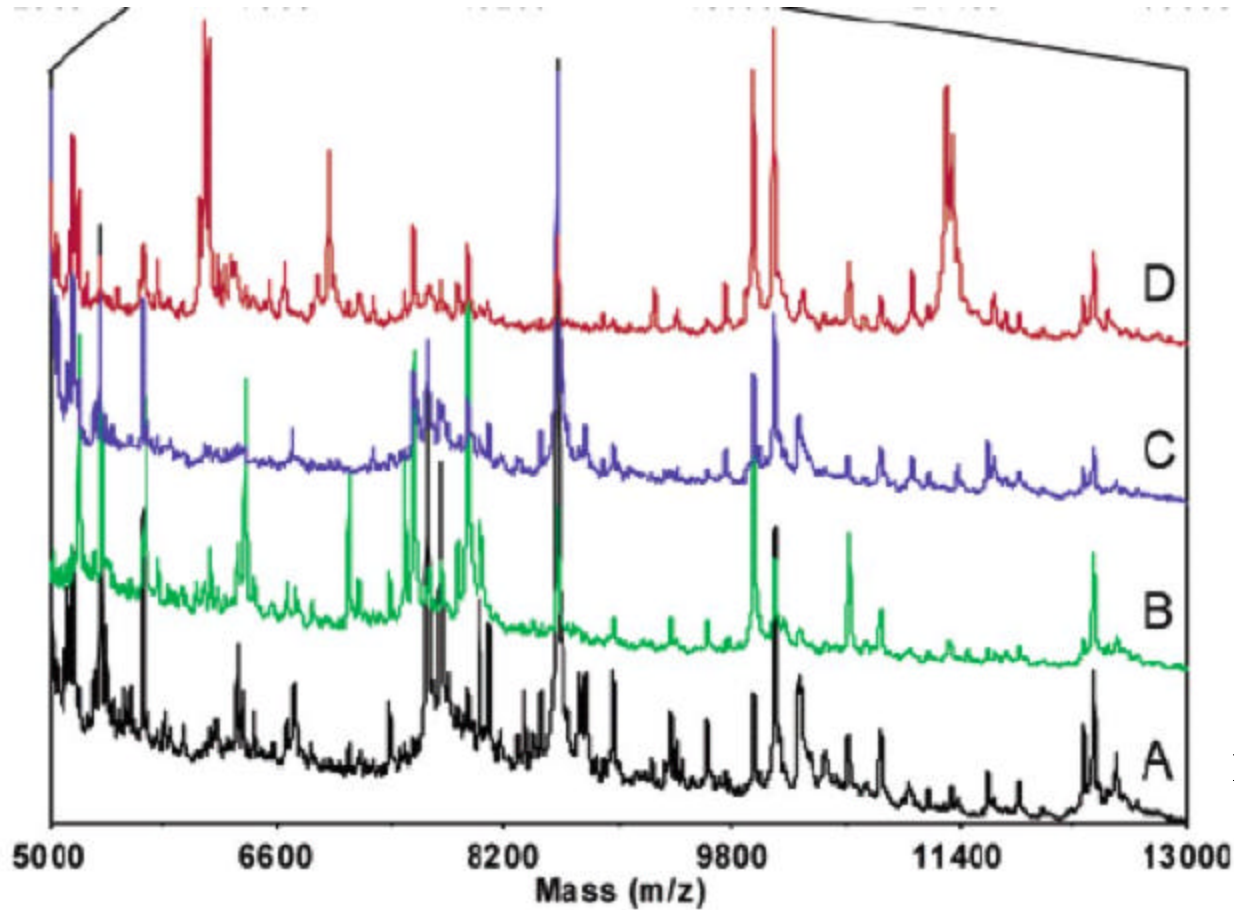






After Petricoin (2004) J Proteome Res.

# Mass spectrometry of brain tumor biopsy samples



D glioblastoma

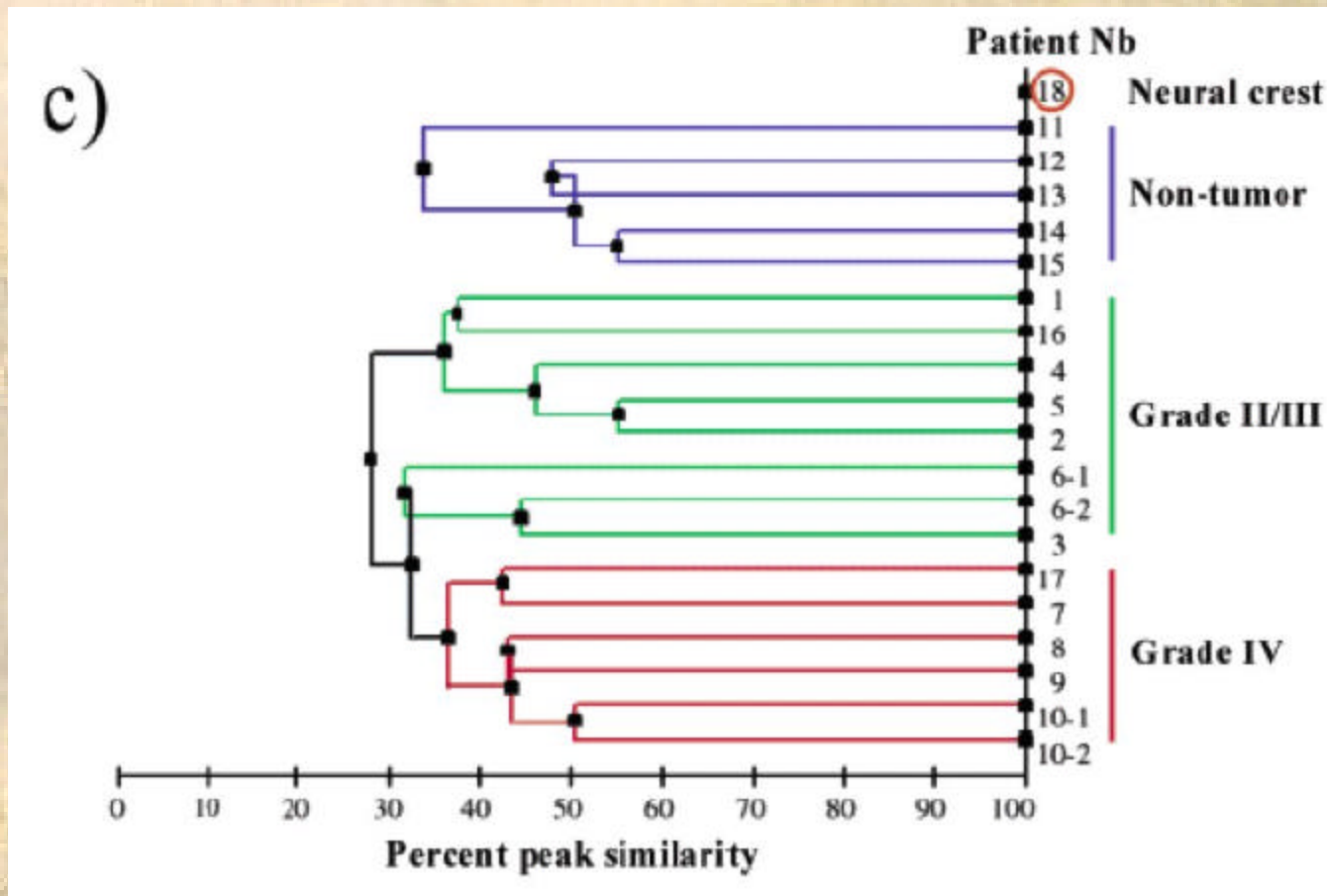
C grade III astrocytoma

B grade II astrocytoma

A normal cells

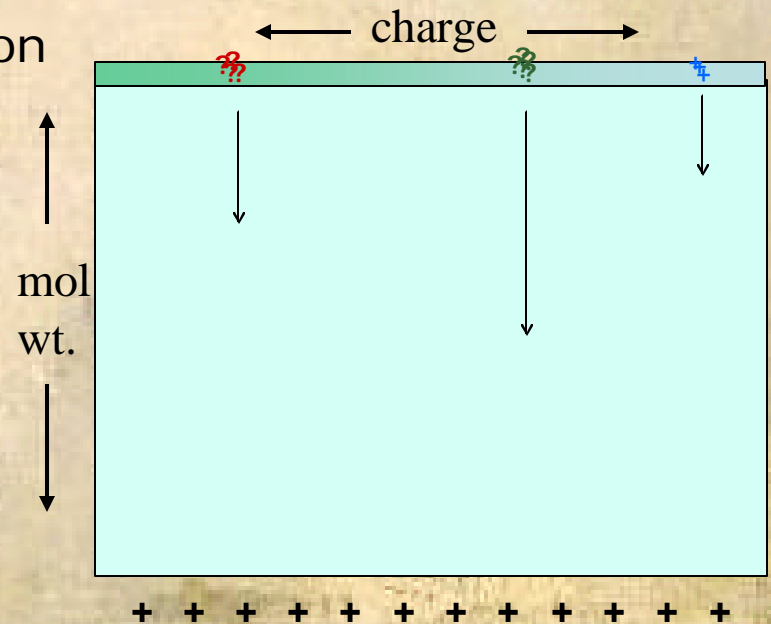
# Mass spectrometry of brain tumor biopsy samples

-classification from mass spectrometry pattern:

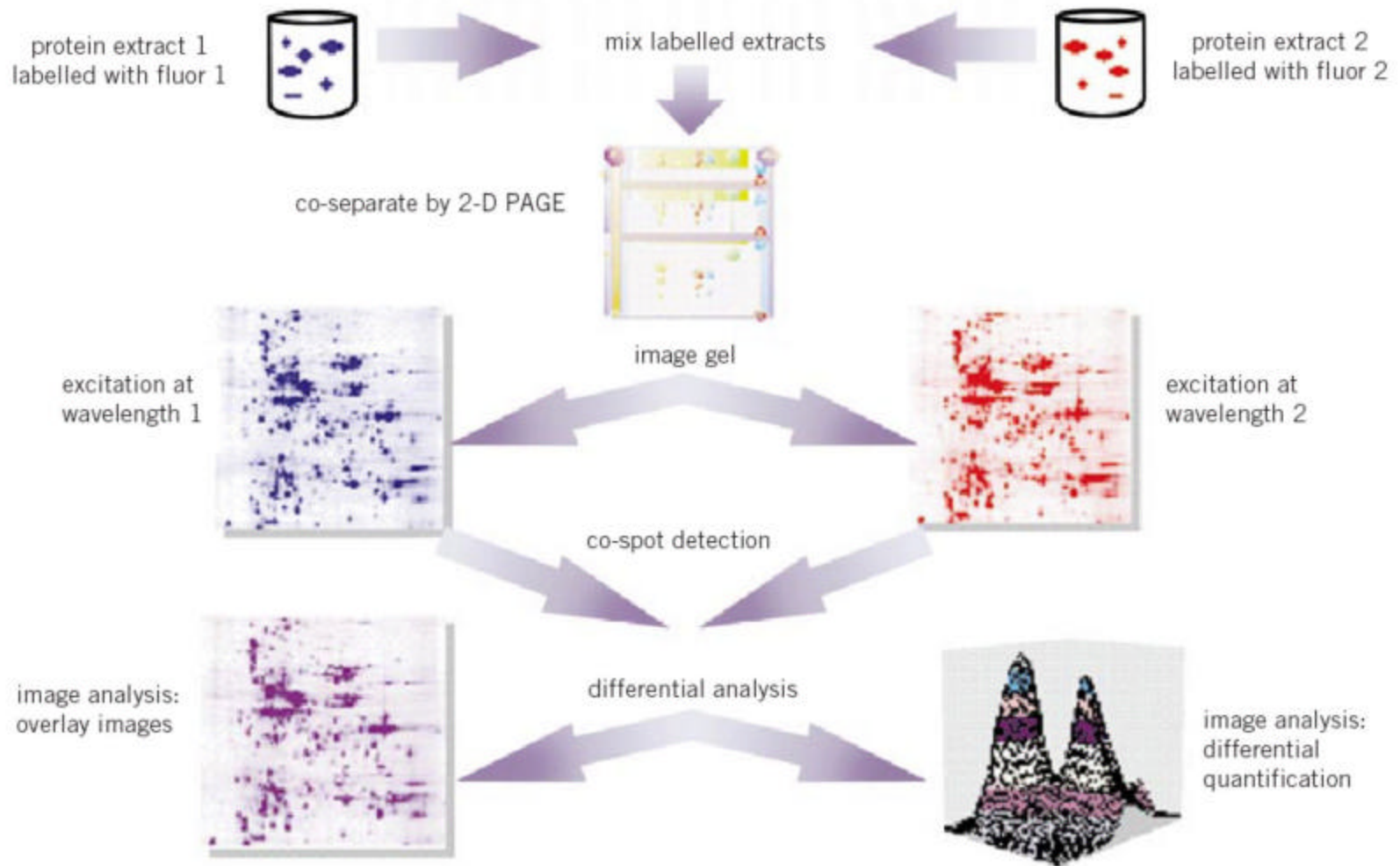


# Proteomics using 2-D electrophoresis

- separation in first dimension by charge, second dimension by molecular weight
- 1st dimension uses an immobilized pH gradient (e.g. pH 3-10)
- high voltage is gradually applied across gel, causing proteins to migrate to the location where they have zero net charge (Isoelectric Focussing - IEF)
- SDS-PAGE used for second dimension separation by M.W.
- proteins are visualized by staining - silver stain, fluorescent dyes, ...



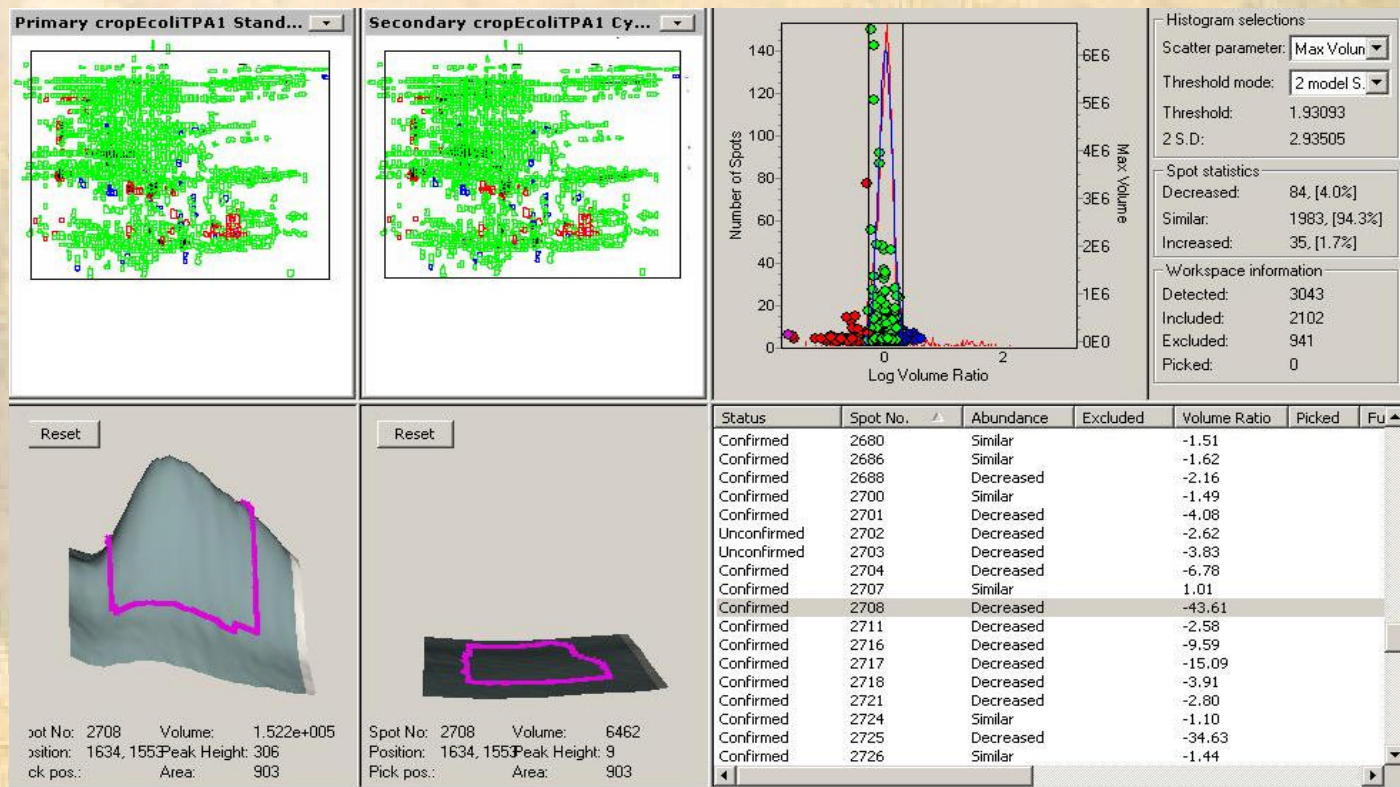
# Differential In-gel Electrophoresis



**Fig 1.** Outline of the 2-D DIGE technology. Protein samples are labelled with different fluorescent dyes, mixed, and co-separated by 2-D electrophoresis. Spots in the gel are visualized in the CCD-based imager and quantitatively analysed using 2-D analysis software. Spots showing quantitatively significant differences are then picked, digested, and analysed by mass spectrometry.



# Statistical analysis using DeCyder Software



Non-Induced Culture



# Advantages of DIGE

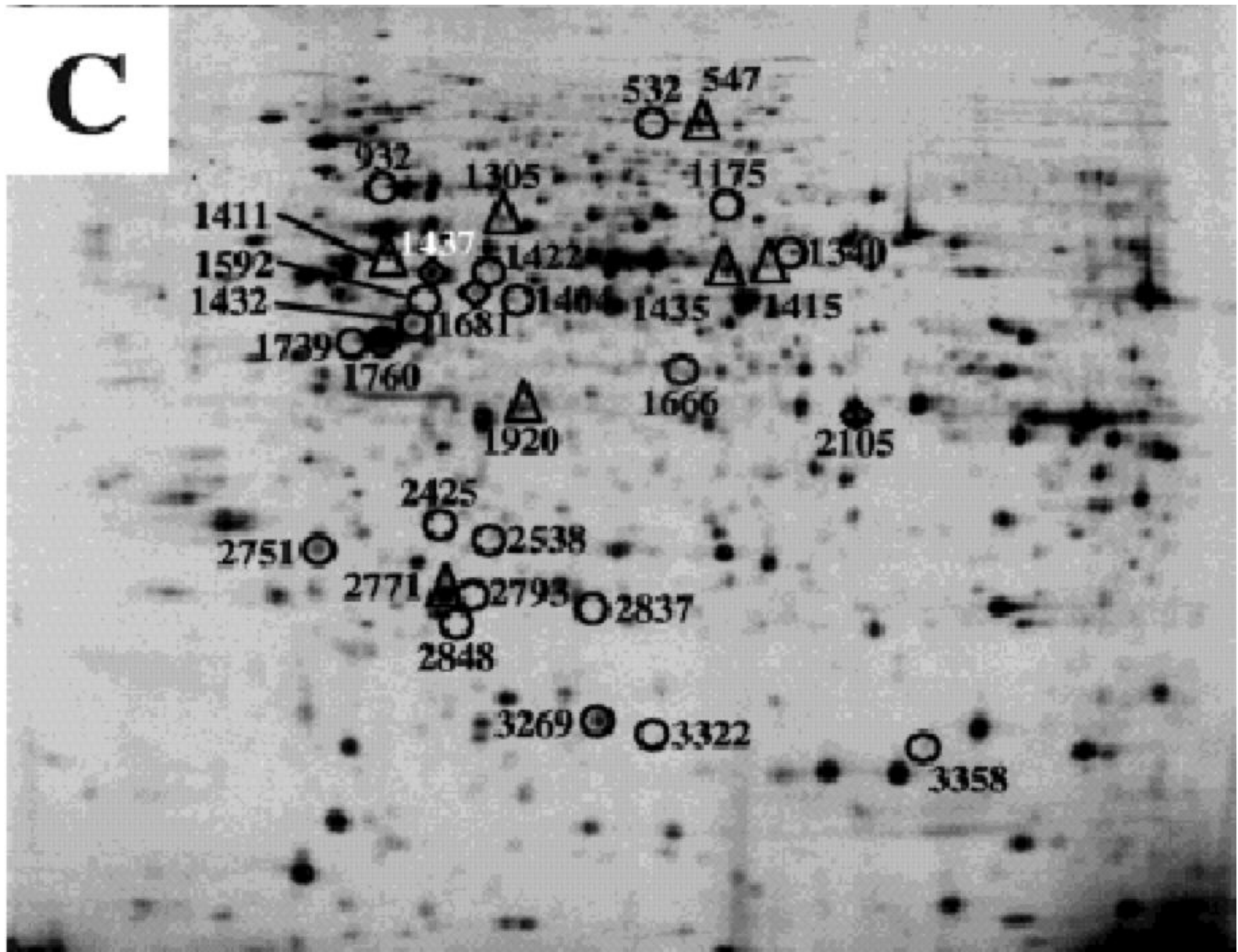
- multiplexed samples in the same gel
- Spot matching of paired samples vs. internal standard
- Automated spot matching
- differential analysis and statistical significance estimates
- Multiple gels analyzed vs. same internal standard
- large reduction in variation when comparing between samples

# Limitations of 2-DE

- No information on location within cell unless pre-fractionation of sample into cellular components
- Low abundance proteins difficult to see but may be overcome by pre-fractionation at cellular level, HPLC, differential solubilization
- No kinetic information
- Denaturing disassociates protein subunits, cofactors, substrates

# Proteomics using 2-D electrophoresis

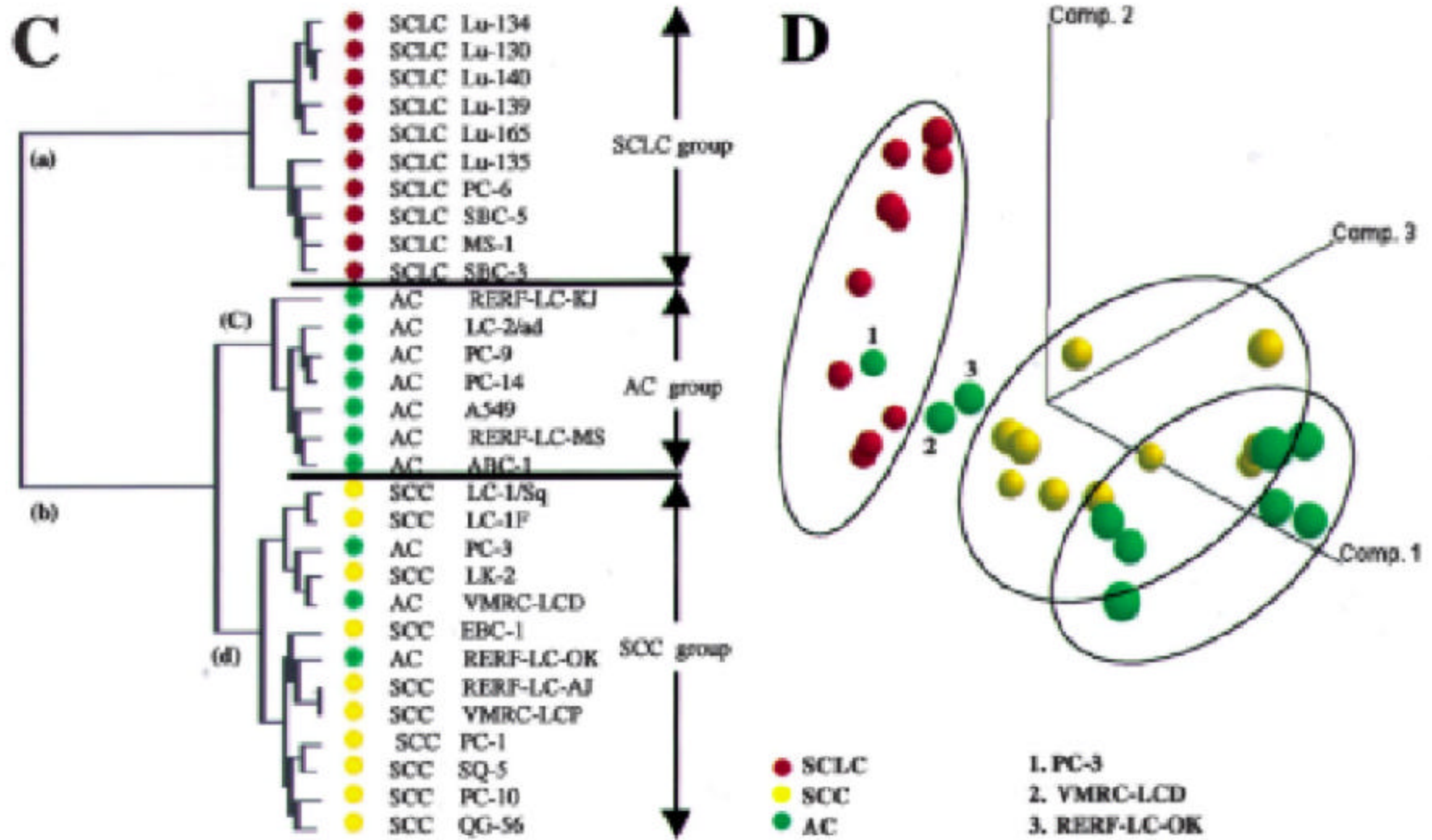
- protein expression patterns may be able to distinguish between cancer types
- determine origin of metastatic tumours
- example: classification of lung cancer subtypes using 2D-DIGE



Spots uniformly up-regulated (triangles) or down-regulated (circles) in lung small cell carcinoma. Seike et al 2004, *Proteomics* 4: 2776.

# Hierarchical cluster analysis of 71 protein spots on DIGE

SCLC, small cell lung carcinoma; SCC, squamous cell carcinoma; AC, adenocarcinoma.





**Table 2.** Protein spots with different intensity between SCLC and AC

Spot no. <sup>a)</sup>	Access no. <sup>b)</sup>	Protein description <sup>c)</sup>	Score <sup>d)</sup>	Number of peaks <sup>e)</sup>	Protein coverage (%) <sup>f)</sup>	Spot ranking <sup>g)</sup>	Fold differences <sup>h)</sup>
High in SCLC							
2848	P32119	Peroxisredoxin 2	815	10	44.4	4	4.12
1218	P40227	T-complex protein 1, zeta subunit	1374	18	38.2	8	1.28
3325	–	Not identified	–	–	–	12	6.12
1729	P20073	Annexin A7	323	8	18.2	14	1.36
547	P13639	Elongation factor-2	1265	19	23.7	15	1.44
995	P20700	Lamin B1	404	7	14.7	18	1.72
932	P11142	Heat shock protein 71 kDa protein	574	12	25.4	19	1.32
1361	P78371	T-complex protein 1, beta-subunit	2139	28	59.6	20	1.35
High in AC							
1681	P05783	Keratin 18	1783	22	46.5	1	0.04
1437	P05787	Keratin 8	1655	23	54.7	2	0.02
2105	P07355	Annexin II	997	14	47.8	3	0.13
1411	P05787	Keratin 8	1686	23	51.6	5	0.06
3358	P18282	Destrin	313	6	40.6	6	0.48
1435	–	Not identified	–	–	–	7	0.92
2405	–	Not identified	–	–	–	9	0.39
2225	–	Not identified	–	–	–	10	0.21
1338	–	Not identified	–	–	–	11	0.37
3322	P23528	Cofilin	334	4	31.9	13	0.77
2669	–	Not identified	–	–	–	16	0.51
1340	P50995	Glucose-6-phosphate 1-dehydrogenase	958	12	20	17	0.83



# **Differential expression of proteins in cellular senescence**

Preliminary work has been done to characterize proteins potentially involved in the human cell aging process

## Experimental setup

- ▶ human fibroblast cell lines
- ▶ senescence induced using *RAS* oncogene
- ▶ control and induced cells compared using DIGE
- ▶ differentially expressed proteins identified by mass spec

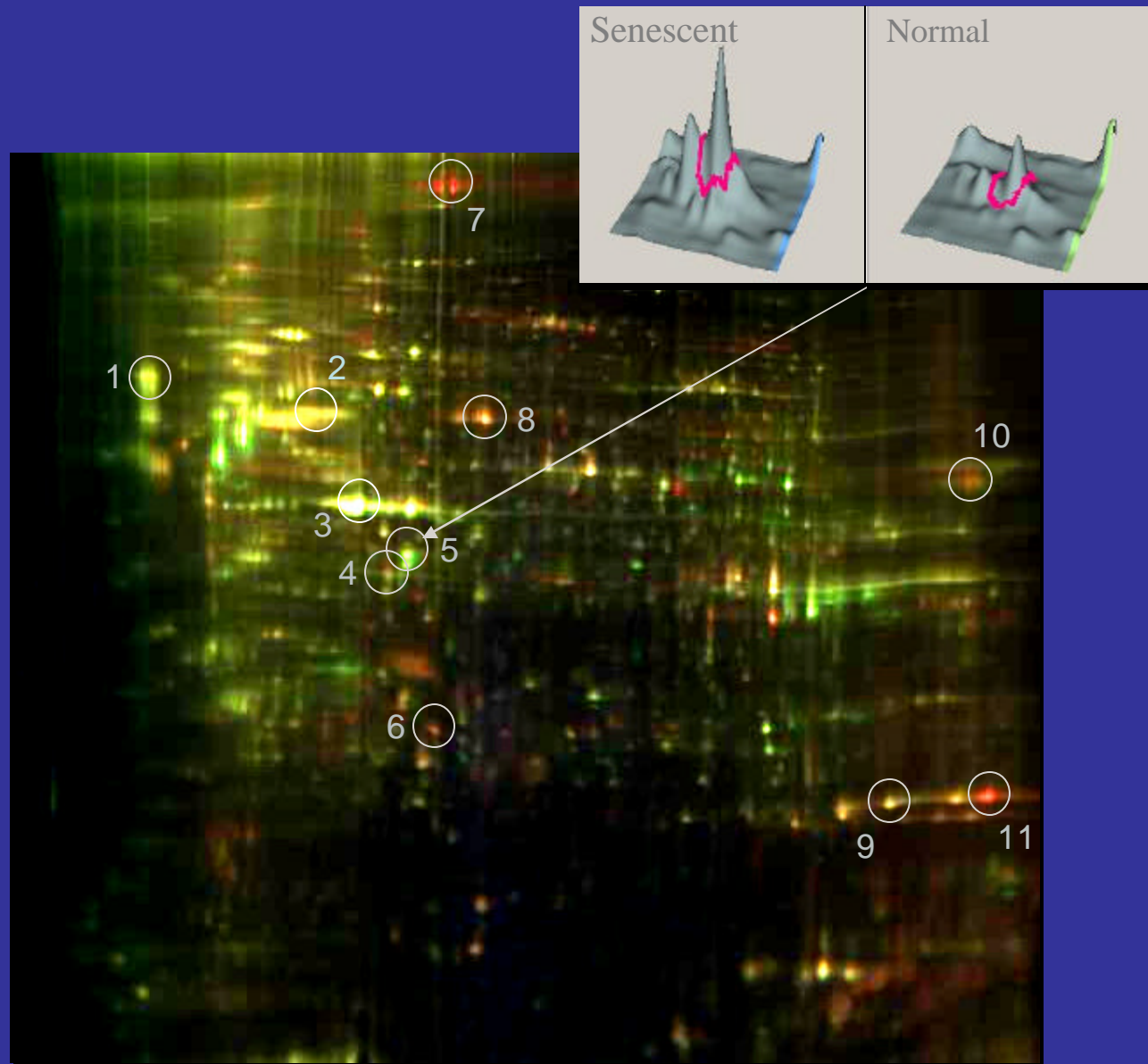


Image of a gel bearing both CyDye-labeled normal (red) and induced-senescent (green) fibroblast proteins. A volume map is shown for spot 5.

Fragmentation spectrum of a peptide obtained from a tryptic digest of disulfide isomerase ER-60. Interpretation of this spectrum yielded the sequence LE<sup>+</sup>LTDDNFESR.

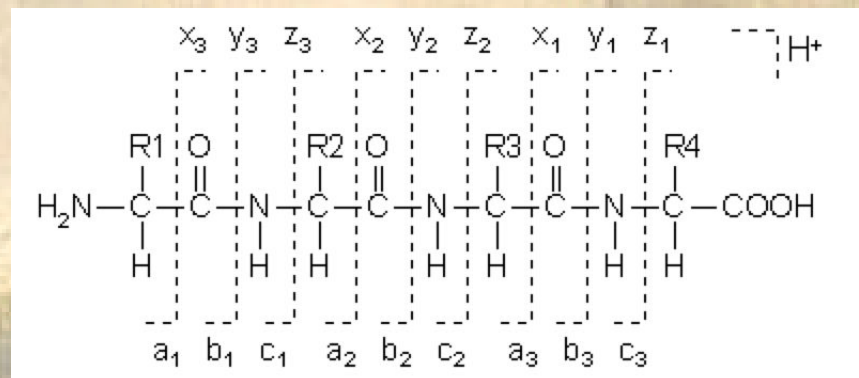
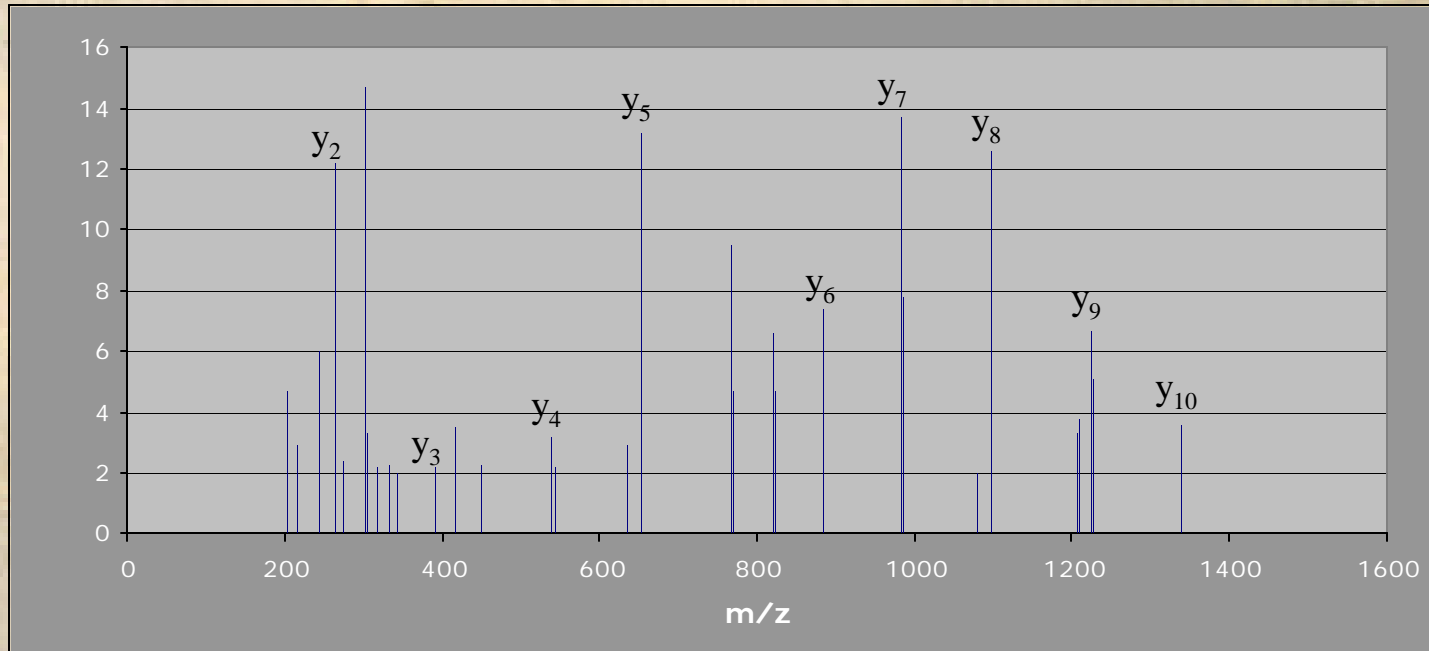
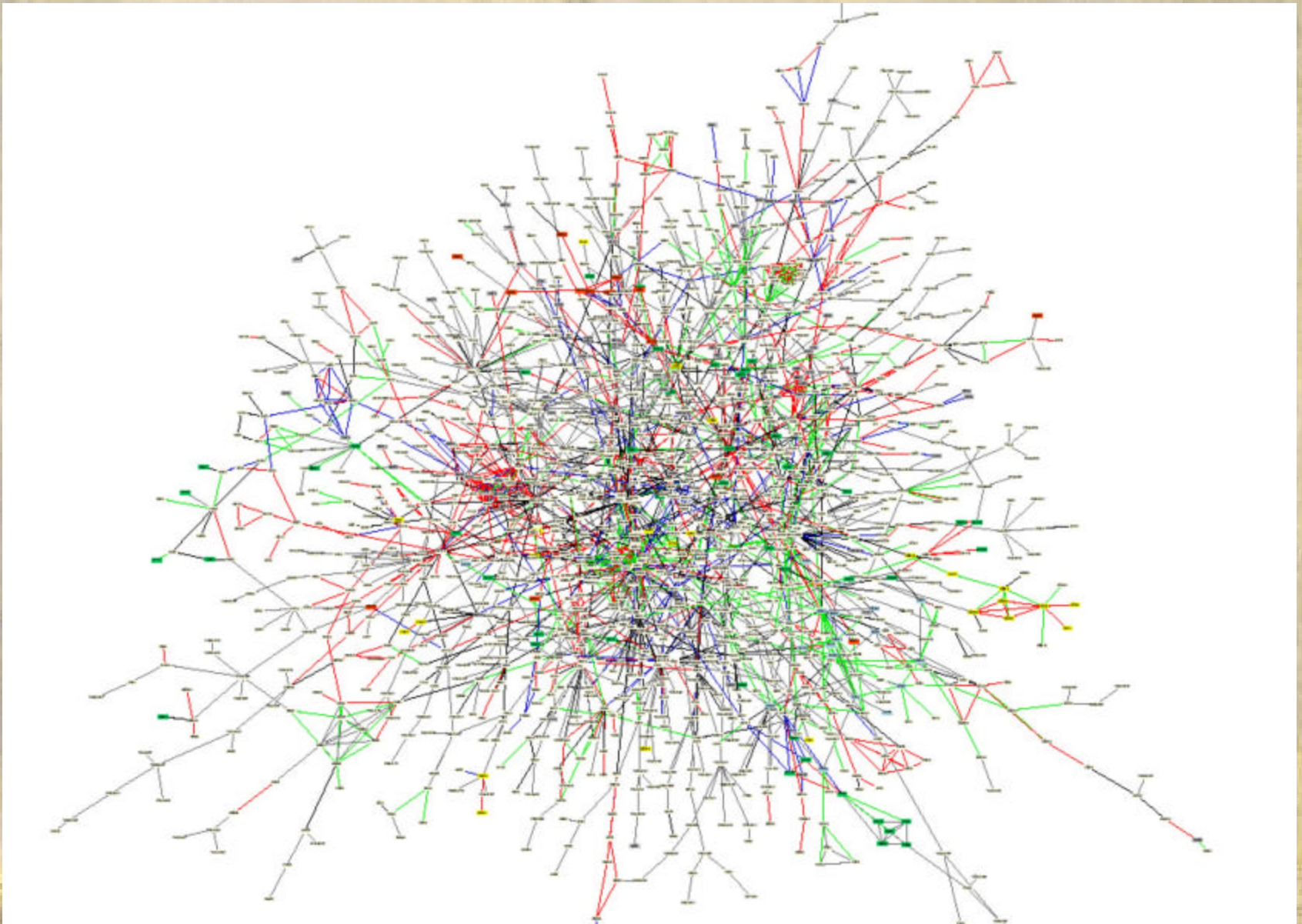


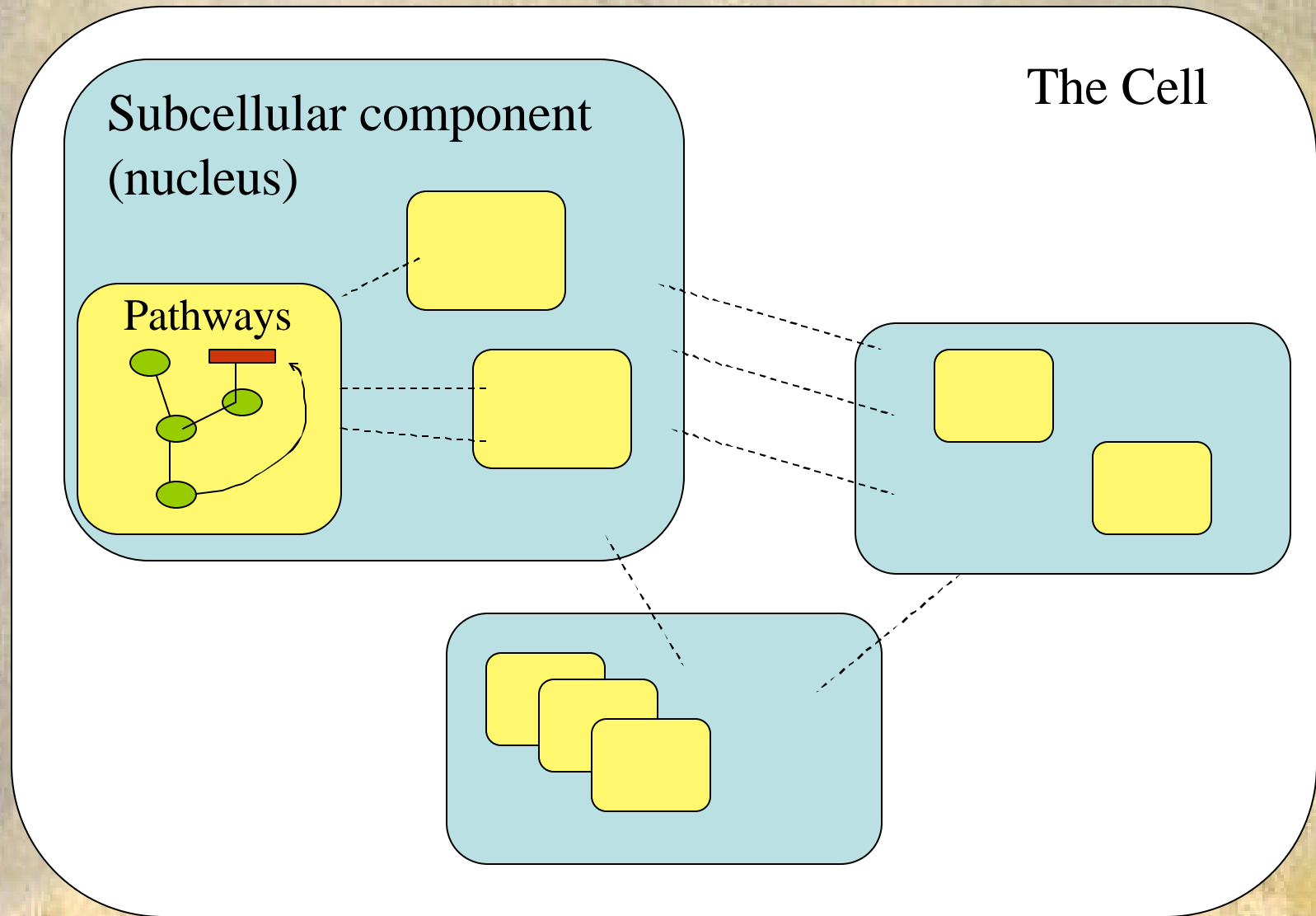
Figure Label	Accession Number	Protein	Average Ratio Senescent: Normal
1	P27797	Calreticulin (CRTC) - lectin, calcium binding chaperone	1.01
2	multiple	Vimentin + $\alpha$ -tubulin + tubulin $\alpha$ -chain 1 - filament protein	1.15
<b>3</b>	<b>P02571</b>	<b><i>g-actin (possibly with b-actin)</i></b> <b>- cytoskeleton</b>	<b>2.51</b>
4	P05218	tubulin $\beta$ -5 chain - microtubules	0.88
<b>5</b>	<b>Q8WU19</b>	<b><i>K-ALPHA-1 protein</i></b> <b>- microtubules, cytoskeleton</b>	<b>2.70</b>
6	P04792	Heat shock protein 27 kDa - stress response, actin organization	0.62
7	P02452	Collagen $\alpha$ 1(I) chain precursor - fibrillar forming, structural protein	<b>0.06</b>
8	P30101	Probable protein disulfide isomerase ER-60 - protein folding	<b>0.50</b>
9	Q06830	Peroxiredoxin precursor - redox regulation, signal cascades via H2O2(?)	0.62
10	P29043	Heat shock protein Hsp 47 precursor - collagen binding	<b>0.35</b>
11	Q01995	<i>Transgelin</i> <b>**contradicts RNA expression</b> <b>- linked to replicative senescence</b>	<b>0.15</b>



A messy way to look at interaction data:

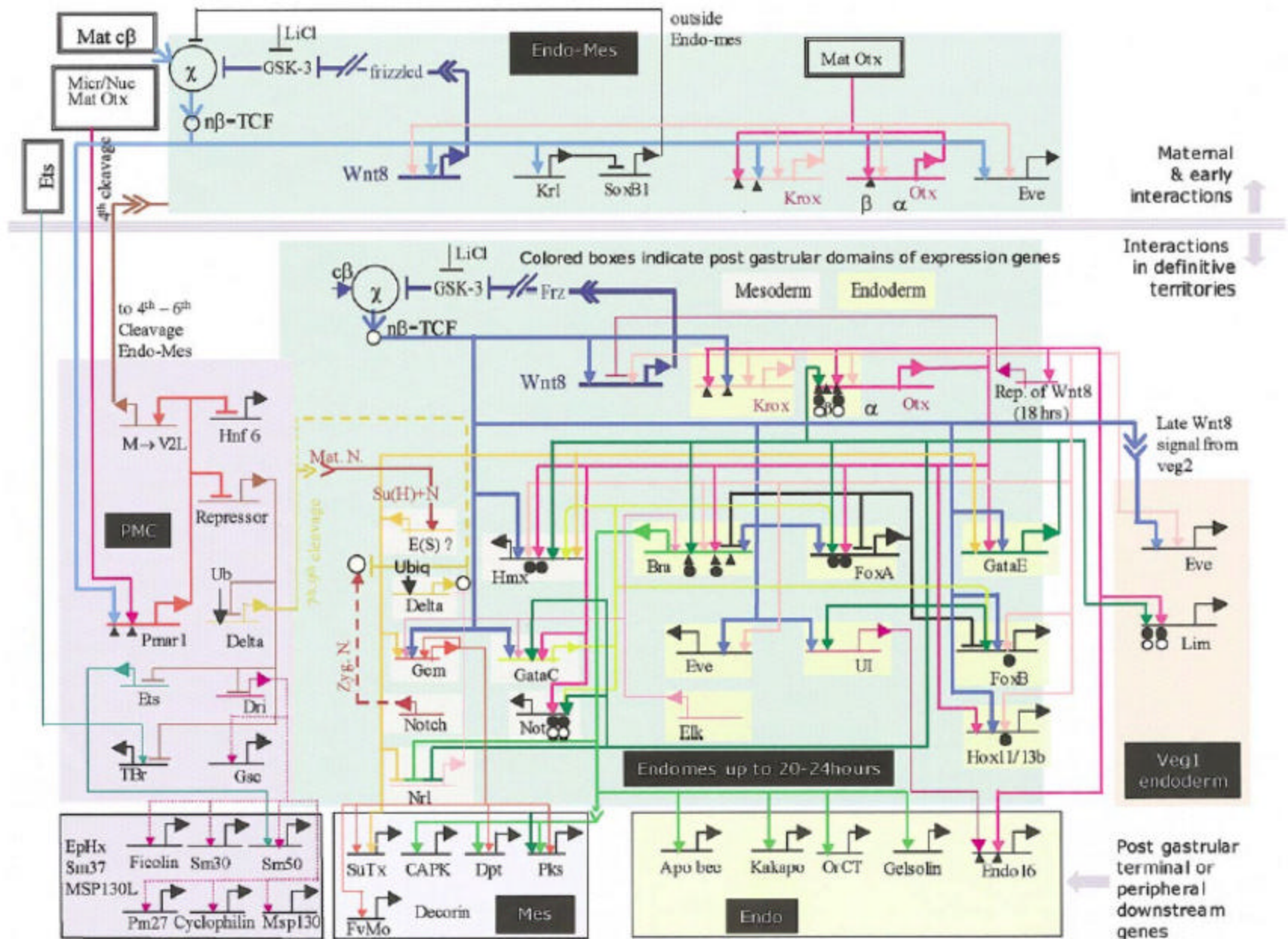


A better way - as a series of nested, interacting organizational units:





# Preliminary Regulatory Network in the Sea urchin for endomesodermal development



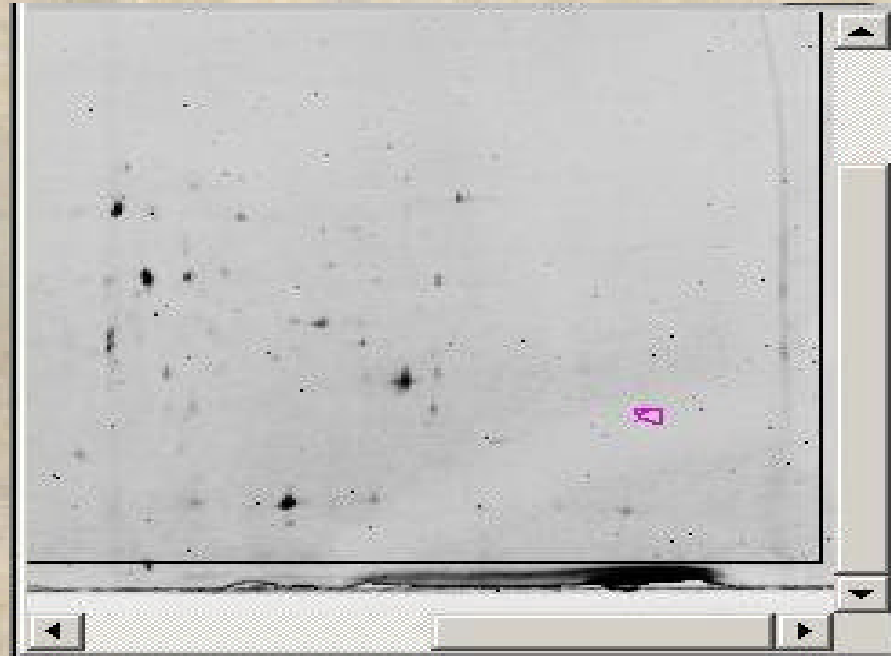
## Cell cycle modeling in yeast

- ▶ focus is the genes, proteins and interactions involved in the initiation of DNA replication (G1 --> S transition)
- ▶ DIGE used to isolate, quantify and identify differentially expressed proteins and protein isoforms
- ▶ protein levels and interactions modeled using differential equation model (Chen/Tyson/Novak)
- ▶ cell cycle model used to predict effect of perturbations on system
- ▶ perturbations to system created by genetic manipulation
- ▶ changes in protein level/type used to refine model

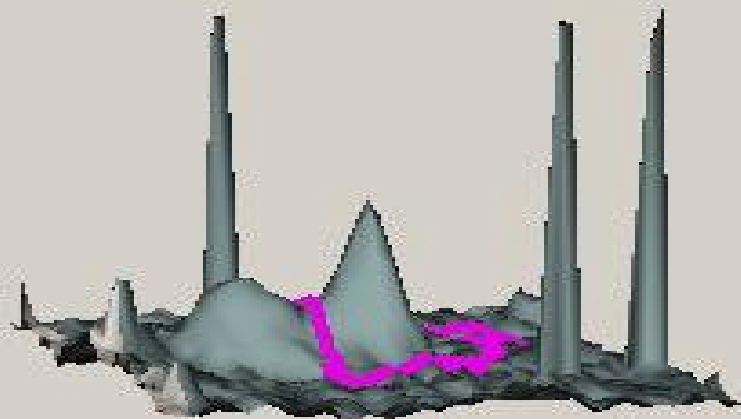




# Tentative identification of hct1 on DIGE gels - wt vs. Hct1 knockout



Reset



Spot No: 1833      Volume: 1.508e+004  
Position: 1326, 808      Peak Height: 244  
Pick pos.:      Area: 356

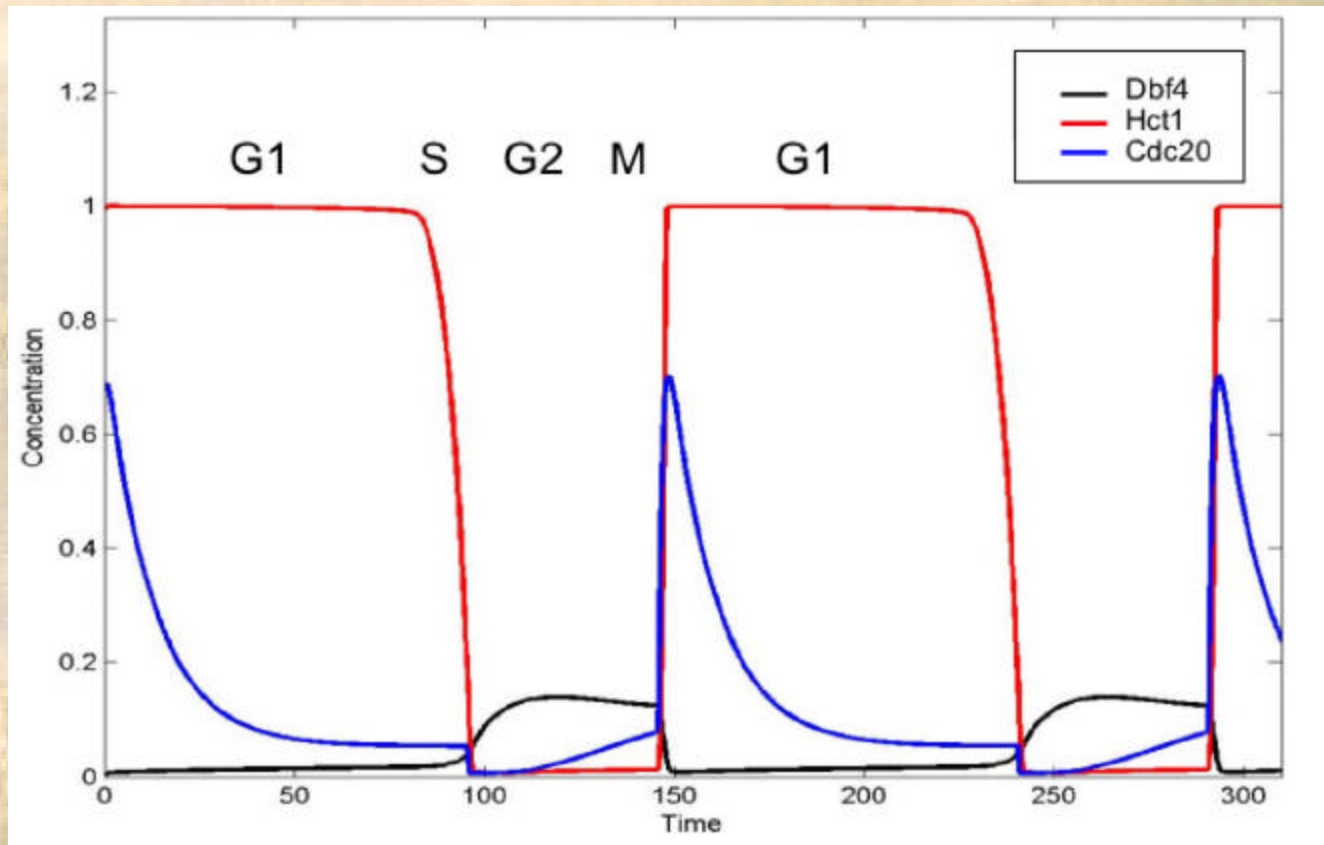
Reset



Spot No: 1833      Volume: 6114  
Position: 1326, 808      Peak Height: 30  
Pick pos.:      Area: 356

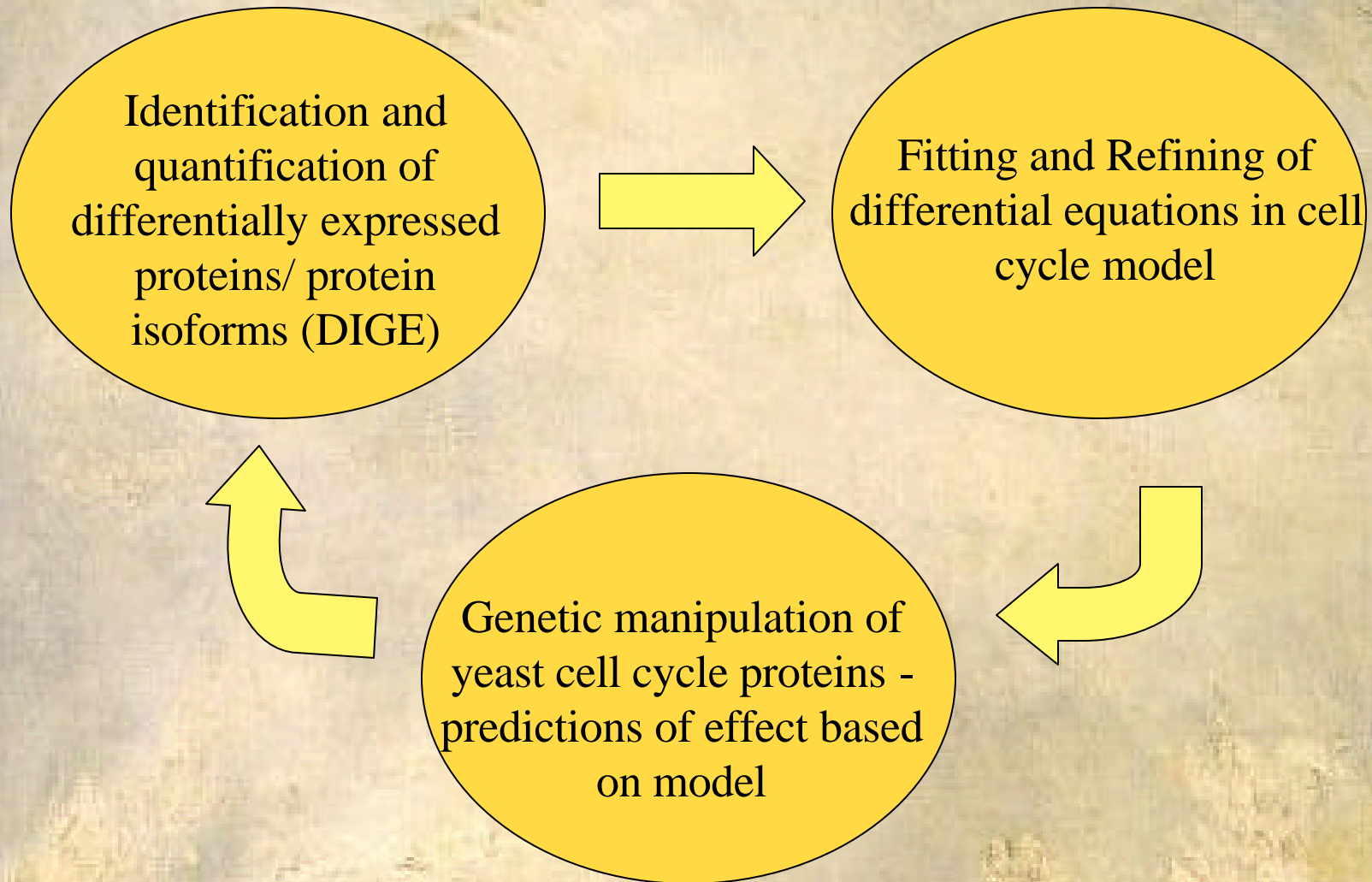


- ▶ model of three selected cell cycle proteins: Dbf4, Cdc20, and Hct1
- ▶ Hct1 and Cdc20 are from Chen/Tyson/Novak model
- ▶ Dbf4 was modeled as dependent on Hct1



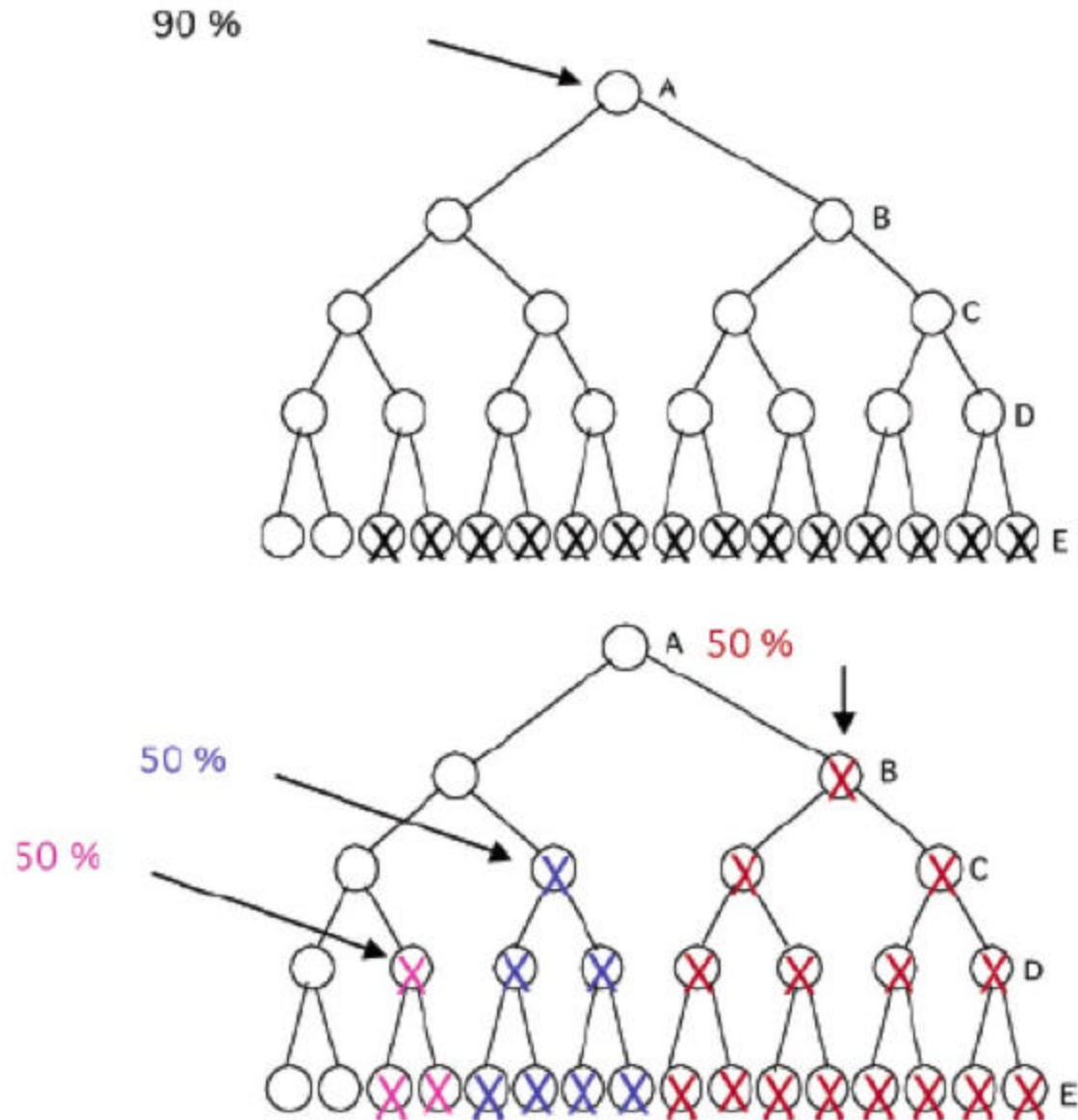


# REFINEMENT OF CELL CYCLE MODEL



Eventual goal - an accurate description of the gene/protein network involved in the initiation of DNA replication

# Combinatorial Therapeutics: reduced toxicity, increased efficacy



## Conclusions

- bioinformatics technologies (microarrays, 2D-DIGE, mass spectrometry, ...) have wide potential application in diagnostics and treatment
- still in a relatively early phase of development- not practical for medical applications yet
- potential for identifying novel drug targets or multiple interacting targets
- remains to be seen whether utility of these methods outweigh instrumentation costs and required expertise

# Acknowledgements

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