

# Bioimaging and Functional Genomics

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# Outline

- Context and aim
- Linking genes with clinical traits
- Joint co-clustering
  - Co-clustering
  - Automated clinical image processing
- Conclusion

# Context and Aim

- Study of multi-factorial genetic pathologies
- Improve clinical diagnosis and develop specific new therapy
  - Understanding of genetic mechanisms underlying clinical observations
  - Increase the amount of confidence in the hypothesized gene expression paths

# Approach

- Linking genes with clinical traits

- How?

- Correlation of **gene** expression with **clinical trait** measurement
- Clinical traits can be obtained through several techniques

- Example:

- Biopsy genetic analysis and tissue imaging data

# Solution

Gene Expression  
Analysis

BIO Medical  
Imaging

Enhanced  
Genetic  
Analysis

More information  
about genetic  
mechanisms

More confidence  
in gene  
expression paths



# Gene Expression Analysis

## ● Objective

- Correlation of gene expression with multifactorial genetic pathologies

## ● Techniques exploited so far

- Real-time PCR (qPCR)
- Microarray data analysis

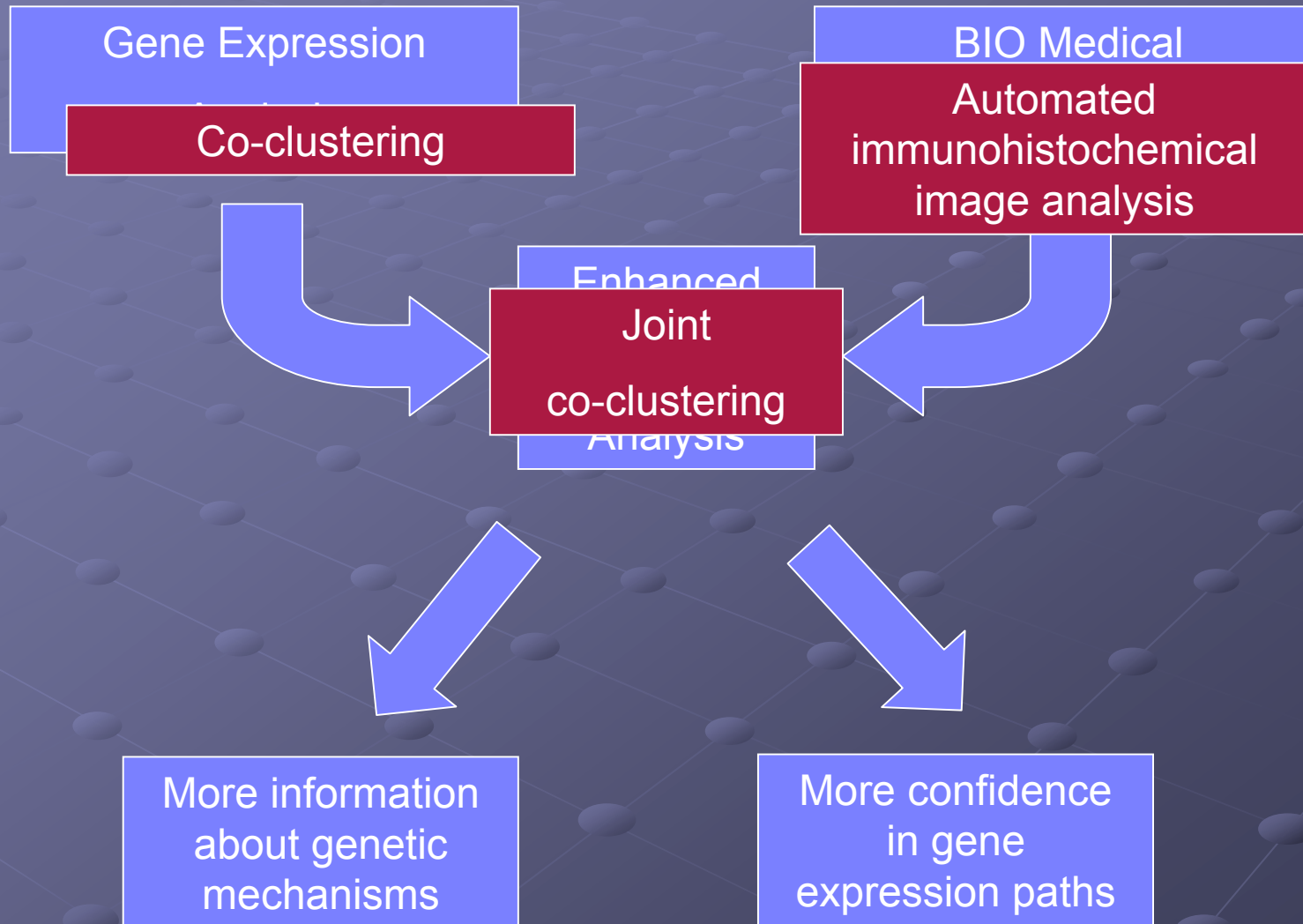
## ● Drawback

- Information provided by genetic clustering techniques is not sufficient to draw pathological conclusions

# Bio and Molecular Imaging

- Extraction of clinical and functional biological information from images of molecules and tissues
- Important roles in clinics
  - E.g, study protein activations in carcinoma tissues
- Objectives
  - Early disease detection
  - Drug response
  - New Therapies

# Our Solution





# *Joint co-clustering*

## ● Correlation between

- Gene expression analysis
- Bioimaging

## ● *Joint co-clustering:*

- *Fully-automated computer-aided image analysis approach* to extract clinical bioimaging parameters
- *Co-clustering* applied to gene expression data and clinical bioimaging parameters

# Outline

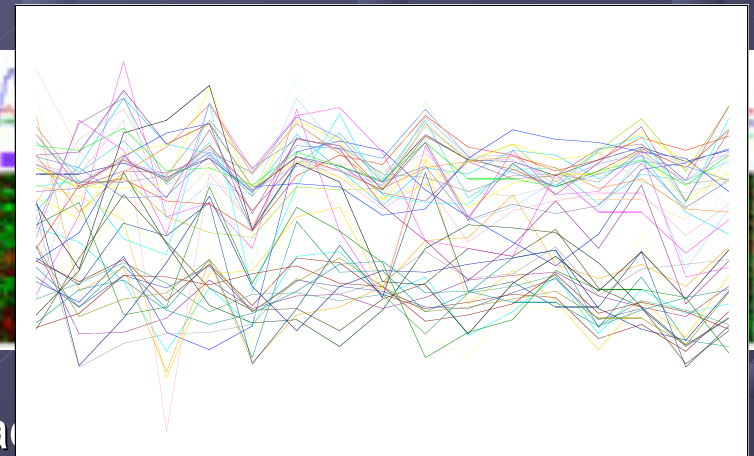
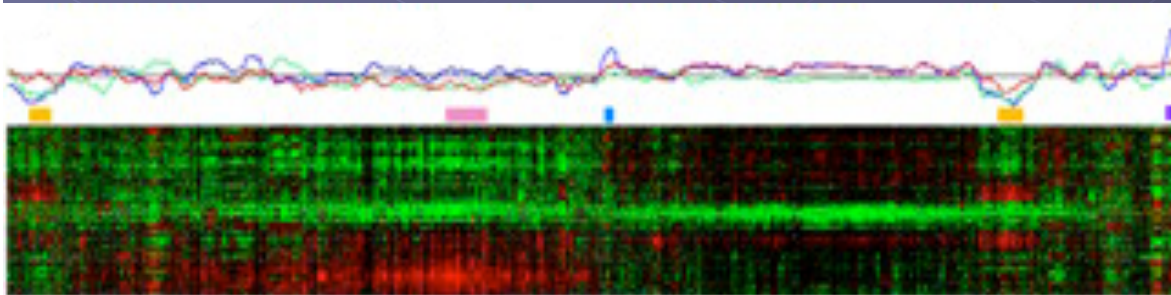
- Context and aim
- Linking genes with clinical traits
- **Joint co-clustering**
  - Co-clustering
  - Automated clinical image processing
- Conclusion

# Co-clustering

- Unsupervised cluster search method in high dimensional spaces
- Linking genes and clinical traits
  - Provides qualitative information about pathology

# Challenges

- As the number of clinical traits increases, the inspection method breaks down
  - We need a systematic approach
- The values of clinical traits are not necessarily continuous or numeric
  - We need a more general statistics than the Pearson correlation coefficient

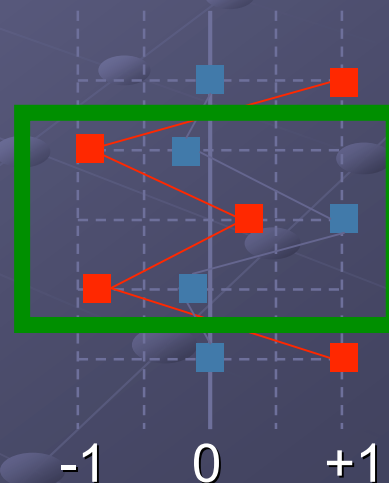


# Definition

- *Co-cluster* of genes and clinical traits
  - A submatrix of the correlation matrix
  - Elements are statistically significant
  - For any pair of column vectors, the inter-column distance is less than a threshold

	Trait 2	Trait 1
⋮	0.0	1.0
	-0.2	-0.9
	0.7	0.2
	-0.2	-0.9
	0.0	1.0

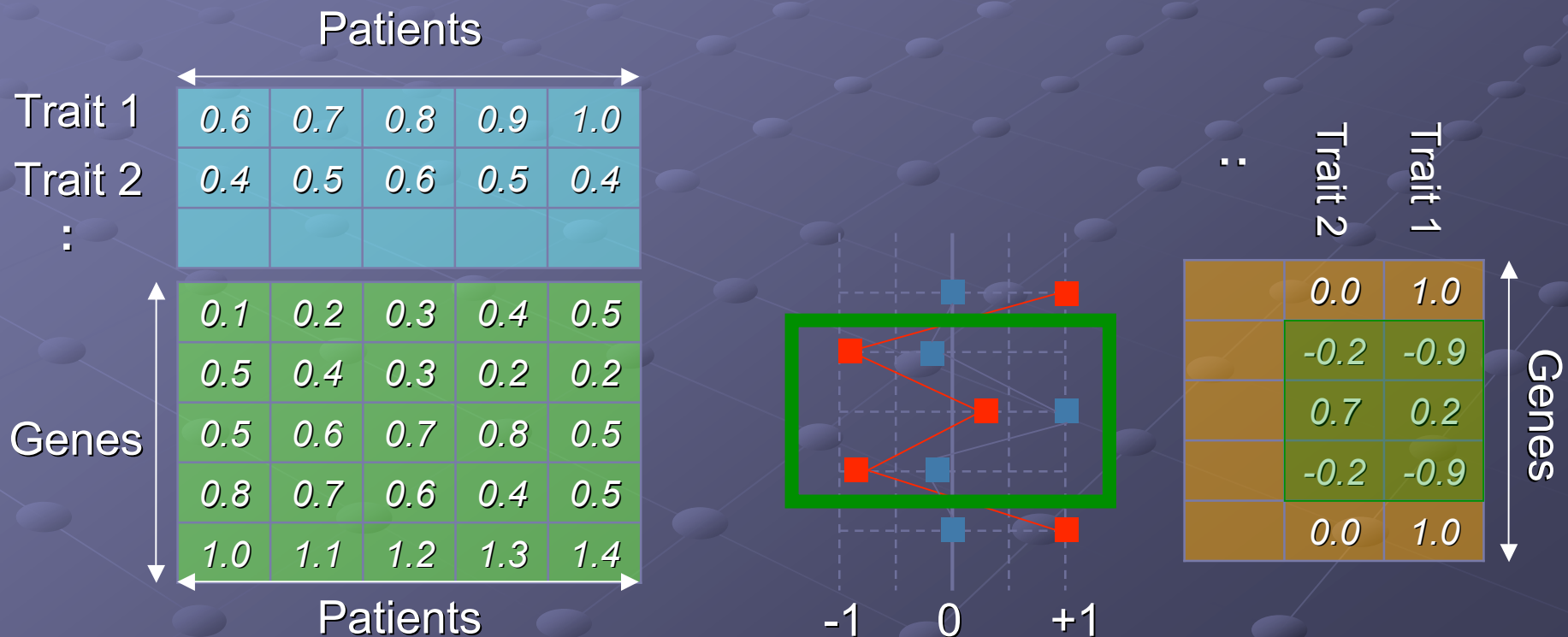
↑ Genes



Correlation matrix

# Our Approach

- Construct a “correlation matrix”
  - Use the SAM statistic (Tusher *et al.*, “Significance analysis of microarrays applied to the ionizing radiation response”, 2001)
- Find co-clusters of genes and traits appearing on the correlation matrix

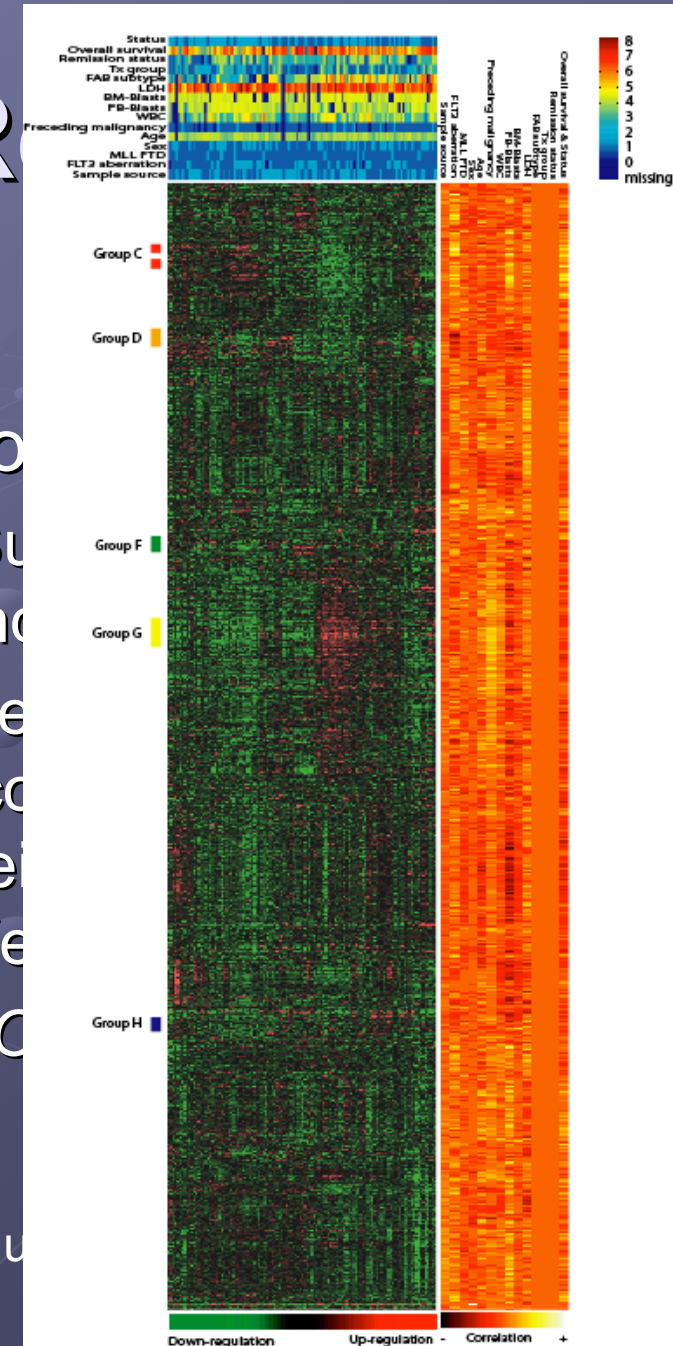
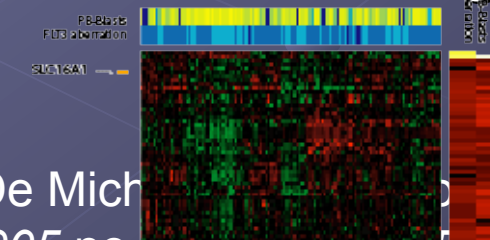
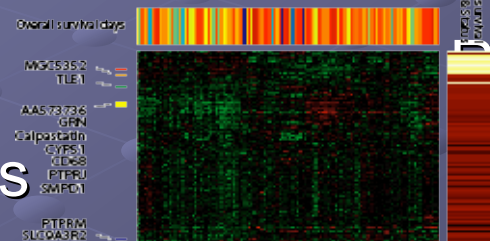
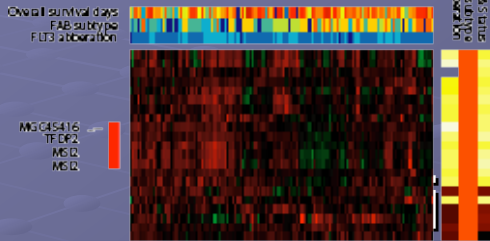


# Experimental R

## ● AML data set

- Bullinger *et al.*, 2004
- 6283 genes
- 119 patients
- 15 clinical traits

## ● 43 GT co-clusters



S. Yoon, L. Benini, and G. De Micheli, "Gene Clustering and Network Analysis of AML Data Set", *IEEE EMBC 2005* no. 11.3.1.4, 2005

# Outstanding Issue

- Extraction clinical parameter from image
- Can this be automated?



# Immunohistochemical (IHC) automated quantification

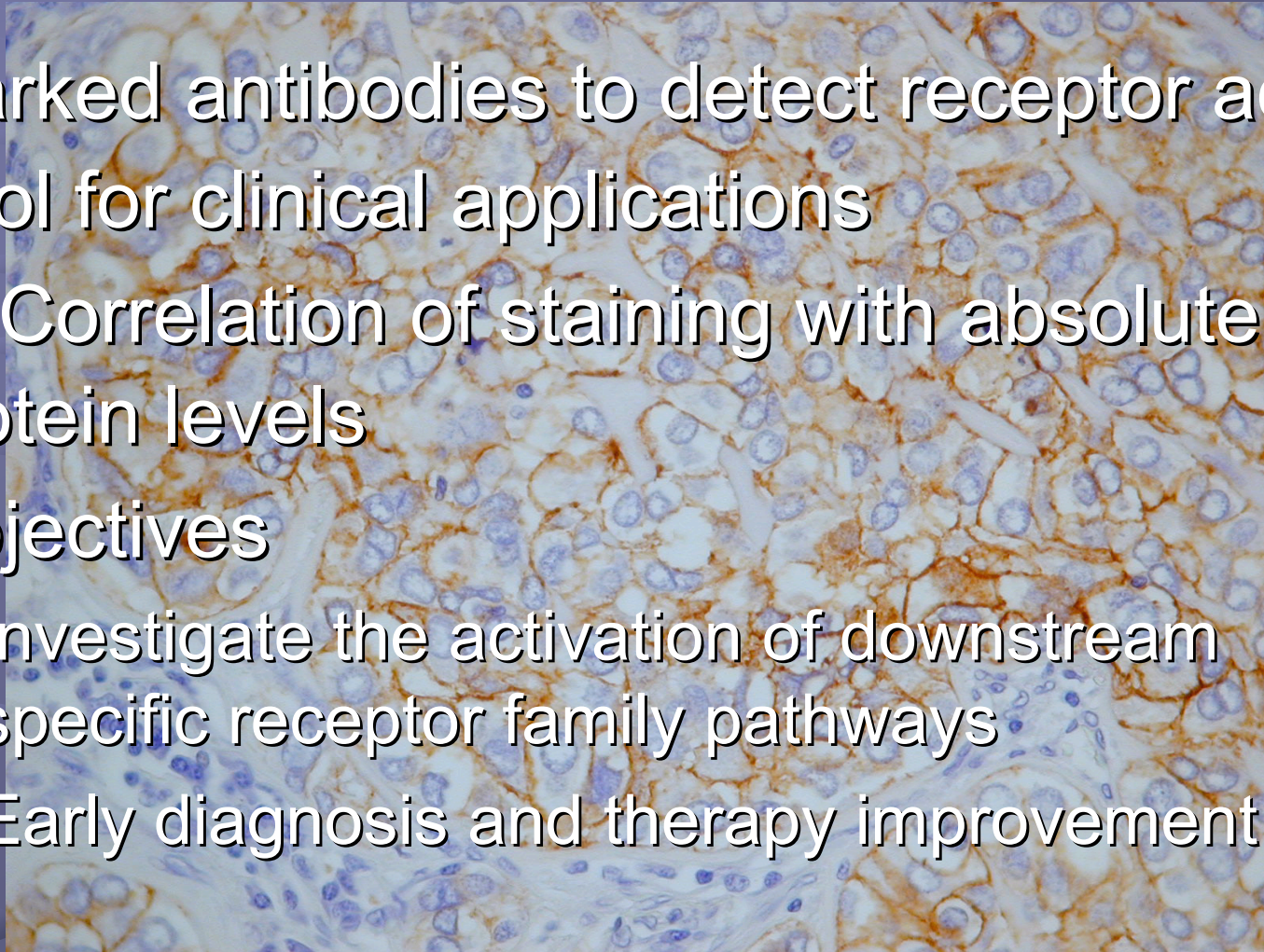
- Development of bioimaging techniques for acquiring quantitative and qualitative information from immunostains



- Development of an automated image processing method to standardize IHC analysis

# Immunohistochemistry (IHC): characteristics and aim

- Marked antibodies to detect receptor activity
- Tool for clinical applications
  - Correlation of staining with absolute protein levels
- Objectives
  - Investigate the activation of downstream specific receptor family pathways
  - Early diagnosis and therapy improvement



# Non small cell lung carcinoma (NSCLC) Project

- The EGFR/erb-B family of receptors
  - Important role for non small cell lung carcinoma (NSCLC) development
- Project objectives
  - Evaluation of the correlation between EGFR genetic alterations
  - Evaluation of the activation of downstream pathways
  - Definition of a subgroup of NSCLC able to respond to EGFR inhibiting therapy

# IHC quantification framework: parameters

- Localization of marker (i.e membrane, cytoplasm, nucleus)
- Reaction intensity
- Percentage of (EGFR) positivity w.r.t. negative protein reaction in the carcinoma cells **computed at the same cellular area**

# Cell segmentation

## ● Challenging issue

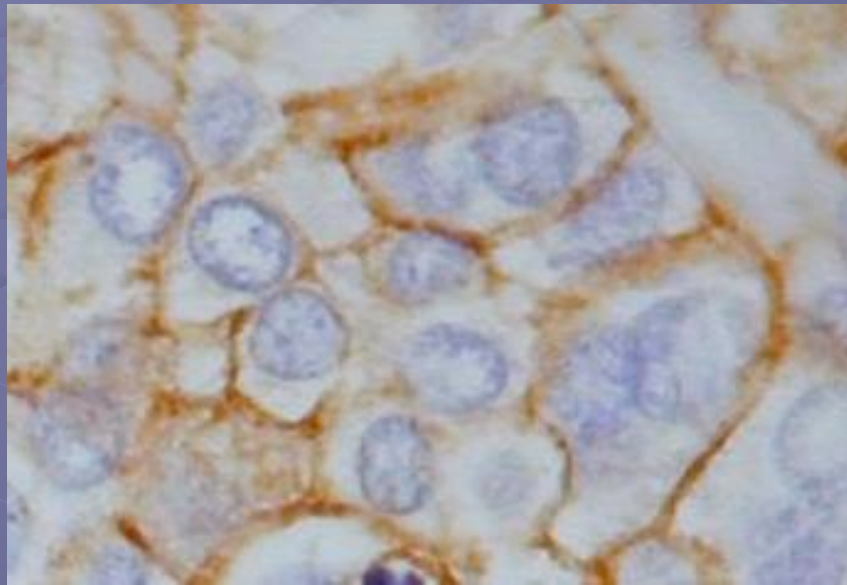
- Morphology variability
  - ⇒ difficult to design geometrical models
- Noise in the images
- Cancer cell membranes with negative protein reaction are not visible

## ● Standard methods based on geometrical and/or gradient variation analysis are not effective

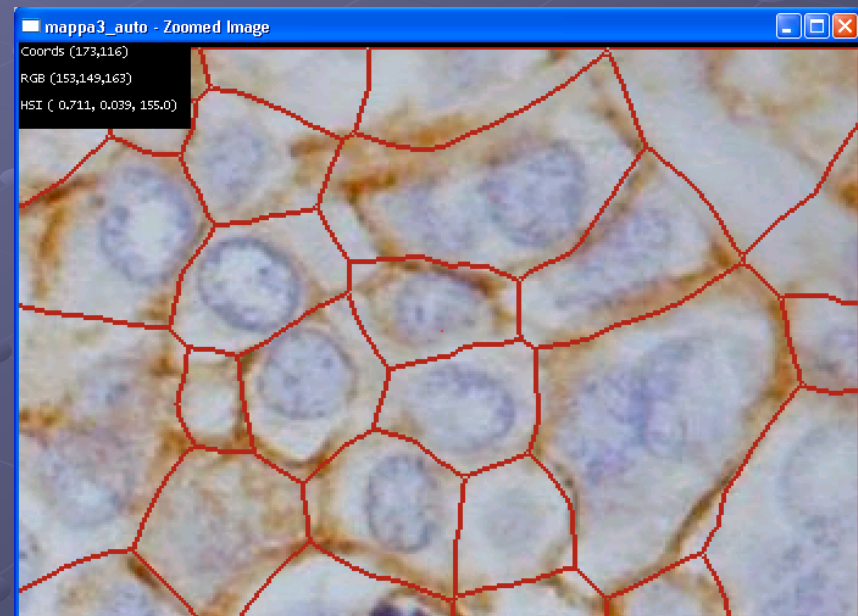
⇒ **Customized automated procedure for cellular segmentation in IHC tissue images**

# Bioimaging tool<sup>(1)</sup>

Original image (a detail)



Virtual cell membrane segmentation



Several steps:

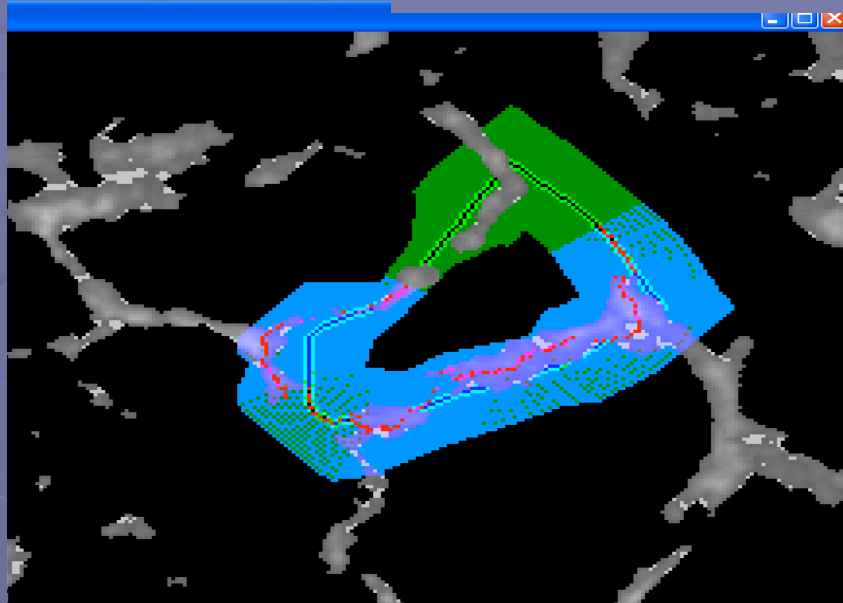
- Color filtering
- Nuclear membrane segmentation
- **Virtual cell membrane definition**



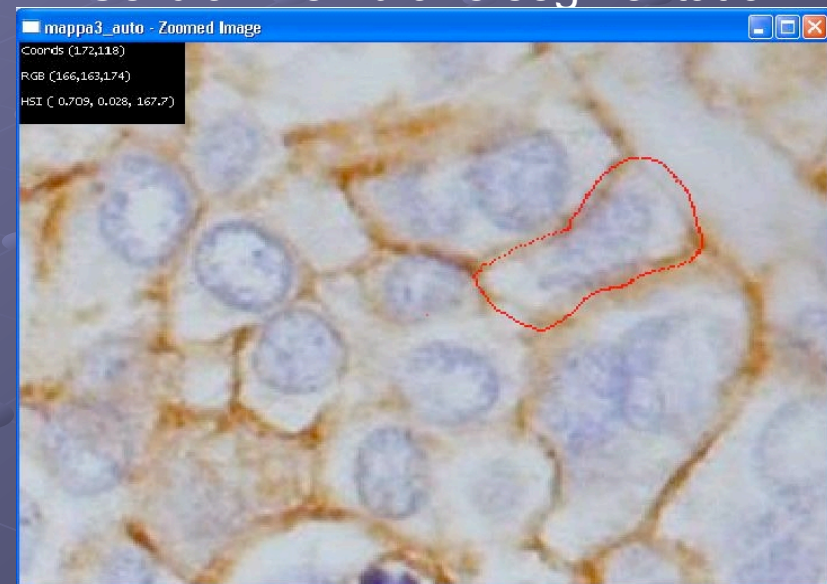
*Original automated  
customized procedure*

- Detection positive reaction (brown) areas when they exist
- Connection positive cellular membranes with cell virtual ones where positive reaction doesn't exist

## Scanning Process



## Cellular membrane segmentation



Several steps:

4. Cellular membrane segmentation:
  - Scanning procedure
  - Fitting membrane points
5. Clinical parameters computation



*Original automated  
customized procedures*

# Clinical Parameter Computation

## ● Protein activity computation

- Percentage of brown pixels in the cellular membranes w.r.t. total number of membrane pixels
- Validation: comparison with manual measurements

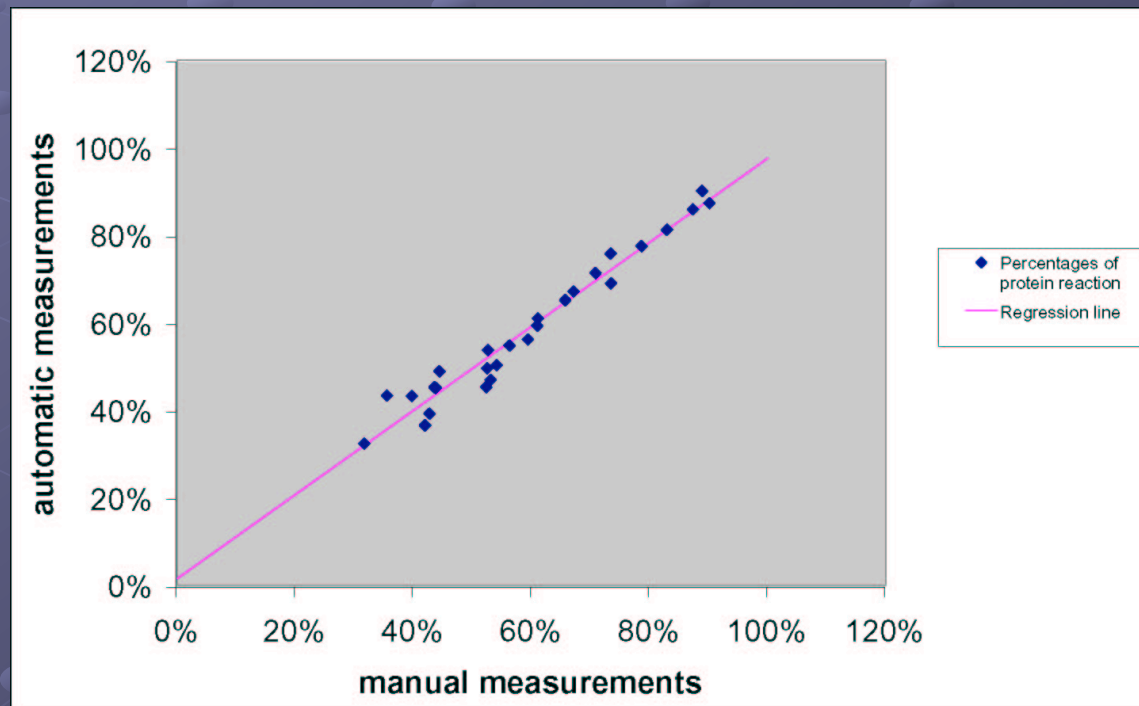
### Results:

Coeff. Correl 0.98

Perc Reac = 59%

Aver Err = -0.77%

RMSE = 3.3%





# Conclusion

- Bioimaging as standardized IHC image analysis
  - Extraction of quantitative and qualitative parameters for activation of downstream pathways analysis of receptor families
- Correlation of these parameters with data coming from gene expression analysis → **joint co-clustering tool**
- Preliminary results (85% correlation) show *joint co-clustering* is promising approach:
  - To analyze largescale biological data
  - To study multi-factorial genetic pathologies through their genetic alterations
  - To enable new opportunities for early diagnosis
  - To provide information in future strategies for therapy