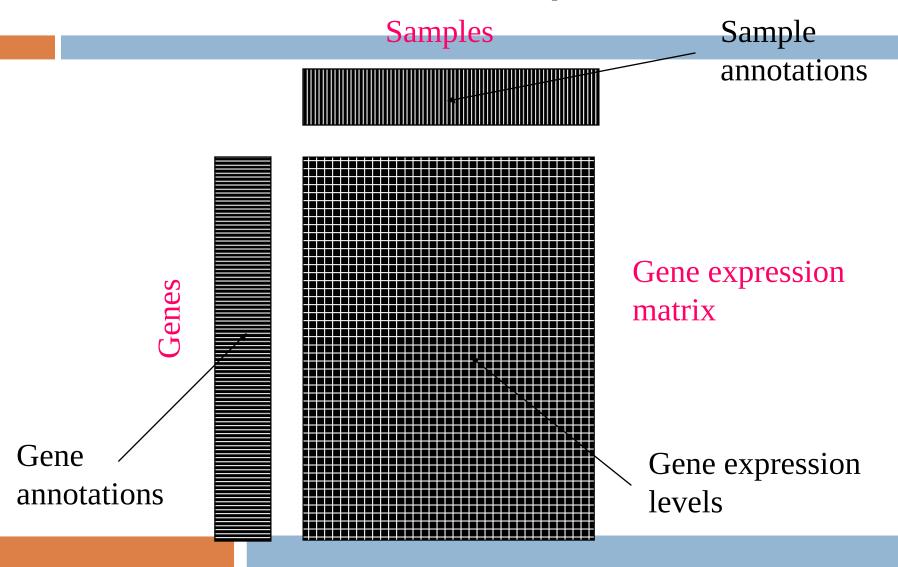
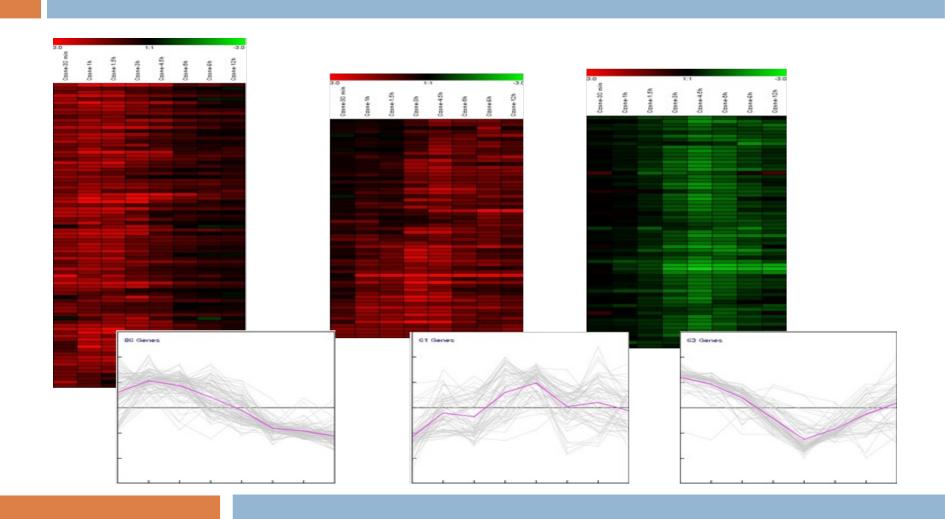
Bioinformatika **Gene expression data analysis**

Michael Anděl

GE data - conceptual view



GE data – image view



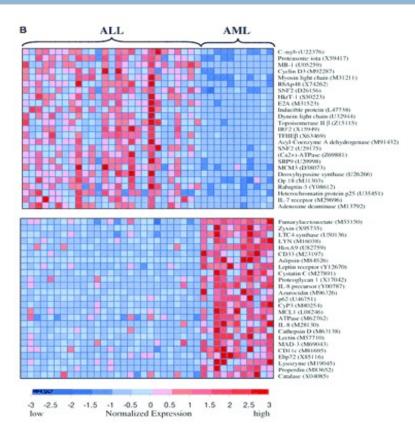
Classification task

Challenge:

- \circ samples (10^1) x features (10^3)
- False hypotheses, overfitting
- Interpretability: are the expressed genes the causal ones?

What to do?

- Decrease number of hypotheses
- Analyze more abstract entities than genes, eg. principal components



Golub et al.: Molecular classification of cancer: Class discovery and class prediction by gene expression monitoring. Science, 1999

PCA - motivation

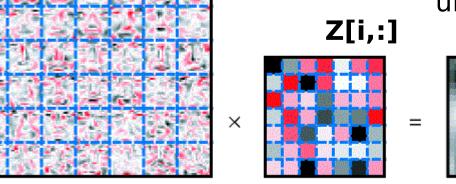
M ... # genes

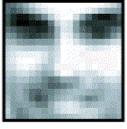
PCA

- N ... # samples
- X (N x M) ... GE data in the space of genes

V[:,:]

- V (M x K) ... transformation basis, eigengenes
- Q (N x K) ... transformed GE data in the space of eigengenes
- K ... # of eigengenes,
 i.e. the number of
 underlying concepts
 X[i,:]





Lee et al.: Learning the parts of objects by non-negative matrix factorization. Science, 1999

Data:

- 7,129 GE profiles of 72 patients
- 25 samples: acute myeloid leucaemia (AML)
- 47 samples: acute lymphoblastic leucaemia (ALL)

Golub, T., et al. *Molecular classification of cancer: class discovery and class prediction by gene expression monitoring*. Science 286.5439 (1999): 531-537.

Task:

 Construct decision model to differentiate these types of tumours. Just complete the code in attached script ge_cv.m

Workflow I:

- 1. Learn a decision tree on subjected data. Use Matlab class ClassificationTree and its method fit.
- 2. Show the tree (method view) and enumerate its training accuracy. How would you interpret this model? Which gene is crucial for decision? Compare it with the original article. Have this gene been previously reported as Leucaemia associated? Use e.g. phenopedia.
- 3. Estimate its *real* accuracy by the means of crossvalidation. Compare the two accuracies. Why are they so different?

Workflow II:

- 1. Learn a basis matrix *V* of the data. Use attached function pca.m.
- 2. For $K = \{5, 10, 20, 50\}$, project the original data X to the top K components V. The result is reduced data $Z = XV_{1:K}^T$.
- 3. On these 4 transformations, learn decision trees, show them and enumerate their *training* accuracy.
- 4. Choose optimal decomposition and corresponding tree-model (use training accuracy and Occam razor). Estimate *real* accuracy of the optimal model by crossvalidation.

Understanding the model:

- 1. Use attached function mineGenes to extract the genes, frequietly appearing in the components presented in your tree-model
- 2. Resulting gene-sets, related to each component in the tree, interpret in overrepresented GO terms. Use e.g. amigo.
- 3. Create a story!