



Data Mining for Biological Data Analysis

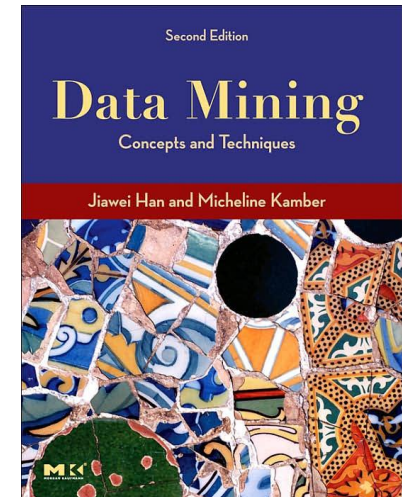
Data Mining and Text Mining (UIC 583 @ Politecnico di Milano)

References

- ❑ Data Mining Course by *Gregory-Platesky Shapiro* available at www.kdnuggets.com

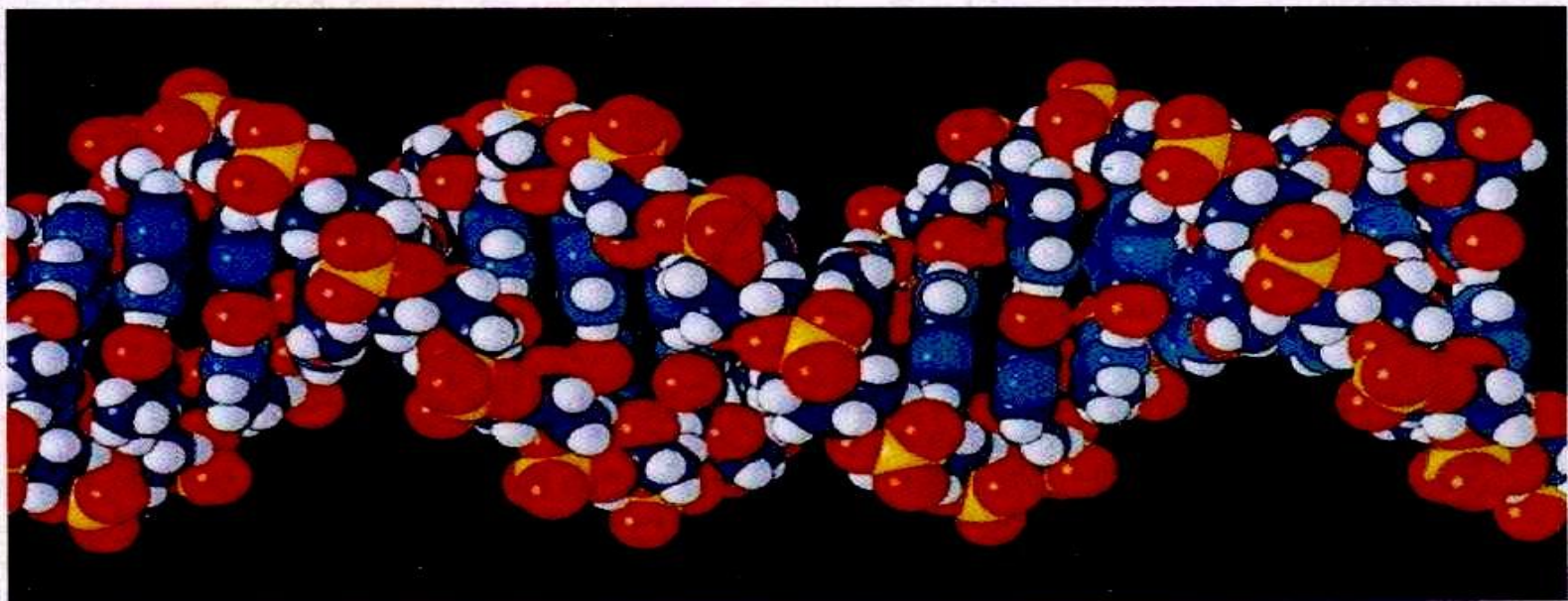
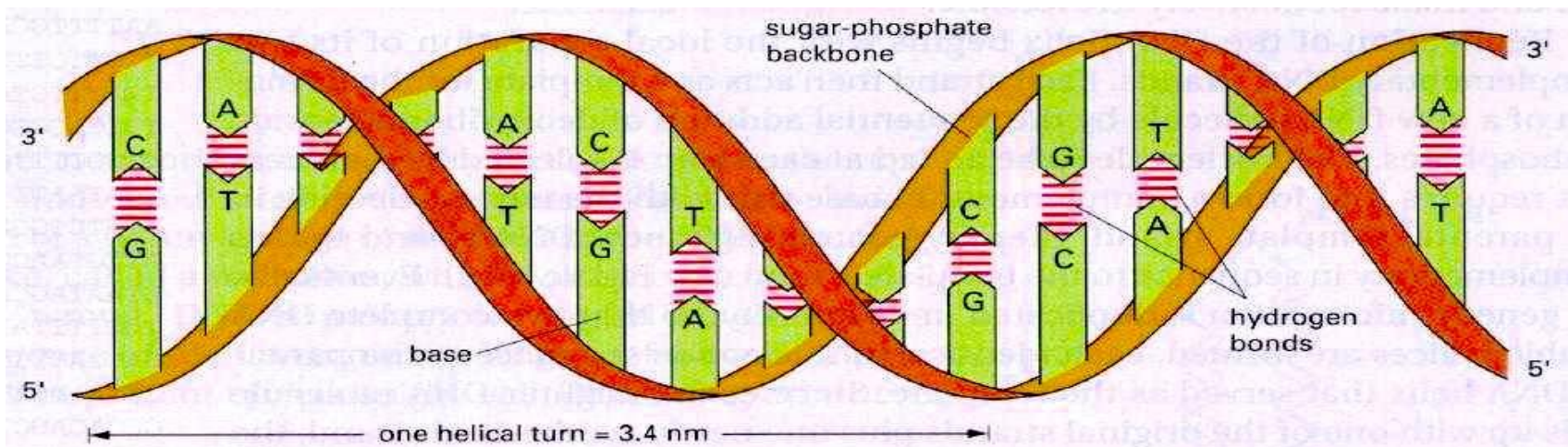


- ❑ Jiawei Han and Micheline Kamber, "Data Mining: Concepts and Techniques", The Morgan Kaufmann Series in Data Management Systems (Second Edition)
 - ▶ Chapter 8



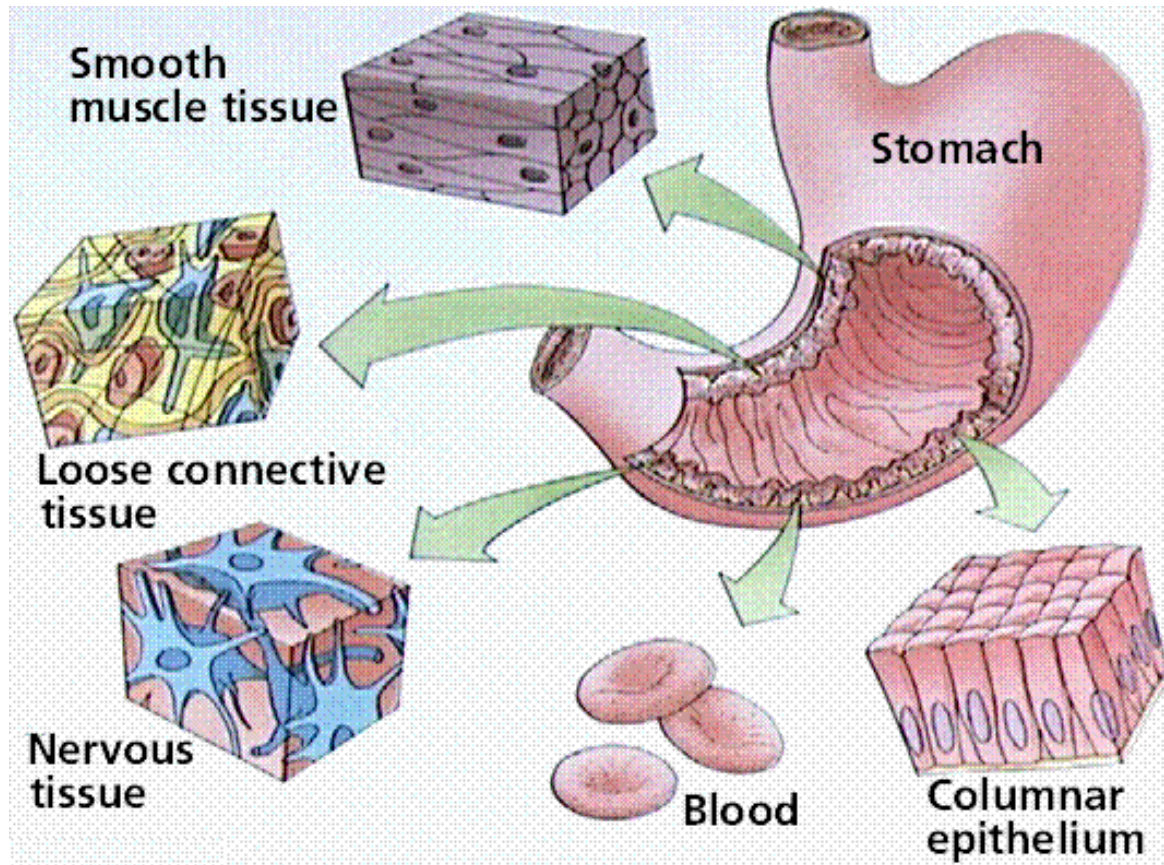
Introduction to Biology

The DNA



Different cell types

- All cells of an organism share the same DNA content (and the same genes) yet there is a variety of cell types.

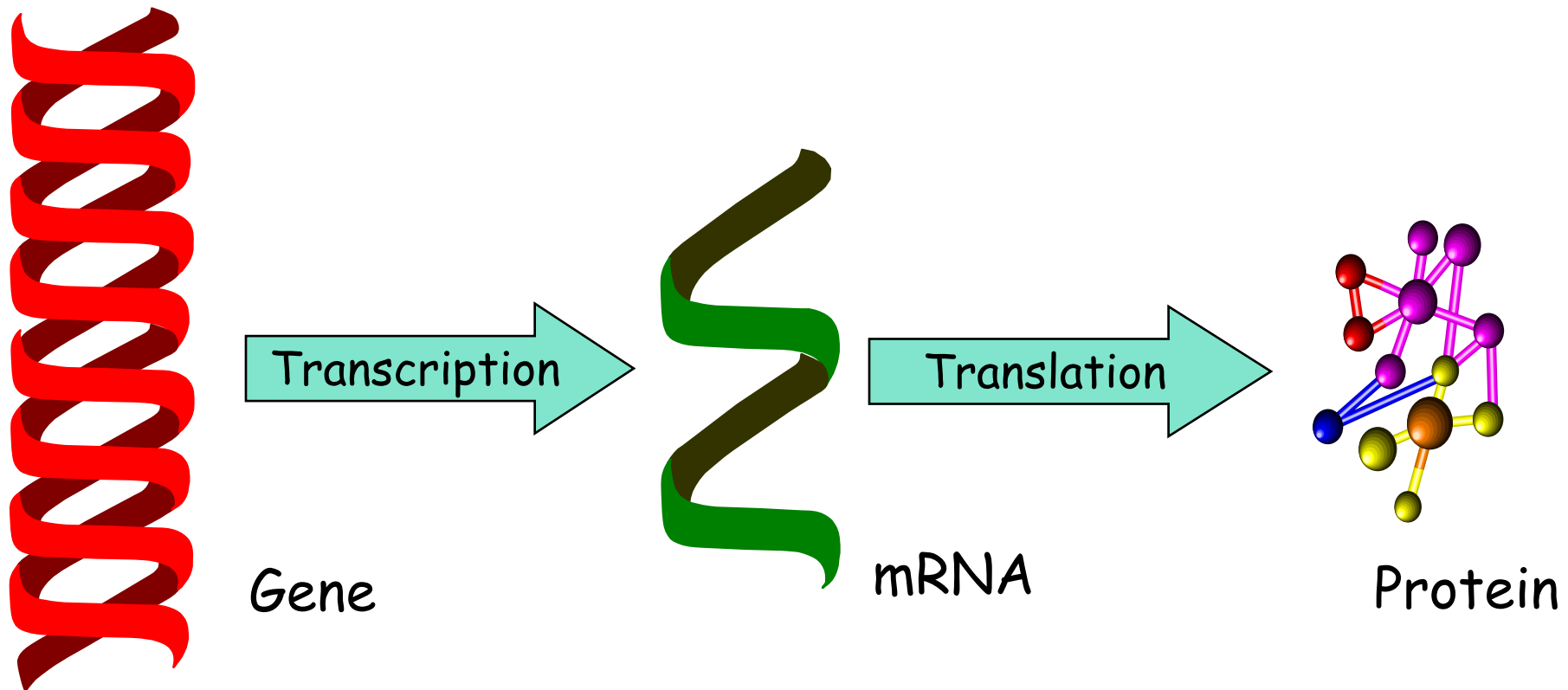


So, how does the cell
use DNA ?

The “Central Dogma”

- ❑ DNA contains thousands of particular segments called **genes**
- ❑ Genes contain “instructions” for making **proteins**
- ❑ In order to be executed these “instructions” have to be transcribed into **mRNA**
- ❑ Proteins are defined by a sequence of **amino acids** (20 types)
- ❑ There are almost one million of proteins that act alone or in complexes to perform many cellular functions

Gene expression



Cells express **different** subset of the genes in different **tissues** and under different **conditions**

Muscle, nervous, blood ...

Disease, mutation...

- ❑ **Thousands** of genes ($\sim 25K$ in human DNA) function in a **complicated** and **orchestrated** way that creates the mystery of life.
- ❑ **Genomic** studies the **functionality** of specific genes, their relations to **diseases**, their **associated proteins** and their participation in **biological processes**
- ❑ **Proteins** ($\sim 1M$ in human organism) are responsible for many regulatory functions in cells, tissues and organism
- ❑ **Proteome**, the collection of proteins produced, evolves dynamically during time depending on environmental signals.
- ❑ **Proteomic** studies the sequences of proteins and their functionalities

- ❑ Semantic integration of genomic and proteomic databases
 - ▶ Data produced by different labs need to be integrated
 - ▶ Data mining can be used to perform data cleaning, integration, object reconciliation to merge heterogeneous databases
- ❑ Alignment of nucleotide/protein sequences
 - ▶ Build phylogenetic trees
 - ▶ Similarity search
 - ▶ Difference search
- ❑ Protein structure analysis
 - ▶ 3D structure of proteins heavily affects their functionalities
 - ▶ Prediction of protein structures
 - ▶ Discovery of regularities

❑ Association and path analysis of gene sequences

- ▶ Analysis of gene associations in diseases
- ▶ Discovery of sequential patterns of genes correlated to different stages of diseases

❑ Visualization

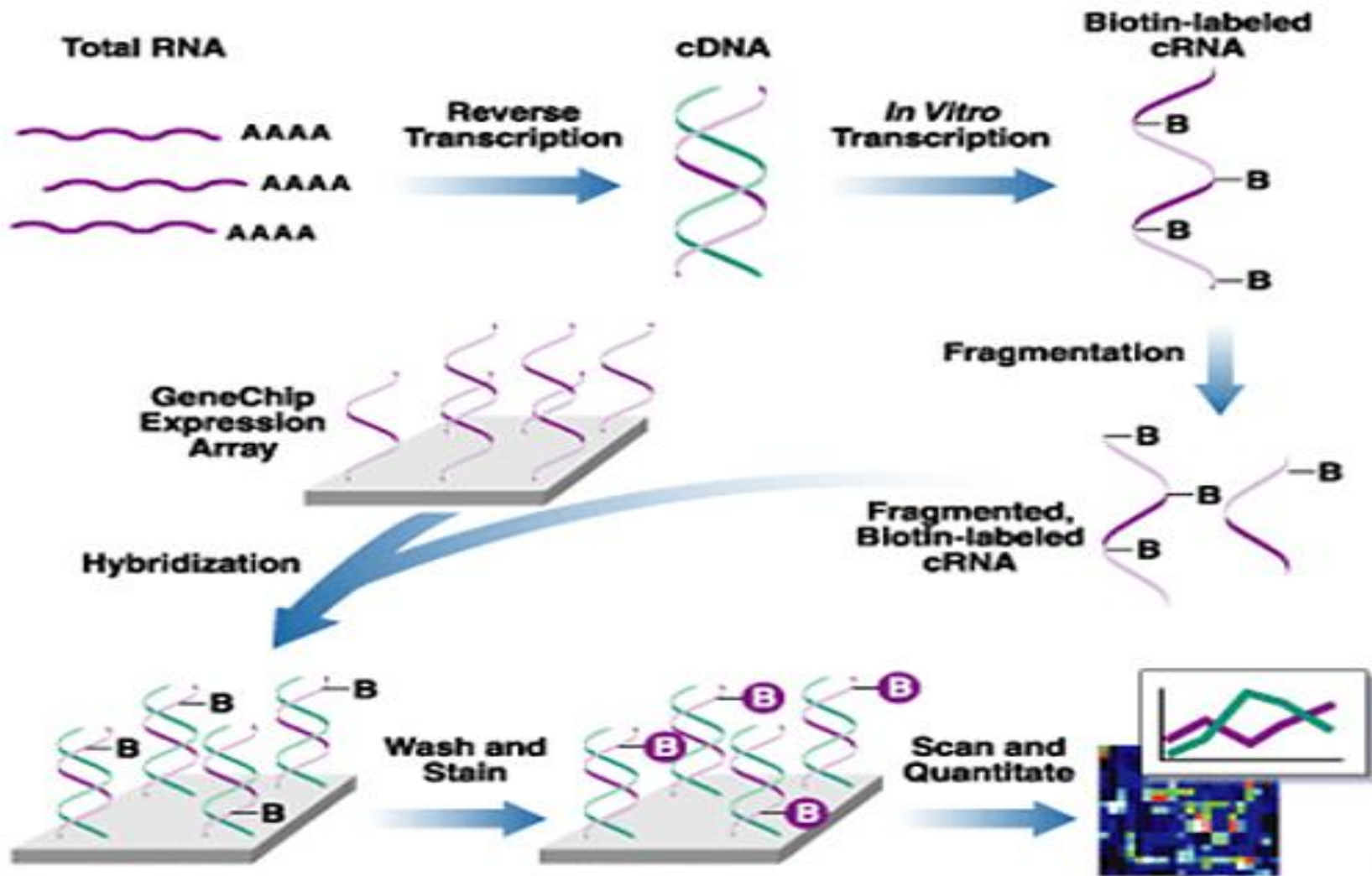
- ▶ Support to knowledge discovery
- ▶ Interactive data exploration

DNA Microarray Analysis

- ❑ The DNA Microarray is a technology that allows the analysis of the gene expression levels in samples collected
- ❑ Such an analysis has many potential applications
 - ▶ Earlier and more accurate diagnostics
 - ▶ New molecular targets for therapy
 - ▶ Improved and individualized treatments
 - ▶ Fundamental biological discovery (e.g. finding and refining biological pathways)
- ❑ Examples
 - ▶ Molecular diagnosis of leukemia, breast cancer, ...
 - ▶ Discovery that genetic signature strongly predicts outcome
 - ▶ A few new drugs, many new promising drug targets

- ❑ Traditional methods in molecular biology generally work on a “one gene in one experiment” basis, which means that the throughput is very limited and the “whole picture” of genes function is hard to obtain
- ❑ “In early 1997, scientists never envisioned looking at more than 25 to 50 gene-expression levels simultaneously. Today everybody tells us that they want to look at the whole genome.” – Kreiner, Affymetrics
- ❑ With a technology for **simultaneously** analyzing the expression levels of **large numbers** of genes we can:
 - ▶ Study the behavior of **co-regulated gene networks**.
 - ▶ Look for groups of **genes involved in a particular biological process** or in a specific disease by identifying genes whose expression levels change under certain circumstances.
 - ▶ Detecting changes in gene expression level in order to have clues on its **product function**.
 - ▶ Compare **normal** organism and **mutant** RNA transcription profiles.

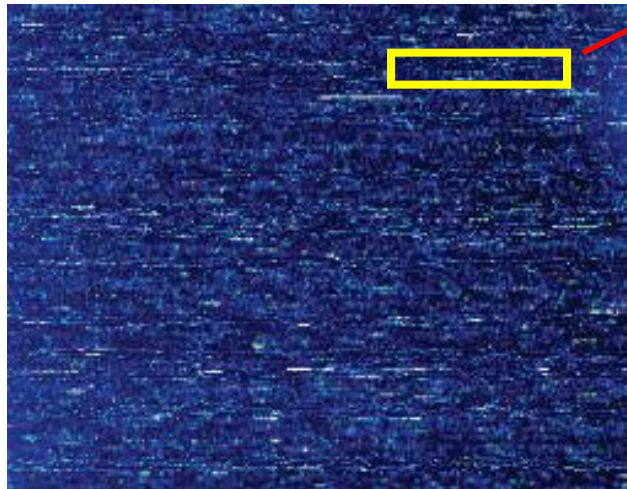
The technology: the whole picture



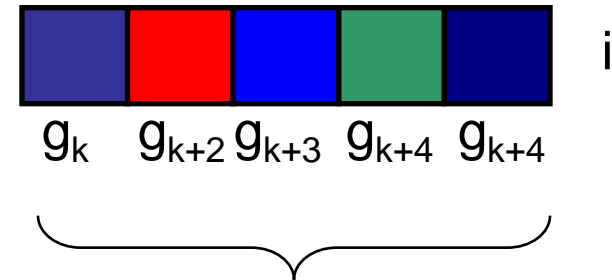
The technology: Affymetrix Chip



The technology: scanning



An observation



Genes expression levels

- ❑ Gene Selection
 - ▶ Find genes for therapeutic targets (new drugs)
- ❑ Classification (supervised)
 - ▶ Identify disease
 - ▶ Predict outcome / select best treatment
- ❑ Clustering (unsupervised)
 - ▶ Find new biological classes / refine existing ones
 - ▶ Exploration (discovery of unknown classes)

❑ Main challenges

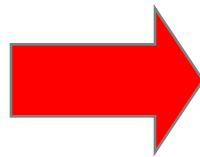
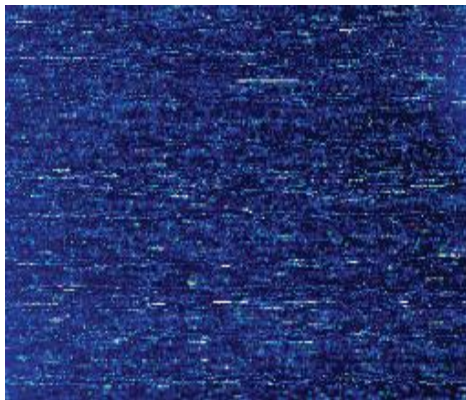
- ▶ Few samples (usually < 100) but many features (usually genes > 1000)
- ▶ High probability of finding **false positives**, that are knowledge discovered due to random noise
- ▶ Models discovered need to be explainable to biologists

❑ Main steps

- ▶ Data preparation
- ▶ Feature selection
- ▶ Apply a classification methods
- ▶ Tuning parameters with cross-validation

Preparing data

- ❑ Microarray data is translated in a **$n \times p$ table**, where n is the number of observations and p is the number of genes tested
- ❑ Each element $\langle i, j \rangle$ of the table is the expression level of gene j in the observation i
- ❑ Thresholds and transformations are applied to data
- ❑ Genes with a not significant variability through the whole dataset are excluded



	Gene 1	Gene 2	Gene 3
Sample 1	104	3208	40
Sample 2	32	1095	41

Genes Selection

- ❑ Most learning algorithms look for non-linear combinations of features
 - ▶ Can easily find *spurious* combinations given few records and many genes (“false positives problem”)
- ❑ Classification accuracy improves if we first reduce number of genes by a linear method
 - ▶ e.g. T-values of mean difference

$$\frac{(\text{Avg}_1 - \text{Avg}_2)}{\sqrt{(\sigma_1^2 / N_1 + \sigma_2^2 / N_2)}}$$

- ❑ Select the top N genes from each class

Genes Selection: Randomization Approach

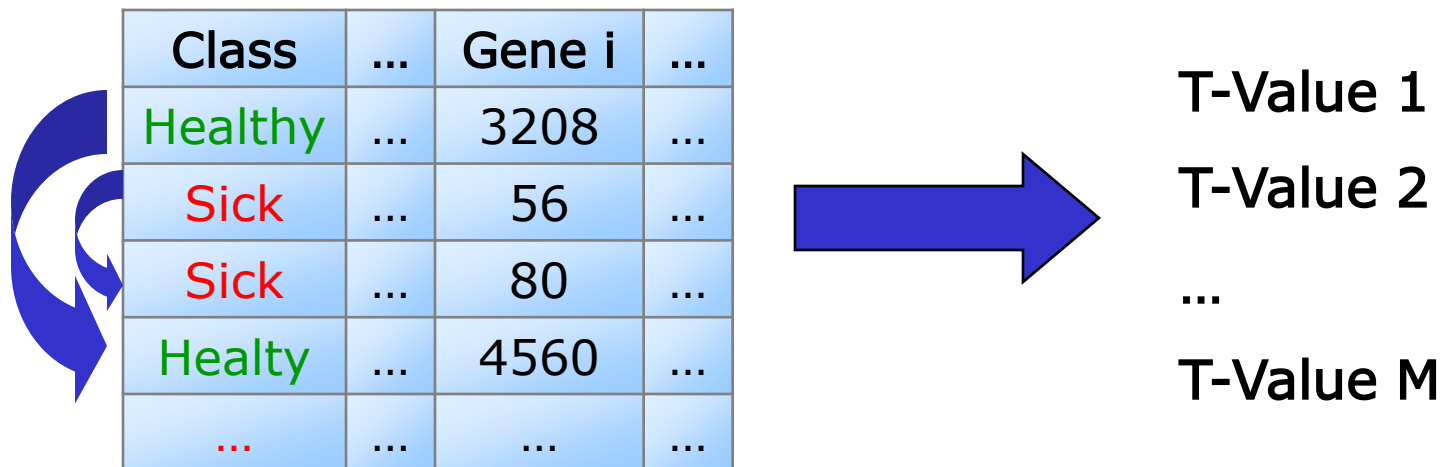
Class	...	Gene i	...
Healthy	...	3208	...
Sick	...	56	...
Sick	...	80	...
Healthy	...	4560	...
...



T-Value

❑ Is T-Value outcome due to chance ?

Genes Selection: Randomization Approach



- ☐ Is T-Value outcome due to chance ?
- ☐ Randomization approach
 - ▶ Generate M random permutations of the class column
 - ▶ Compute T-values for each permutation and for each gene
 - ▶ How frequent a big T-value occurs for a random permutation?
 - ▶ Keep genes with high T-value and desired significance
- ☐ Limitations
 - ▶ Genes are assumed independent
 - ▶ Randomization is a conservative approach

- ❑ Generate several models and evaluate them
 - ▶ Apply T-Values to identify the top N genes
 - ▶ Evaluate (with cross-validation) the accuracy of the model learned using all the subset of genes selected
 - ▶ Choose the simplest model that reaches the best performance
- ❑ Issues
 - ▶ Computationally expensive
 - ▶ Validation sets used in the genes selection process cannot be used to assess the final performance of the model!

- ❑ Decision Trees/Rules
 - ▶ Model easy to understand
 - ▶ Find smallest gene sets, but not robust
 - ▶ Poor performance
- ❑ Neural Networks
 - ▶ Work well for reduced number of genes
 - ▶ Model is difficult to understand
- ❑ K-nearest neighbor
 - ▶ Good results for small number of genes, but no model
- ❑ Naïve Bayes
 - ▶ Simple, robust, but ignores gene interactions
- ❑ Support Vector Machines (SVM)
 - ▶ Good accuracy, automatic gene selection, but hard to understand
- ❑ ...

Biological Sequence Alignment

Alignment of biological sequence (1)

- ❑ Given two or more input biological sequences, **identify similar sequences with long conserved subsequences**
- ❑ Sequences can be either nucleotides (DNA/RNA) or amino acids (proteins)
 - ▶ Nucleotides align with if they are identical
 - ▶ Amino acids align if identical or if one can be derived from the other
- ❑ **Tasks**
 - ▶ Pairwise sequence alignment
 - ▶ Multiple sequence alignment
- ❑ **Applications**
 - ▶ Discovering phylogentic trees
 - ▶ Similarity searches

- ❑ **Substitution matrix** is used to define
 - ▶ cost of substitutions
 - ▶ cost of insertions and deletions
- ❑ Cost is inversely proportional to the probability that a substitution/insertion/deletion occurred
- ❑ Gaps ("—") can be used to indicate positions where it is preferable not to align two symbols
- ❑ The introduction of a gap ("—") is usually associated to a negative cost (**penalty**)

Example

- Align the following sequences:

HEAGAWGHEE

PAWHEAE

- Evaluate the following alignments according to the substitution matrix provided and the a gap penalty of -8

	A	E	G	H	W
A	5	-1	0	-2	-3
E	-1	6	-3	0	-3
H	-2	0	-2	10	-3
P	-1	-1	-2	-2	-4
W	-3	-3	-3	-3	15

<i>H</i>	<i>E</i>	<i>A</i>	<i>G</i>	<i>A</i>	<i>W</i>	<i>G</i>	<i>H</i>	<i>E</i>	—	<i>E</i>	
<i>P</i>	—	<i>A</i>	—	—	<i>W</i>	—	<i>H</i>	<i>E</i>	<i>A</i>	<i>E</i>	
-2	-8	+5	-8	-8	+15	-8	+10	+6	-8	+6	= 0
<i>H</i>	<i>E</i>	<i>A</i>	<i>G</i>	<i>A</i>	<i>W</i>	<i>G</i>	<i>H</i>	<i>E</i>	—	<i>E</i>	
—	—	<i>P</i>	—	<i>A</i>	<i>W</i>	—	<i>H</i>	<i>E</i>	<i>A</i>	<i>E</i>	
-8	-8	-1	-8	+5	+15	-8	+10	+6	-8	+6	= +1

- ❑ Two major approaches
 - ▶ **Local alignment**, works on segments and merge them
 - ▶ **Global alignment**, works on entire sequence
- ❑ Global alignment approaches search for the optimal alignment starting from optimal subsequences
- ❑ *Needleman-Wunsch* and *Smith-Waterman* algorithms exploit **dynamic programming** to find the optimal solution
- ❑ Both these algorithm have a **computational complexity** that is **quadratic w.r.t. sequences length!**
- ❑ Local alignment approaches (e.g. BLAST and FASTA) may be not able to find the best alignment but are more suitable to deal with long sequences

- ❑ BLAST breaks the sequences in small fragments called **words**
- ❑ A word is a **k-tuple** of elements (typically 11 nucleotides or 3 amino acids)
- ❑ BLAST first builds an **hash tables** of **neighborhood** words, that are closely matching
- ❑ Starting from a fragment, the alignment is extended in both the direction by choosing the best scoring matches
- ❑ BLAST has computationally complexity linear w.r.t. to the sequence length
- ❑ Several specialized versions of BLAST have been introduced
 - ▶ Protein similarity searches (BLASTP)
 - ▶ Variable word size (BLASTN)
 - ▶ Not contiguous alignments (MEGABLAST)

- ❑ Is important both in phylogenetic analysis and in the discovery of protein structures
- ❑ Multiple alignment is computationally more challenging
- ❑ **Freng-Doolittle alignment**
 - ▶ Performs the pairwise alignments
 - ▶ Merge them following a guide tree generated with a hierarchical clustering approach
- ❑ Hidden Markov Models
 - ▶ More sophisticated probabilistic approach to represent statistical regularities in the sequences