

# ***Bioinformatics Solutions Using R and Bioconductor***

---

Dr. Alla Bulashevskaya  
AG Prof. Jens Timmer

14.07.2009

- C and Perl were the languages of choice for the first generation of bioinformaticians: massive amounts of sequence data
- R is well suited to scientific challenges in **functional genomics** which employs technologies to measure abundances of biomolecules under different conditions
- Since 2001 **Bioconductor** project  
Packages for analyzing high-dimensional (time series) data obtained from high-throughput **functional genomics** assays:  
e.g. expression microarrays, metabolic profiling:
  - Identifying Interesting Genes with **siggenes**
  - R package **limma**: Linear Models for Microarray Data
  - Reverse Engineering Genetic Networks using **GeneNet**
  - Package **GAGE**: Generally Applicable Gene Set Enrichment

## Identifying Interesting Genes with **sigggenes**

- Common task in microarray experiments: identification of genes whose expression values differ substantially between groups or conditions - **differentially expressed genes (degs)**
- **multiple testing** problems in which thousands of hypotheses are tested simultaneously
- Testing statistic and corresponding p-value are computed for each gene
- Raw p-values are adjusted for multiple testing:
  - **Significance Analysis of Microarrays (SAM)**  
observed test statistics are plotted against expected under null hypotheses; points that differ from diagonal correspond to degs
  - **Empirical Bayes Analysis of Microarrays (EBAM)**  
Models the distribution of observed test statistics as mixture of two components: one for degs and the other for not

## R package **limma**: Linear Models for Microarray Data

- A number of summary statistics are computed for each gene and each contrast:
  - moderated t-statistic*: ratio of log<sub>2</sub>-fold change to its standard error  
Standard errors have been moderated across genes
  - p-value*:  
obtained from *moderated t-statistic* after adjustment for multiple testing using Benjamini and Hochberg's method to control **False Discovery Rate (FDR)**
  - B-statistic*: log-odds that the gene is differentially expressed  
B=0 corresponds to 50/50 chance
- *moderated F-statistic*  
for each gene combines the *t-statistics* for all contrasts  
tests whether any of the contrasts are non-zero

## R Package **GAGE**:

### Generally Applicable Gene Set Enrichment

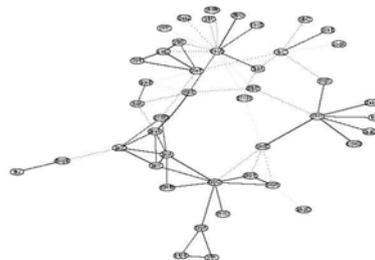
---

- **Gene Set Analysis (GSA)** focuses on sets of related genes
- Incorporates prior knowledge of biological pathways in form of gene sets
- Small coordinated gene expression changes in a pathway can have major biological effect even if these changes are not statistically significant for any individual gene
- Uses all data instead of prefiltering for a short list of strongly differentially expressed genes
- **GAGE** can handle multiple microarray datasets with different sample sizes, experimental designs, profiling techniques

## Reverse Engineering Genetic Networks using **GeneNet**

---

- output - graph where each gene corresponds to a node and edges depict dependencies between nodes
- based on **Graphical Gaussian Model (GGM)** which represent multivariate dependencies by means of *partial correlation*
- model selection - assigning statistical significance to edges using local False Discovery Rate
- exploratory approach that may help to identify „**hubs**“ or **clusters of genes** that are functionally related or co-regulated



**Bioconductor** is more than just a repository of biology-related packages!!!

The project has driven a number of technological innovations:

- Management of package dependence hierarchies
- Interfaces between R and other software systems
- Biological metadata (e.g. genome annotations) packages
- More structured data formats than basic data types of R
- Common data structures that allow efficient exchange of data and computational results between different packages  
(Example is *ExpressionSet* - a class for storage data and info on a microarray experiment)