

Tools and Algorithms in Bioinformatics GCBA815, Fall 2013

Week7: Microarray data analysis BRB tools

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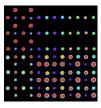
Microarrays

- · GeneChips
 - Contain short (25-mer) oligonucleotides as probes
 - About 400,000 different DNA spots per chip
 - Space for a lot of controls and multiple spots (16-20) per gene
 - Mismatched oligos for background correction



- Concentrated DNA is prepared and spotted on glass slides
- Much cheaper than GeneChips
- Designed to contain only the genes of interest
- Usually limited to about 15,000 spots per slide
- Limits controls and duplicates for each gene

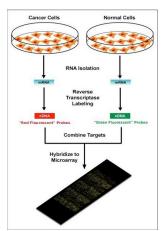






Two-channel vs one-channel detection

- Two-channel (or two-color)
 - Typically used to compare expression in two different treatments
 - Two types of cDNA labeling dyes are used
 - Cy3- has emission at 570nm (corresponds to green)
 - Cy5- has emission at 670nm (corresponds to red)
 - The two Cy-labeled cDNA samples are mixed and hybridized to the a microarray
 - Relative intensities are are used detect up or down-regulated genes



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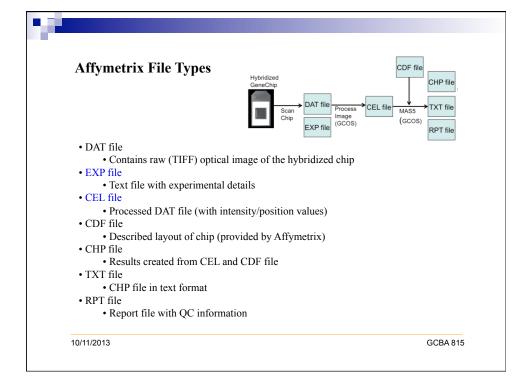
Two-channel vs one-channel detection

- Single-channel (one-color)
 - Uses a single dye to measure intensity for each probe
 - Multiple treatments are hybridized to multiple arrays and the intensities are compared to determine up or down-regulated genes
 - One bad quality sample does not affect the other array results (which is a problem in two-channel arrays
 - Results are easily comparable to arrays from different experiments
 - Requires twice as many microarrays compared to twochannel arrays



Experimental design

- Controls (RNA spike-ins)
- Technical replicates
 - Multiple aliquots of the same sample are run separately to account for technical variation
- · Biological replicates
 - Multiple independent samples of the same kind are run separately to account for sample to sample variation
- Multiple spots for each gene on a single array to measure statistical significance.





Basic statistics

- · Raw intensities
 - Pre-processing, background correction
 - Transformed into expression values (called normalization)
 - Use algorithms such as MAS5, RMA/GCRMA, etc.
- Fold change
 - Ratio of normalized intensities of experimental/control
- Log(intensity) or log(ratio)
 - Brings the differential expression values from multiplicative scale to additive scale
- T-test (Gosset 'student' t-test)
- P-value (obtained from a test statistic such as t-test)

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Log transformations

Normalized Intensity	Natural Log (ln)	Log_2	Log 10
1	0	0	0
10	2.30	3.32	1
20	3	4.32	1.3
100	4.61	6.62	2
1000	6.91	9.97	3
10000	9.21	13.29	4
50000	10.82	15.61	4.7
100000	11.51	16.61	5



Public Repositories of Microarray Datasets

- NCBI's GEO (Gene Expression Omnibus)
 - GEO data sets
 - · Original expression datasets
 - Curated gene expression datasets
 - · Cluster tools and differential expression queries
 - · Geo Profiles
 - · Contains gene expression profiles for each gene
- EBI's Array Express

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Microarray Data Analysis Using BRB-ArrayTools

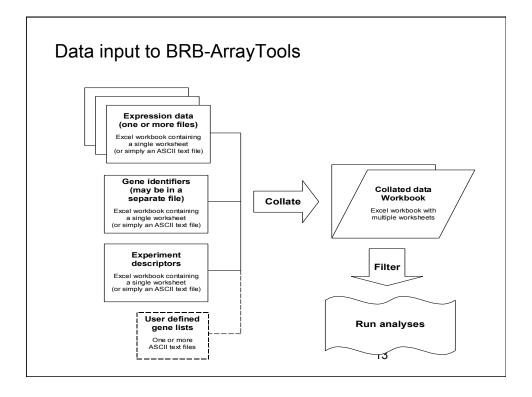
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BRB-ArrayTools

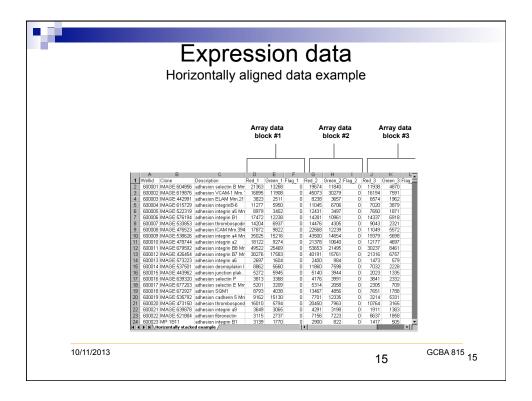
An Integrated Software Tool for DNA Microarray Analysis

- Developed under the direction of Dr. Richard Simon of the Biometrics Research Branch, NCI.
- Software was developed with the purpose of deploying powerful statistical tools for use by biologists.
- Analyses are launched from user-friendly Excel interface. Also requires installation of a free software called R for running back-end programs.



Expression data

- Input data as tab-delimited ASCII files (or Excel spreadsheets) in one of the following three formats:
 - 1. Horizontally aligned
 - 2. Separate files
 - 3. Multi-chip sets
- Files may contain expression data in the form of signal (or single-channel expression summary), dual-channel intensities, or expression ratios (for dual-channel data). Data may or may not have been already log-transformed. Flags, detection call, and spot size may also be used. All other variables will be ignored.
- For Affymetrix data, expression data files should be PROBESETlevel data if using the Data Import Wizard. Affymetrix CEL files should be imported using a specialized utility included with BRB-ArrayTools

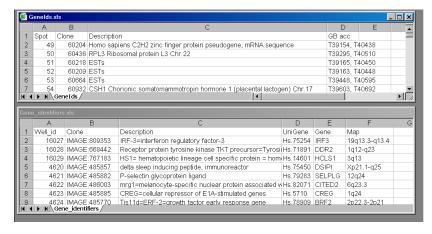


Gene identifiers

- A gene identifiers file is optional, but highly recommended for annotation purposes.
- Gene identifiers which may be used for hyperlinking are: clone ids, UniGene cluster id or gene symbol, GenBank accessions, and probe set ids.

Gene identifiers

Two examples of a gene identifier file



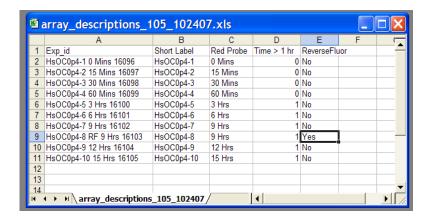
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Experiment (Array) descriptors

- An experiment descriptors file describes the samples used for each array, and is mandatory.
- After the header row, each row in this file represents one array or sample, and each column represents one descriptor variable.
- First column contains array id, subsequent columns contain descriptions, phenotype class labels, patient outcome, and other sample or experiment information.
- A COPY of the original experiment descriptor file will appear in the experiment descriptor sheet of the collated project workbook. The experiment descriptor sheet in the collated project workbook may be further edited as you analyze the data.

Experiment descriptors

Describes the samples used for each array



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Automatic data importers

- General format data: The data import wizard can be used to guide you through the specification of the data components.
- Affymetrix data: Automatically imports data by searching for "Probe Set Name", "Signal" (or "Avg Diff"), and "Detection" (or "Abs_Call") column header labels.
- For importing Affymetrix CEL files, go to the following menu items: (Data Import Wizard), find a data folder containing the .CEL files, and provide an Experiment Descriptors file. Gene identifiers will be imported automatically from the BRB server.
- Can automatically import a GDS dataset from the NCBI Gene Expression Omnibus (GEO) database into BRB-ArrayTools.
- Can directly import dual channel Agilent data into BRB-ArrayTools using the data import wizard.
- Ability to import illumina data using the data import wizard with the lumi package.

Data filtering options

Single-Channel

<u>Intensity filter:</u> May filter out spots with low intensity in single channel or threshold low intensity in forming log intensities.

<u>Detection Call</u>: Exclude a probeset if the Detection call value is "A","M", "P" or "No Call".

Dual channel

Background correction and averaging replicate spots can be performed.

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Data filtering options

Normalization and truncation

- Normalization and truncation steps are applied after data has been spot-filtered, but before screening out genes
- Arrays are normalized before outlying expression levels are truncated.
- Purpose of truncation is primarily to prevent extremely large ratios from being formed by small denominators in dual-channel data. The truncation option is useful if the dual-channel intensities have not been thresholded.

Data filtering options

Normalization and truncation

Normalization:

For single-channel data: Default option is to mediancenter all arrays to a reference array, based on all genes or only a set of housekeeping genes. The reference array may be explicitly chosen, or a "median" array can be automatically found.

■ <u>Truncation:</u> Truncate extreme values (large logintensities for single-channel data, or large absolute log-ratios for dual channel data)

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Data filtering options

Gene filters: Gene variation

- <u>Fold-change filter:</u> Specify a minimum percentage of log-expression values which must meet a specified fold-change criteria
- Log-ratio (or log-intensity) variation filter: Screen genes which do not vary much over the set of samples:
 - 1. Significance criterion compares the variance of each gene against the "average" gene
 - 2. Percentile criterion screens a specified percentage of genes with smallest variance

Data filtering options

Gene filters: Gene quality

- Missing value filter: Screens out genes which contain too many missing values over the set of samples
- Percent absent filter: For Affymetrix data, can filter out a probeset if too many expression values had an Absent call
- Minimum Intensity: This option is only available for single channel data. It filters out genes whose 50th percentile normalized log intensity is less than the log of the user defined value.

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Data filtering options

Gene subsets

- Select genelists for analysis: User may subset the data by selecting one or more genelists to INCLUDE or EXCLUDE. If more than one genelist is selected, then the UNION of all genes on those genelists will be used.
- Specify gene labels to exclude: User may exclude genes based on gene identifier labels. For example, all genes with "Empty" in the gene description field may be excluded.
- CAUTION: Gene subsetting is applied globally to the entire dataset, not just to a specific analysis.
- Probe reduction: Reduce multiple probe sets per gene by choosing the most variably expressed or the maximally expressed probe/probeset.

Some important analysis tools

Finding Genes

Finding differentially expressed genes/gene sets amongst classes.

Prediction

Develop a classifier for predicting the class of a sample

Clustering/Visualizing

Visualizing/Clustering of Genes and Samples.

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Finding Genes

- Comparing classes (Class Comparison)
- Correlated with a quantitative trait (Quantitative Trait Analysis)
- Correlated with survival (Survival Analysis)
- Time Course Analysis (Plug-in) 28

Tools for Class Comparison

- Class Comparison Between groups of arrays
- SAM
- Gene Set Expression Comparison.
- ANOVA models

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Classification of samples

- Cluster analysis vs. classification
- Use cluster analysis to discover new classes, or for visualization purposes
- Use classification when classes are already specified
- Use the Class Prediction tool when the primary interest is to form a classifier to predict the class of new samples.

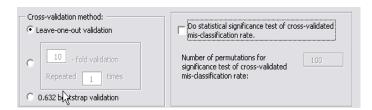
Components of Class Prediction

- C1. Feature(gene) selection
 - -which genes will be included in the model.
- C2. Select model type.
 - -choose prediction method (SVM, k-NN etc) Fit the parameters for the model.
- C3. Evaluating the Classifier
 - Cross-validation

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Cross-validating the classifier

- Leave-One-Out cross validation.
- K-Fold cross validation.
- +0.632 bootstrap cross-validation.



Permutation test

- Use a permutation test to assess the significance of the misclassification rate and univariate significance of each gene
- For each permutation of the class labels, re-run the cross-validation and obtain a new crossvalidated misclassification rate
- The permutation p-value is based upon the rank of the misclassification rate using the original data, compared to all permutations

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Other tools

- Plugins: allows users to create their own tools by writing their own scripts written in the R language which can call C and Fortran programming.
- Methylation tools: find frequently methylated probes and correlate methylation with gene expression.
- Utility tools: annotate data, create gene lists, DrugBank information for significant genes.
- CGHTools: analyze copy number variation.