

## The Bioconductor Project: Open-source Statistical Software for the Analysis of Microarray Data

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EMBO Practical Course on Analysis and Informatics of Microarray Data Wellcome Trust Genome Campus, Hinxton, Cambridge, UK March 18, 2003

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Materials from Bioconductor short courses developed with Robert Gentleman, Rafael Irizarry.

Expanded version of this course: Fred Hutchinson Cancer Research Center, December 2002



### **Statistical computing**

#### Everywhere ...

- Statistical design and analysis:
  - image analysis, normalization, estimation, testing, clustering, prediction, etc.
- Integration of experimental data with biological metadata from WWW-resources
  - gene annotation (GenBank, LocusLink);
  - literature (PubMed);
  - graphical (pathways, chromosome maps).

## Outline

- Overview of the Bioconductor project
- Annotation
- Visualization
- Pre-processing: spotted and Affy arrays
- Differential gene expression
- Clustering and classification

## Acknowledgments

- Bioconductor core team
- Vince Carey, Biostatistics, Harvard
- Yongchao Ge, Statistics, UC Berkeley
- Robert Gentleman, Biostatistics, Harvard
- Jeff Gentry, Dana-Farber Cancer Institute
- Rafael Irizarry, Biostatistics, Johns Hopkins
- Yee Hwa (Jean) Yang, Biostatistics, UCSF
- Jianhua (John) Zhang, Dana-Farber Cancer Institute

### References

#### Personal webpage

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articles and talks on: image analysis; normalization; identification of differentially expressed genes; cluster analysis; classification.

- Bioconductor <u>www.bioconductor.org</u>
  - software, data, and documentation (vignettes);
  - training materials from short courses;
  - mailing list.
- R <u>www.r-project.org</u>
  - software; documentation; RNews.

 Bioconductor is an open source and open development software project for the analysis and comprehension of biomedical and genomic data.

• Software, data, and documentation are available from <a href="https://www.bioconductor.org">www.bioconductor.org</a>.

- The project was started in the Fall of 2001 by Robert Gentleman, at the Biostatistics Unit of the Dana Farber Cancer Institute.
- There are currently 21 core developers, at various institutions in the US and Europe.
- R and the R package system are used to design and distribute software (<u>www.r-project.org</u>).
- First release (v 1.0): May 2<sup>nd</sup>, 2002, 15 packages.
- Second release (v 1.1): November 18<sup>th</sup>, 2002, 5 new packages.

There are two main classes of packages

- End-user packages:
  - aimed at users unfamiliar with R or computer programming;
  - polished and easy to use interfaces to a wide variety of computational and statistical methods for the analysis of genomic data.
- Developer packages: aimed at software developers, in the sense that they provide ``software to write software".

#### **Bioconductor packages** Release 1.1, November 18<sup>th</sup>, 2002

• General infrastructure:

Biobase, reposTools, rhdf5, tkWidgets.

• Annotation:

annotate, AnnBuilder  $\rightarrow$  data packages.

• Graphics:

geneplotter, hexbin.

- Pre-processing for Affymetrix oligonucleotide chip data: affy, vsn, CDF packages.
- Pre-processing for spotted DNA microarray data: marrayClasses, marrayInput, marrayNorm, marrayPlots, marrayTools, vsn.
- Differential gene expression:

edd, genefilter, multtest, ROC.

• Graphs:

graph.

# **Ongoing efforts**

- Variable (feature) selection;
- Prediction;
- Cluster analysis;
- Cross-validation;
- Multiple testing;
- Quality measures for microarray data;
- Interactions with MAGE-ML;
- Biological sequence analysis;
- Etc.

## **Computing needs**

- Mechanisms for facilitating the design and deployment of portable, extensible, and scalable software.
- Support for interoperability with software written in other languages.
- Tools for integrating biological metadata from the WWW in the analysis of experimental metadata.
- Access to a broad range of statistical and numerical methods.
- High-quality visualization and graphics tools that support interactivity.
- An effective, extensible user interface.
- Tools for producing innovative, high-quality documentation and training materials.
- Methodology that supports the creation, testing, and distribution of software and data modules.

 Interactive tools for linking experimental data in real time, to biological metadata from WWW resources.

E.g. PubMed, GenBank, LocusLink.

- Scenario. Normalize spotted array data with marrayNorm, obtain list of differentially expressed genes from multtest or genefilter, use the annotate package
  - to retrieve and search PubMed abstracts for these genes;
  - to generate an HTML report with links to LocusLink for each gene.

- Widgets. Small-scale graphical user interfaces (GUI), providing point & click access for specific tasks (tkWidgets).
- E.g. File browsing and selection for data input, basic analyses.
- Object-oriented class/method design. Allows efficient representation and manipulation of large and complex biological datasets of multiple types (cf. MIAME standards).

## Object-oriented programming

 The Bioconductor project has adopted the object-oriented programming – OOP – paradigm presented in

J. M. Chambers (1998). Programming with Data.

- Tools for programming using the class/method mechanism are provided in the R methods package.
- Tutorial: <a href="http://www.omegahat.org/RSMethods/index.html">www.omegahat.org/RSMethods/index.html</a>

# OOP

- A class provides a software abstraction of a real world object. It reflects how we think of certain objects and what information these objects should contain.
- Classes are defined in terms of slots which contain the relevant data.
- An object is an instance of a class.
- A class defines the structure, inheritance, and initialization of objects.

## OOP

- A method is a function that performs an action on data (objects).
- Methods define how a particular function should behave depending on the class of its arguments.
- Methods allow computations to be adapted to particular data types, i.e., classes.
- A generic function is a dispatcher, it examines its arguments and determines the appropriate method to invoke.
- Examples of generic functions include plot, summary, print.

### Data

- Issues:
  - complexity;
  - size;
  - evolution.

• We distinguish between biological metadata and experimental metadata.

## **Experimental metadata**

- Gene expression measures
  - scanned images, i.e., raw data;
  - image quantitation data, i.e., output from image analysis;
  - normalized expression measures, i.e., log ratios M or Affy measures.
- Reliability information for the expression measures.
- Information on the probe sequences printed on the arrays (array layout).
- Information on the target samples hybridized to the arrays.
- See Minimum Information About a Microarray Experiment – MIAME – standards.

## **Biological metadata**

- Biological attributes that can be applied to the experimental data.
- E.g. for genes
  - chromosomal location;
  - gene annotation (LocusLink, GO);
  - relevant literature (PubMed).
- Biological metadata sources are large, complex, evolving rapidly, and typically distributed via the WWW.

### marrayRaw class

#### **Pre-normalization intensity data for a batch of arrays**



## AffyBatch class

#### **Probe-level intensity data for a batch of arrays (same CDF)**



## exprSet class



#### Reading in phenoData

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	FALSE MALE
Back	Finish
	Sample Names

MINNE Information	
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Laboratory:	
Contact Information:	
Experiment Title:	
Experiment Description:	
URL:	
	Exit

#### tkphenoData



## Pedagogy

Extensive documentation and training resources for R and Bioconductor are available on the WWW.

- R manuals and tutorials are available from the R website.
- R help system
  - detailed on-line documentation, available in text, HTML, PDF, and LaTeX formats;
  - e.g. help(genefilter), ?pubmed.
- R demo system
  - user-friendly interface for running demonstrations of R scripts;
  - e.g. demo (marrayPlots), demo (affy).
- Bioconductor short courses
  - modular training segments on software and statistical methodology;
  - lectures and computer labs available on WWW for self-instruction.

## Vignettes

- Bioconductor has adopted a new documentation paradigm, the vignette.
- A vignette is an executable document consisting of a collection of documentation text and code chunks.
- Vignettes form dynamic, integrated, and reproducible statistical documents that can be automatically updated if either data or analyses are changed.
- Vignettes can be generated using the Sweave function from the R tools package.

## Vignettes

- Each Bioconductor package contains at least one vignette, located in the doc subdirectory of an installed package and accessible from the help browser.
- Vignettes provide task-oriented descriptions of the package's functionality and can be used interactively.
- Vignettes are available separately from the Bioconductor website or as part of the packages.

## Vignettes

- Tools are being developed for managing and using this repository of step-by-step tutorials
  - Biobase: openVignette Menu of available vignettes and interface for viewing vignettes (PDF).
  - tkWidgets: vExplorer Interactive use of vignettes.
  - reposTools.

### Sweave

- The Sweave system allows the generation of integrated statistical documents intermixing text, code, and code output (textual and graphical).
- Functions are available in the R tools package.
- See ? Sweave and manual www.ci.tuwien.ac.at/~leisch/Sweave/

## **Sweave input**

- <u>Input:</u> a text file which consists of a sequence of code and documentation chunks, or segments (noweb file).
  - Documentation chunks
    - start with @
    - can be text in a markup language like LaTeX.
  - Code chunks
    - start with <<name>>=
    - can be R or S-Plus code.
  - File extension: .rnw, .Rnw, .snw, .Snw.

## Sweave output

- <u>Output:</u> a single document, e.g., .tex file or .pdf file containing
  - the documentation text,
  - the R code,
  - the code output: text and graphs.
- The document can be automatically regenerated whenever the data, code, or documentation text change.
- **Stangle** or **tangleToR**: extract only the code.



#### Annotation

### annotate package

- One of the largest challenges in analyzing genomic data is associating the experimental data with the available biological metadata, e.g., sequence, gene annotation, chromosomal maps, literature.
- Bioconductor provides two main packages for this purpose:
  - annotate (end-user);
  - AnnBuilder (developer).

## WWW resources

- Nucleotide databases: e.g. GenBank.
- Gene databases: e.g. LocusLink, UniGene.
- Protein sequence and structure databases: e.g. SwissProt, Protein DataBank (PDB).
- Literature databases: e.g. PubMed, OMIM.
- Chromosome maps: e.g. NCBI Map Viewer.
- Pathways: e.g. KEGG.
- Entrez is a search and retrieval system that integrates information from databases at NCBI (National Center for Biotechnology Information).
Important tasks

- Associate manufacturers or in-house probe identifiers to other available identifiers.
  - E.g.

Affymetrix IDs  $\rightarrow$  LocusLink LocusID Affymetrix IDs  $\rightarrow$  GenBank accession number.

- Associate probes with biological data such as chromosomal position, pathways.
- Associate probes with published literature data via PubMed (need PMID).

Affymetrix identifier HGU95A chips	~41046_s_at″
LocusLink, LocusID	``9203 <i>''</i>
GenBank accession #	``X95808″
Gene symbol	"ZNF261"
PubMed, PMID	<pre>``10486218" ``9205841" ``8817323"</pre>
Chromosomal location	"X", "Xq13.1"

### Annotation data packages

- The Bioconductor project provides packages that contain only data.
- These data packages are built using **AnnBuilder**.
- They can be downloaded from the Bioconductor website and also using update.packages.
- Data packages contain many different mappings to interesting data.
  - Mappings between Affy IDs and other probe IDs: hgu95a for HGU95A GeneChip series, also, hgu133a, hu6800, mgu74a, rgu34a.
  - Affy CDF data packages.
- The packages are updated and expanded regularly as updated and new data become available.

- Much of what annotate does relies on matching symbols.
- This is basically the role of a hash table in most programming languages.
- In R, we rely on environments.
- The annotation data packages provide R environment objects containing key and value pairs for the mappings between two sets of probe identifiers.
- Keys can be accessed using the R 1s function.
- Matching values in different environments can be accessed using the get or multiget functions.

> library(hgu95a) > get("41046 s at", env = hgu95aACCNUM) [1] "X95808" > get("41046 s at", env = hgu95aLOCUSID) [1] "9203" > get("41046 s at", env = hgu95aSYMBOL) [1] "ZNF261" > get("41046 s at", env = hgu95aGENENAME) [1] "zinc finger protein 261" > get("41046 s at", env = hgu95aSUMFUNC) [1] "Contains a putative zinc-binding motif (MYM) | Proteome" > get("41046 s at", env = hgu95aUNIGENE) [1] "Hs.9568"

> get("41046 s at", env = hgu95aCHR) [1] "X" > get("41046 s at", env = hgu95aCHRLOC) [1] "66457019@X" > get("41046 s at", env = hgu95aCHRORI) [1] "-@X" > get("41046 s at", env = hgu95aMAP) [1] "Xq13.1" > get("41046 s at", env = hgu95aPMID) [1] "10486218" "9205841" "8817323" > get("41046 s at", env = hgu95aGO)[1] "GO:0003677" "GO:0007275"

- Instead of relying on the general R functions for environments, new userfriendly functions have been written for accessing and working with specific identifiers.
- E.g. getGO, getGOdesc, getLL, getPMID, getSYMBOL.

- > getSYMBOL("41046\_s\_at",data="hgu95a")
   41046\_s\_at
   "ZNF261"
- > gg<- getGO("41046\_s\_at",data="hgu95a")</pre>
- > getGOdesc(gg, "MF")
   \$"c("GO:0003677", "GO:0007275")"
   [1] "DNA binding"
- > getLL("41046\_s\_at",data="hgu95a")
   41046\_s\_at
   9203
- > getPMID("41046\_s\_at",data="hgu95a")
   \$"41046\_s\_at"
   [1] 10486218 9205841 8817323

### annotate: querying databases

The **annotate** package provides tools for

- Searching and processing information from various WWW biological databases
  - GenBank,
  - LocusLink,
  - PubMed.
- Regular expression searching of PubMed abstracts.
- Generating nice HTML reports of analyses, with links to biological databases.

### annotate: WWW queries

 Functions for querying WWW databases from R rely on the browseURL function

browseURL("www.r-project.org")

• The **XML** package is used to parse query results.

### annotate: querying GenBank

www.ncbi.nlm.nih.gov/Genbank/index.html

- Given a vector of GenBank accession numbers or NCBI UIDs, the genbank function
  - opens a browser at the URLs for the corresponding GenBank queries;
  - returns an **XMLdoc** object with the same data.

genbank("X95808",disp="browser")

http://www.ncbi.nih.gov/entrez/query.fcgi?tool=bioconductor&cmd=Search&db=Nucleotide&term=X95808

genbank(1430782,disp="data",
 type="uid")

### annotate: querying LocusLink

www.ncbi.nlm.nih.gov/LocusLink/

 locuslinkByID: given one or more LocusIDs, the browser is opened at the URL corresponding to the first gene.

locuslinkByID("9203")

http://www.ncbi.nih.gov/LocusLink/LocRpt.cgi?l=9203

• **locuslinkQuery**: given a search string, the results of the LocusLink query are displayed in the browser.

locuslinkQuery("zinc finger")
http://www.ncbi.nih.gov/LocusLink/list.cgi?Q=zinc finger&ORG=Hs&V=0

### annotate: querying PubMed

www.ncbi.nlm.nih.gov

- For any gene there is often a large amount of data available from PubMed.
- The annotate package provides the following tools for interacting with PubMed
  - pubMedAbst: a class structure for PubMed abstracts in R.
  - pubmed: the basic engine for talking to PubMed.

### annotate: pubMedAbst class

Class structure for storing and processing PubMed abstracts in R

- pmid
- authors
- abstText
- articleTitle
- journal
- pubDate
- abstUrl

### annotate: high-level tools for querying PubMed

- pm.getabst: download the specified PubMed abstracts (stored in XML) and create a list of pubMedAbst objects.
- **pm.titles**: extract the titles from a list of PubMed abstracts.
- **pm.abstGrep**: regular expression matching on the abstracts.

### annotate: PubMed example

pmid <-get("41046\_s\_at", env=hgu95aPMID)
pubmed(pmid, disp="browser")</pre>

http://www.ncbi.nih.gov/entrez/query.fcgi?tool=bioconductor&cmd=Retrie ve&db=PubMed&list\_uids=10486218%2c9205841%2c8817323

absts <- pm.getabst("41046\_s\_at", base="hgu95a")

pm.titles(absts)

pm.abstGrep("retardation",absts[[1]])

### annotate: PubMed example

RGui - [R Console]				
R File Edit Misc Packages Windows Help				_ B ×
Slot "articleTitle": [1] "Prediction of the coding sequences of unidenti	fied human genes. VII.	The complete sequences of	of 100 new cDNA clones from brain whic	h can\$
Slot "journal": [1] "DNA Res"				
Slot "pubDate": [1] "Apr 1997"				
Slot "abstUrl": [1] "No URL Provided"				
[[3]] An object of class "pubMedAbst" Slot "authors": [1] "S M SM van der Maarel" "I H IH Scholten"	"I I Huber"	"C C Philippe"	"R F RF Suijkerbuijk"	
[6] "S S Gilgenkrantz" "J J Kere"	"F P FP Cremers"	"H H HH Ropers"		
Slot "abstText": [1] "In several families with non-specific X-linked	d mental retardation (X	LMR) linkage analyses hav	ve assigned the underlying gene defect	to t\$
Slot "articleTitle": [1] "Cloning and characterization of DXS6673E, a ca	andidate gene for X-lin	ked mental retardation in	n Xq13.1."	
Slot "journal": [1] "Hum Mol Genet"				
Slot "pubDate": [1] "Jul 1996"				
Slot "abstUrl": [1] "No URL Provided"				
> pm.titles(absts) [[1]]				
<ol> <li>"Cloning and mapping of members of the MYM fami [2] "Prediction of the coding sequences of unidenti</li> </ol>		The complete sequences (	of 100 new cDNA clones from brain whic	\$ h can\$
[3] "Cloning and characterization of DXS6673E, a ca				\$
<pre>&gt; pm.abstGrep("retardation",absts[[1]]) [1] TRUE FALSE TRUE &gt;</pre>				-
D 1 5 1 A Longuage and Environment				

### annotate: PubMed HTML report

• The new function pmAbst2html takes a list of pubMedAbst objects and generates an HTML report with the titles of the abstracts and links to their full page on PubMed.

pmAbst2html(absts[[1]],filename="pm.htm
1")

Image: Search Netscape     Print     Security     Shop     Stop		
V Bookmarks 🙏 Location: file:///Cl/Sandrine/Current/Talks/EMBD03/pm.html	▼ ( What's Relat	
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<b>BioConductor Abstract List</b>		
Article Title	Publication Date	
Conditional targeting of the DNA repair enzyme hOGG1 into mitochondria.	Nov 2002	
nter-individual variation, seasonal variation and close correlation of OGG1 and ERCC1 mRNA levels in ull blood from healthy volunteers.	Sep 2002	
A limited association of OGG1 Ser326Cys polymorphism for adenocarcinoma of the lung	May 2002	
Protection of human lung cells against hyperoxia using the DNA base excision repair genes hOgg1 and <u>Ppg.</u>	Jul 2002	
The human OGG1 DNA repair enzyme and its association with orolaryngeal cancer risk.	Jul 2002	
Iuman OGG1 undergoes serine phosphorylation and associates with the nuclear matrix and mitotic hromatin in vivo.	Jun 2002	
hOGG1 Ser(326)Cys polymorphism and modification by environmental factors of stomach cancer risk in Chinese.	Jun 2002	
Association of the hOGG1 Ser326Cys polymorphism with lung cancer risk.	Apr 2002	
Reciprocal "flipping" underlies substrate recognition and catalytic activation by the human 8-oxo-guanine DNA glycosylase.	Mar 2002	
Expression of 8-oxoguanine DNA glycosylase is reduced and associated with neurofibrillary tangles in Alzheimer's disease brain.	Jan 2002	
Structure and chromosome location of human OGG1.	Month 1999	
Expression and differential intracellular localization of two major forms of human 8-oxoguanine DNA lycosylase encoded by alternatively spliced OGG1 mRNAs.	May 1999	
Genetic polymorphisms and alternative splicing of the hOGG1 gene, that is involved in the repair of 8-hydroxyguanine in damaged DNA.	Jun 1998	
Augmented expression of a human gene for 8-oxoguanine DNA glycosylase (MutM) in B lymphocytes of the dark zone in lymph node germinal centers.	Nov 1997	
Opposite base-dependent reactions of a human base excision repair enzyme on DNA containing 7,8-dihydro-8-oxoguanine and abasic sites.	Oct 1997	
Molecular cloning and functional expression of a human cDNA encoding the antimutator enzyme 3-hydroxyguanine-DNA glycosylase.	Jul 1997	
Cloning and characterization of hOGG1, a human homolog of the OGG1 gene of Saccharomyces cerevisiae.	Jul 1997	
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## pmAbst2html function from annotate package

pm.html

Document: Done

**-**

### annotate: analysis reports

- A simple interface, <u>ll.htmlpage</u>, can be used to generate an HTML report of analysis results.
- The page consists of a table with one row per gene, with links to LocusLink.
- Entries can include various gene identifiers and statistics.

#### **BioConductor Gene Listing**

#### Golub et al. data, genes with permutation maxT adjusted p-value < 0.01

Locus Link Genes

LocusID	Gene name	Chromosome	ALL mean	AML mean	t-statistic	raw p-value	adj p-value
7 <u>91</u>	X95735_at	7	-0.295	1.59	-10.6	2e-05	2e-05
<u>71</u>	M27891_at	20	-0.81	2.08	-9.78	2e-05	2e-05
<u>84</u>	M55150_at	15	0.488	1.24	-8.03	2e-05	0.00014
<u>067</u>	M16038_at	8	-0.284	1.1	-7.98	2e-05	0.00016
<u>34</u>	 L09209_s_at	11	-0.162	1.36	-7.97	2e-05	2e-04
<u>929</u>	M31523_at	19	0.855	-0.391	7.55	2e-05	5e-04
<u>928</u>	X74262_at	1	0.869	-0.565	7.42	2e-05	0.00078
<u>155</u>	Z15115_at	3	1.94	0.945	7.35	2e-05	0.001
<u>6999</u>	 L47738_at	5	0.734	-0.779	7.31	2e-05	0.00114
602	 U22376_cds2_s_at	6	1.86	0.294	7.28	2e-05	0.00116
<u>5108</u>	 HG1612-HT1612_at		1.91	0.888	7.11	2e-05	0.0017
4	 M91432_at	1	0.431	-0.771	7.08	2e-05	0.0018
925	 L41870_at	13	-0.438	-1.3	7.08	2e-05	0.0018
<u>46</u>	 U72936_s_at	NA	-0.097	-1.07	7.07	2e-05	0.0018
<u>430</u>	X51521_at	6	1.92	1.07	7.06	2e-05	0.00186
056		5	0.71	1.51	-6.97	2e-05	0.00232
<u>4741</u>	 Y12670_at	1	-0.167	0.892	-6.96	2e-05	0.00238
203	 X74801_at	1	0.611	-0.183	6.95	2e-05	0.00238
<u>576</u>	 Y00787_s_at	4	-0.371	2.32	-6.87	2e-05	0.00288
<u>'09</u>	J05243_at	9	0.413	-0.982	6.86	2e-05	0.00288
7 <u>25</u>	 U26266_s_at	19	-0.209	-1.16	6.85	4e-05	0.00294
205	 U82759_at	7	-0.64	0.504	-6.82	2e-05	0.00306
5	 M23197_at	19	-0.881	0.354	-6.79	2e-05	0.0033
509	 M63138_at	11	1.21	2.12	-6.77	2e-05	0.00344
9 <u>55</u>	 M12959_s_at	14	1.13	0.132	6.76	2e-05	0.00352
<u>57</u>	 X62654_ma1_at	12	0.0513	1.33	-6.76	2e-05	0.00352
<u>341</u>	 X07743_at	2	-0.959	0.535	-6.74	2e-05	0.00378
<u>40465</u>	 M31211_s_at	12	0.108	-0.953	6.71	2e-05	0.00404
<u>336</u>	U62136_at	8	-0.163	-0.92	6.68	2e-05	0.00428
<u>560</u>	 X15949_at	4	-0.541	-1.33	6.61	2e-05	0.00492
	1122014 -+	1.4	0.026	n 260	к к1	2. 05	0.00402

11.htmlpage
function from
annotate
package

4

#### genelist.html

100%

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### annotate: chromLoc class

Location information for <u>one gene</u>

- chrom: chromosome name.
- **position**: starting position of the gene in bp.
- **strand**: chromosome strand +/-.

## annotate: chromLocation class

Location information for a set of genes

- **species**: species that the genes correspond to.
- datSource: source of the gene location data.
- **nChrom**: number of chromosomes for the species.
- chromNames: chromosome names.
- **chromLocs**: starting position of the genes in bp.
- chromLengths: length of each chromosome in bp.
- **geneToChrom**: hash table translating gene IDs to location.

Function buildChromClass.

### Visualization

### geneplotter: cPlot



### geneplotter: alongChrom



### geneplotter: alongChrom



### Pre-processing DNA microarray data

### **Pre-processing**

- **affy**: Affymetrix oligonucleotide chips.
- **marray**: Spotted DNA microarrays.
- **vsn**: Variance stabilization for both types of arrays.

Reading in intensity data, diagnostic plots, normalization, computation of expression measures.

The packages start with very different data types, but produce similar objects of class **exprSet**.

One can then use other Bioconductor packages, e.g., genefilter, geneplotter.

### marray packages



Save data to file using write.exprs or continue analysis using other Bioconductor packages



Save data to file using write.exprs or continue analysis using other Bioconductor packages

### Pre-processing: spotted DNA microarrays



### Normalization

- After image processing, we have measures of the red and green fluorescence intensities, R and G, for each spot on the array.
- Normalization is needed to ensure that differences in intensities are indeed due to differential expression, and not some printing, hybridization, or scanning artifact.
- Normalization is necessary before any analysis which involves within or between slides comparisons of intensities, e.g., clustering, classification, testing.

### Normalization

- Identify and remove the effects of systematic variation in the measured fluorescence intensities, other than differential expression, for example
  - different labeling efficiencies of the dyes;
  - different amounts of Cy3- and Cy5-labeled mRNA;
  - different scanning parameters;
  - print-tip, spatial, or plate effects, etc.

### Normalization

- The need for normalization can be seen most clearly in self-self hybridizations, where the same mRNA sample is labeled with the Cy3 and Cy5 dyes.
- The imbalance in the red and green intensities is usually not constant across the spots within and between arrays, and can vary according to overall spot intensity, location, plate origin, etc.
- These factors should be considered in the normalization.

# marray: Pre-processing spotted DNA microarray data

#### marrayClasses:

- class definitions for spotted DNA microarray data (cf. MIAME);
- basic methods for manipulating microarray objects: printing, plotting, subsetting, class conversions, etc.

#### • marrayInput:

- reading in intensity data and textual data describing probes and targets;
- automatic generation of microarray data objects;
- widgets for point & click interface.
- **marrayPlots**: diagnostic plots.
- marrayNorm: robust adaptive location and scale normalization procedures.
- **marrayTools**: miscellaneous tools for functional genomics cores facilities at UCB and UCSF.
## marrayLayout class

#### **Array layout parameters**



Total number of spots

Dimensions of grid matrix

Dimensions of spot matrices

Current subset of spots

Plate IDs for each spot

Control status labels for each spot

Any notes

## marrayRaw class

#### **Pre-normalization intensity data for a batch of arrays**



## marrayNorm class

#### Post-normalization intensity data for a batch of arrays



## marrayInput package

- marrayInput provides functions for reading microarray data into R and creating microarray objects of class marrayLayout, marrayInfo, and marrayRaw.
- Input
  - Image quantitation data, i.e., output files from image analysis software.

E.g. .gpr for GenePix, .spot for Spot.

- Textual description of probe sequences and target samples.
  - E.g. gal files, god lists.

## marrayInput package

 Widgets for graphical user interface

widget.marrayLayout,

widget.marrayInfo,

widget.marrayRaw.

🥼 MarrayRaw builder						
Files						
Name of the marrayRaw object:						
swir						
Foreground and background intensities						
Green Foreground Gmean Green Background morphG						
Red Foreground Rmean	Red Background m	orphR				
Weights						
Layout:						
swirl.layout	Browse					
Target Information:						
swirl.samples	Browse					
Gene Information:						
swirl.gnames		Browse				
Notes:						
Layout Target Genes B	uild Quit					

## marrayPlots package

- See demo (marrayPlots).
- Diagnostic plots of spot statistics.
  - E.g. red and green log intensities, intensity log ratios M, average log intensities A, spot area.
    - maImage: 2D spatial color images.
    - maBoxplot: boxplots.
    - maPlot: scatter-plots with fitted curves and text highlighted.
- Stratify plots according to layout parameters such as print-tip-group, plate.
   E.g. MA-plots with loess fits by print-tipgroup.

### **2D** spatial images maImage



#### Cy3 background intensity

#### **Cy5 background intensity**

3

4

- 140

- 130

- 120

- 110

- 95

- 83

- 71

- 60

- 48

- 36

### Boxplots by print-tip-group maBoxplot

Swirl 93 array: pre-normalization log-ratio M



### MA-plot by print-tip-group maPlot $M = \log_2 R - \log_2 G$ , $A = (\log_2 R + \log_2 G)/2$

Swirl 93 array: pre-normalization log-ratio M

Intensity log ratio, M



Average log intensity, A

## marrayNorm package

- maNormMain: main normalization function, allows robust adaptive location and scale normalization for a batch of arrays
  - intensity or A-dependent location normalization (maNormLoess);
  - 2D spatial location normalization (maNorm2D);
  - median location normalization (maNormMed);
  - scale normalization using MAD (maNormMAD);
  - composite normalization;
  - your own normalization function.
- maNorm: simple wrapper function.
   maNormScale: simple wrapper function for scale normalization.

## marrayTools package

- The marrayTools package provides additional functions for handling two-color spotted microarray data (see devel. version).
- The **spotTools** and **gpTools** functions start from Spot and GenePix image analysis output files, respectively, and automatically
  - read in these data into R,
  - perform standard normalization (within print-tipgroup loess),
  - create a directory with a standard set of diagnostic plots (jpeg format), excel files of quality measures, and tab delimited files of normalized log ratios M and average log intensities A.

## swirl dataset

- Microrrays:
  - 8,448 probes (768 controls);
  - 4 x 4 grid matrix;
  - 22 x 24 spot matrices.
- 4 hybridizations: swirl mutant and wild type mRNA.
- Data stored in object of class marrayRaw: data(swirl).
- > maInfo(maTargets(swirl))[,3:4]

experiment Cy3 experiment Cy5

1	swirl	wild type
2	wild type	swirl
3	swirl	wild type
4	wild type	swirl

## Pre-processing: oligonucleotide chips





## **Probe-pair set**



# Terminology

- Each gene or portion of a gene is represented by 16 to 20 oligonucleotides of 25 base-pairs.
- Probe: an oligonucleotide of 25 base-pairs, i.e., a 25-mer.
- Perfect match (PM): A 25-mer complementary to a reference sequence of interest (e.g., part of a gene).
- Mismatch (MM): same as PM but with a single homomeric base change for the middle (13<sup>th</sup>) base (transversion purine <-> pyrimidine, G <->C, A <->T).
- Probe-pair: a (PM,MM) pair.
- Probe-pair set: a collection of probe-pairs (16 to 20) related to a common gene or fraction of a gene.
- Affy ID: an identifier for a probe-pair set.
- The purpose of the MM probe design is to measure non-specific binding and background noise.

# **Affymetrix files**

- Main software from Affymetrix company, *MicroArray Suite - MAS*, now version 5.
- **DAT** file: Image file, ~10^7 pixels, ~50 MB.
- **CEL** file: Cell intensity file, probe level PM and MM values.
- CDF file: Chip Description File. Describes which probes go in which probe sets and the location of probe-pair sets (genes, gene fragments, ESTs).

# affy: Pre-processing Affymetrix data

- Class definitions for probe-level data: AffyBatch, ProbSet, Cdf, Cel.
- Basic methods for manipulating microarray objects: printing, plotting, subsetting.
- Functions and widgets for data input from **CEL** and **CDF** files, and automatic generation of microarray data objects.
- Diagnostic plots: 2D spatial images, density plots, boxplots, MA-plots, etc.

# affy: Pre-processing Affymetrix data

- Background estimation.
- Probe-level normalization: quantile and curve-fitting normalization (Bolstad et al., 2002).
- Expression measures: MAS 4.0 AvDiff, MAS 5.0 Signal, MBEI (Li & Wong, 2001), RMA (Irizarry et al., 2003).
- Main functions: ReadAffy, rma, expresso, express.

## affy classes: AffyBatch

#### **Probe-level intensity data for a batch of arrays (same CDF)**



## affy classes

- **ProbeSet**: PM, MM intensities for individual probe sets.
  - **pm**: matrix of PM intensities for one probe set, rows → 16-20 probes, columns → arrays.
  - mm: matrix of MM intensities for one probe set, rows  $\rightarrow$  16-20 probes, columns  $\rightarrow$  arrays.

Apply **probeset** to **AffyBatch** object to get a list of **ProbeSet** objects.

- Cel: Single array cel intensity data.
- Cdf: Information contained in a CDF file.

# **CDF** data packages

- Data packages containing necessary CDF information are available at <u>www.bioconductor.org</u>.
- Packages contain environment objects, which provide mappings between AffyIDs and matrices of probe locations,

rows  $\rightarrow$  probe-pairs, columns  $\rightarrow$  PM, MM (e.g., 20X2 matrix for hu6800).

- cdfName slot of AffyBatch.
- HGU95Av2 and HGU133A provided in affy package.

## Reading in data: ReadAffy

🗶 File Browser 🧏				_ 0	-	
Choose CEL files						
	Direc	tory: /tmp				
Files in c	lirectory		Files selec	ted		
affyR/		🛆 ctrid4Tcellam	np.CEL.gz	Z	Ζ	
affy.Rcheck/					Ш	
annaffy/					Ш	
Biobase/						
bioconductor101420	3462/				Ш	
com/						
ctrid4Tcellamp.CEL.g	z				Ш	
ctrid4Tcell.CEL.gz					Ш	
dataset/						
demo/						
Dilution/						
exo1/					Ш	
exo2/					Ш	
exo2_new/						
exo2_orig/		-		-		
fonts/					21	
Up	Select >>	<< Re	move	Clear		
Finish						

Creates object of class AffyBatch

# Accessing PM and MM data

- **probeNames**: method for accessing AffyIDs corresponding to individual probes.
- pm, mm: methods for accessing probelevel PM and MM intensities → probes x arrays matrix.
- Can use on AffyBatch objects.

# **Diagnostic plots**

- See demo(affy).
- Diagnostic plots of probe-level intensities, PM and MM.
  - image: 2D spatial color images of log intensities (AffyBatch, Cel).
  - boxplot: boxplots of log intensities (AffyBatch).
  - mva.pairs: scatter-plots with fitted curves (apply exprs, pm, or mm to AffyBatch object).
  - hist: density plots of log intensities (AffyBatch).

#### image

read from file: HIVControl4A.CEL.gz



read from file: HIVControl4B.CEL.gz

1.0

0.8

0.6

0.4

0.2

0.0

0.0

1.0 0.8 0.6 0.4 0.2 0.0 Т 0.2 0.4 0.6 0.8 1.0 0.0 0.2 0.4 0.6 0.8

read from file: HIVControl4B.CEL.gz

1.0

#### hist



hist(Dilution,col=1:4,type="1",lty=1,lwd=3)



Small part of dilution study



boxplot(Dilution,col=1:4)

### mva.pairs



## **Expression measures**

- **expresso**: Choice of common methods for
  - background correction: bgcorrect.methods
  - normalization: normalize.AffyBatch.methods
  - probe specific corrections: pmcorrect.methods
  - expression measures:
     express.summary.stat.methods.
- rma: Fast implementation of RMA (Irizarry et al., 2003): model-based background correction, quantile normalization, median polish expression measures.
- **express**: Implementing your own method for computing expression measures.
- normalize: Normalization procedures in normalize.AffyBatch.methods Or normalize.methods (object).

### **Expression meassures:** expresso













# Probe sequence analysis

- Examine probe intensity based on location relative to 5' end of RNA sequence of interest.
- Expect probe intensities to be lower at 5' end compared to 3' of mRNA.
- E.g.

deg<-AffyRNAdeg(Dilution)
plotAffyRNAdeg(deg)</pre>