The Bioconductor Project for Reproducible Analysis of High Throughput Genomic Data

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Analysis and Comprehension of High Throughput Genomic Data

Hallmarks of effective computational software

- 1. Extensive: data, annotation
- 2. Statistical: volume, technology, experimental design
- 3. Reproducible: long-term, multi-participant science
- 4. Leading edge: novel, technology-driven
- 5. Accessible: affordable, transparent, usable

1. Extensive Data and Annotation

Data

- Expression, tiling, methylation, custom arrays.
- Sequence analysis, e.g., ChIP-, RNA-seq
- Other high-throughput assays, e.g., flow cytometry, mass spec., imaging
- Public repositories, e.g., GEO, ArrayExpress

Annotation, e.g.,

- Well-curated: NCBI, Biomart, UCSC, MsigDB, GO, KEGG
- ▶ Loosely curated: emerging, specialized, & lab-based
- Consortium: HapMap, 1000 genomes, TCGA

Bioconductor

Goal Help biologists understand their data

Focus Expression and other microarray; flow cytometry

► High-throughput sequencing

Themes Contributions from 'core' members and (primarily academic) user community

▶ Based on the R programming language – statistics, visualization, interoperability

Reproducible – scripts, vignettes, packages

Open source / open development

Success > 400 packages; publications; 8,000 web visits / week; 75,000 unique IP downloads / year; very active mailing list; annual conferences; courses; . . .

Bioconductor: Sample Work Flow

```
> ##
> ## Pre-processing
> library(affy)
> eset <- just.rma()</pre>
> ##
> ## Quality assessment
> library(arrayQualityMetrics)
> arrayQualityMetrics(eset)
> ##
> ## Differential expression
> library(limma)
> status <-
+ c("Trt", "Trt", "Trt", "Ctrl", "Ctrl", "Ctrl")
> design <- model.matrix( ~status )</pre>
> fit <- eBayes(lmFit(eset, design))</pre>
> topTable(fit, coef=2)
```

Technology

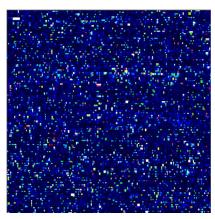
- Acknowledging artifacts and biases
- Accomodate using statistical models, e.g., RMA

Volume of data

Data reduction essential

Experimental design

- Exploratory analysis
- Hypothesis-driven; designed experiments
- Cost-effective, but not too clever



Expression array. Pseudocolors represent hybridisation intensities of RNA to features. Source: url

Technology

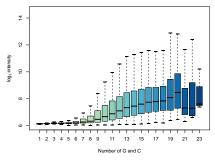
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Measured intensity increases with GC content; Chronic Lymphocytic Leukemia (CLL) dataset.

Technology

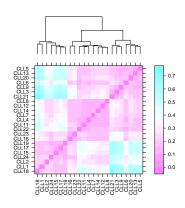
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Heatmap summarizing distance between CLL arrays

Technology

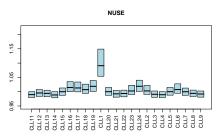
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Normalized unscaled standard error (NUSE) suggests array CLL1 is an outlier.

Technology

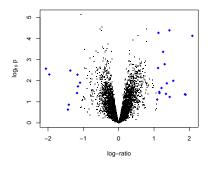
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'Progressive' vs. 'stable' status. log *P* vs. log-fold change, CLL data set. Probe sets with extreme differentiation highlighted.

3. Reproducible Research

Long-term

 Returning to analysis after days, weeks, months of other activity

Multi-participant: communicating with...

- Other statisticians / bioinformaticians
- Biologists and others without specialized statistical knowledge

Science: reproducibility...

- Facilitates third-party verification
- Allows critical assessment
- ► Challenging, even in high-profile journals requiring archived raw data (loannidis *et al.*, 2009, Nat Genet 41: 149-155).

Original research

- Potti et al., 2006; Hsu et al., 2007
- NCI60 cell line drug sensitivity signature
- Clinical trial allocation

Reproducibility

- ► Baggerly & Coombes, 2009
- Off-by-one cisplatin gene signature
- Four 'interesting' genes not supported by analysis (two not on array)

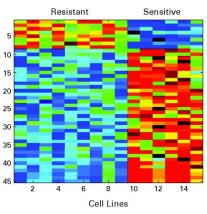
References

- Potti et al. 2006 Nat Med
 12: 1294-1300; (retracted)
- Hsu et al. 2007 J Clin Oncol 25: 4350-4357. (retracted)
- ► Baggerly & Coombes 2009 Ann Appl Stat 3: 1309-1334

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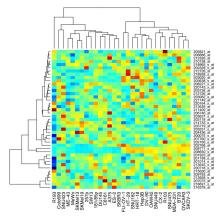


Hsu et al., cisplatin, fig. 1a

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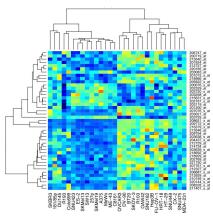


Baggerly & Coombes, fig. 2a

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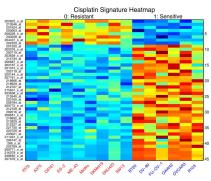


Baggerly & Coombes, fig. 2b

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Baggerly & Coombes, fig. 2d

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... results incorporate several simple errors that may be putting patients at risk. One theme that emerges is that the most common errors are simple (e.g., row or column offsets); conversely, it is our experience that the most simple errors are common – Baggerly & Coombes, 2009

Reproducible Research: Bioconductor

Script-based Data transformations necessarily documented
'Literate programming' Text documents embed scripts, scripts evaluated when text document processed

Versioned software and repositories Record which package versions used, and retrieve from Bioconductor archives

Integrated data containers Sample descriptions and expression data in a single object. Subsetting expression data automatically subsets sample descriptions

The ALL dataset

```
> library(ALL); data(ALL); ALL
ExpressionSet (storageMode: lockedEnvironment)
assayData: 12625 features, 128 samples
  element names: exprs
protocolData: none
phenoData
  sampleNames: 01005 01010 ... LAL4
    (128 total)
  varLabels: cod diagnosis ... date
    last seen (21 total)
  varMetadata: labelDescription
featureData: none
experimentData: use 'experimentData(object)'
  pubMedIds: 14684422 16243790
Annotation: hgu95av2
```

4. Leading Edge

Technological innovations

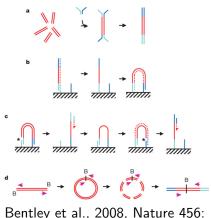
- ► E.g., SNP, miRNA arrays
- ► E.g., lab sequencing platforms; novel protocols

Fast-changing

- Commercial software products not yet developed, or already out-of-date
- Research questions require novel solutions

Sequencing technologies

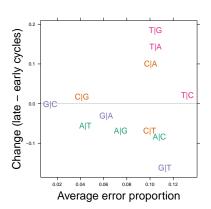
- Historically (e.g., 2 years ago): short reads, low 'tail' quality, tail base call bias, data volume
- Current: count models, read bias, designed experiments, variant representations, annotation



Bentley et al., 2008, Nature 456: 53-9

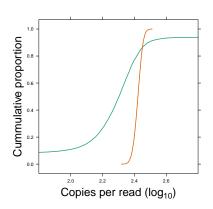
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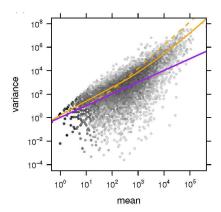
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Poisson (purple) and negative binomial (orange) fit to RNA-seq data. Anders & Huber, 2010, Genome Biol, 11:R106

5. Accessible

Affordable

▶ Purchase / licensing; time

Transparent

- Algorithms, e.g., RMA
- Code reuse

Challenges and solutions

- ▶ Research questions requiring 'one-off' solutions
- Software bugs

Usable

- Documentation
- Training, such as today!

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Documentation

- ► Help pages
- Vignettes
- Archived course and conference material
- Mailing list

BioC2011

- Annual conference user and scientific presentations, workshops, poster session
- Seattle July 27-29

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