# Lab: Introduction to Bioconductor's ExpressionSet Class

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## 1 Introduction

In this lab you will learn how to create and manipulate *ExpressionSet* objects. In the processes you will have an opportunity to practice some basic R skills.

## 2 Loading Packages

The definition of the *ExpressionSet* class along with many methods (OOP-speak for functions) for manipulating *ExpressionSet* objects are defined in the Biobase package. In general, you need to load class and method definitions before you use them. When using Bioconductor, this means loading R packages using library or require.

## > library("Biobase")

Exercise 1

What happens when you try to load a package that is not installed?

## 3 Building an ExpressionSet From Scratch

The data from many high-throughput genomic experiments, such as microarray experiments, can be summarized by a matrix of expression. The matrix has F rows and S columns, where F is the number of features on the chip and S is the number of samples. In addition, one will have a data table that provides information on the samples (e.g., sex, age, and treatment status). The information describing the samples, or *phenotypes*, can be represented as an S by V table, where V is the number of covariates. In R, we use a *data.frame* to hold this "phenoData". Note that the columns of the expression matrix must align with the rows of the phenoData table. The *ExpressionSet* class provides a container for the expression matrix and phenoData and keep the two properly aligned.

In the exercises below, you will learn how to create a new *ExpressionSet* instance given a matrix of expression values and a *data.frame* containing the phenoData. Other labs will cover the creation of the expression matrix from raw CEL files for microarray data.

## 3.1 Loading the Expression Matrix

## 3.1.1 R Binary Files

You can save objects in your R session to a file using the **save** function. By default, this will create a file in R's internal binary format. The same binary file produced by a call to **save** can be used on Linux, OS X, and Windows.

You can load the objects saved in an rda file using the load function.

## 3.1.2 Loading ALLmat

Below, we will load a large *matrix* of expression values stored in the file ALLmat.rda. The example assumes that the file is in the current working directory. You can change the working directory using setwd.

> getwd()

```
[1] "/Users/seth/proj/COURSES/bioc_R_intro/Bioc_intro"
```

> ls()

character(0)

```
> load("ALLmat.rda")
> ls()
```

[1] "ALLmat"

The ls function lists the R objects in your current working environment. You should see a new object named ALLmat appear after the call to load.

### Exercise 2

Open and read the help page for load.

### Exercise 3

Determine the class and dimension of the matrix.

## 3.2 Loading the Phenotype Data

Covariates describing the samples in this experiment have been saved to a whitespacedelimited text file called ALL-sample-info.txt. Delimited text files are common and can be produced from Microsoft Excel by saving as a "csv" file (this stands for comma separated values, but the separator does not have to be a comma).

R's read.table function is a powerful tool for reading delimited text files. Below, you will use it to read in the phenoData.

```
> samples <- read.table("ALL-sample-info.txt", header = TRUE,
+ check.names = FALSE)
```

## Exercise 4

What class does read.table return?

### Exercise 5

Determine the column names of samples. Hint: apropos("name").

### Exercise 6

Use **sapply** to determine the classes of each column of **samples**. Hint: read the help page for **sapply**.

### Exercise 7

Examine the sex and age of the 15th and 30th samples. Do the same for the sample with cod matching 11005.

To make the phenoData more self-documenting, we have a file ALL-varMeta.txt that gives a description for each column of samples. You can use read.table to read this file into an R object.

```
> varInfo <- read.table("ALL-varMeta.txt", header = TRUE,
+ colClasses = "character")
> varInfo[c("sex", "cod", "mol.biol"), , drop = FALSE]
```

|          | labelDescription      |  |  |
|----------|-----------------------|--|--|
| sex      | Gender of the patient |  |  |
| cod      | Patient ID            |  |  |
| mol.biol | molecular biology     |  |  |

Bioconductor's Biobase package provides a class called *AnnotatedDataFrame* that allows you to store the column descriptions with the data. Create an *AnnotatedDataFrame* instance for our phenoData by following the example below.

```
> pd <- new("AnnotatedDataFrame", data = samples, varMetadata = varInfo)</pre>
```

## 3.3 Creating an *ExpressionSet*, finally

Now that you have a *matrix* of expression values (ALLmat) and an *AnnotatedDataFrame* containing the phenotype information (pd), you are ready to put the pieces together and create an *ExpressionSet*.

```
> ALLSet <- new("ExpressionSet", exprs = ALLmat, phenoData = pd,
+ annotation = "hgu95av2")
```

The annotation argument is intended to hold the name of the R package that provides annotation data for the chip used in the experiment. In this case, the appropriate annotation package is hgu95av2.

## 3.4 *ExpressionSet* Basics

Now that you have an *ExpressionSet* instance, let's explore some of the basic operations. You can get an overview of the structure and available methods for *ExpressionSet* objects by reading the help page:

```
> help("ExpressionSet-class")
> "?"(class, ExpressionSet)
```

When you print an *ExpressionSet* object, a brief summary of the contents of the object is displayed. All of the data contained by the *ExpressionSet* is not shown. This would not be useful as it would fill your screen with data.

> ALLSet

```
Instance of ExpressionSet
assayData
 Storage mode: lockedEnvironment
 featureNames: 1000_at, 1001_at, 1002_f_at, ..., AFFX-YEL021w/URA3_at, AFFX-YEL024w/RIF
 Dimensions:
        exprs
Rows
        12625
Samples
          128
phenoData
 sampleNames: 01005, 01010, 03002, ..., 83001, LAL4 (128 total)
 varLabels:
    cod: Patient ID
    diagnosis: Date of diagnosis
         Gender of the patient
    sex:
         Age of the patient at entry
    age:
    BT: does the patient have B-cell or T-cell ALL
    . . . : . . .
    relapse: Relapse? Derived from f.u
    transplant: did the patient receive a bone marrow transplant? Derived from f.u
    f.u: follow up data available
    date last seen: date patient was last seen
    (21 total)
Experiment data
 Experimenter name:
 Laboratory:
 Contact information:
```

Title: URL: PMIDs: No abstract available.

Annotation [1] "hgu95av2"

### 3.4.1 Accessing Data Elements

A number of accessor functions are available to extract data from an *ExpressionSet* instance. You can access the columns of the phenotype data (an *AnnotatedDataFrame* instance) using \$:

```
> ALLSet$sex[1:5] == "F"
```

[1] FALSE FALSE TRUE FALSE FALSE

> ALLSet\$"t(9;22)"[1:5]

[1] TRUE FALSE NA FALSE FALSE

You can retrieve the names of the features using featureNames. For many microarray datasets, the feature names are the probeset identifiers.

```
> featureNames(ALLSet)[1:5]
```

[1] "1000\_at" "1001\_at" "1002\_f\_at" "1003\_s\_at" "1004\_at"

The unique identifiers of the samples in the data set are available via the sampleNames method. The varLabels method lists the column names of the phenotype data:

```
> sampleNames(ALLSet)[1:5]
```

```
[1] "01005" "01010" "03002" "04006" "04007"
```

```
> varLabels(ALLSet)
```

| [1]  | "cod"        | "diagnosis"      | "sex"            |
|------|--------------|------------------|------------------|
| [4]  | "age"        | "BT"             | "remission"      |
| [7]  | "CR"         | "date.cr"        | "t(4;11)"        |
| [10] | "t(9;22)"    | "cyto.normal"    | "citog"          |
| [13] | "mol.biol"   | "fusion protein" | "mdr"            |
| [16] | "kinet"      | "ccr"            | "relapse"        |
| [19] | "transplant" | "f.u"            | "date last seen" |

You can extract the expression *matrix* and the *AnnotatedDataFrame* of sample information using exprs and phenoData, respectively:

> mat <- exprs(ALLSet)
> adf <- phenoData(ALLSet)</pre>

#### 3.4.2 Subsetting

Probably the most useful operation to perform on *ExpressionSet* objects is subsetting. Subsetting an *ExpressionSet* is very similar to subsetting the expression *matrix* that is contained within the *ExpressionSet*, the first argument subsets the features and the second argument subsets the samples. Here are some examples:

A new *ExpressionSet* consisting of the 5 features and the first 3 samples:

```
> vv <- ALLSet[1:5, 1:3]
> dim(vv)
Rows Samples
5 3
> featureNames(vv)
[1] "1000_at" "1001_at" "1002_f_at" "1003_s_at" "1004_at"
> sampleNames(vv)
[1] "01005" "01010" "03002"
```

A subset consisting of only the male samples:

> males <- ALLSet[, ALLSet\$sex == "M"]</pre>

Samples that have B-cell type ALL:

```
> anyB <- grep("^B", ALLSet$BT)
> bcell <- ALLSet[, anyB]</pre>
```

## 4 What was used to create this document

The version number of R and the packages and their versions that were used to generate this document are listed below.

- Version 2.3.1 Patched (2006-06-08 r38315), powerpc-apple-darwin8.6.0
- Base packages: base, datasets, grDevices, graphics, methods, stats, tools, utils
- Other packages: Biobase 1.10.0