Creating annotated output with affycoretools and Reporting Tools

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1 Overview

The *affycoretools* package is intended to help people easily create useful output from various analyses. While affycoretools was originally intended for those using Affymetrix microarrays, this is no longer the case. While some functions remain Affy-centric, most are now much more general, and can be used for any microarray or RNA-Seq experiment.

2 Introduction

This package has evolved from my work as a service core biostatistician. I routinely analyze very similar experiments, and wanted to create a way to minimize all the cutting and pasting of code that I found myself doing. In addition, I wanted to come up with a good way to make an analysis reproducible, in the sense that I (or somebody else) could easily re-create the results.

In the past this package relied on the *annaffy* package, and was intended to be used in concert with a 'Sweave' document that contained both the code that was used to analyze the data, as well as explanatory text that would end up in a pdf (or HTML page) that could be given to a client. In the intervening period, people have developed other, better packages such as *knitr* and *ReportingTools* that make it much easier to create the sort of output I like to present to my clients.

3 Using affycoretools

For this section we will be using the sample. ExpressionSet data set that comes with the *Biobase* package. Remember that you can always run this code at home by doing this:

```
library(knitr)
purl(system.file("doc/RefactoredAffycoretools.Rnw", package="affycoretools"))
```

And then you will have a file called RefactoredAffycoretools.R in your working directory that you can either source or open with RStudio or Emacs/ESS, and run by chunk or line by line.

We first load and rename the data:

```
suppressMessages(library(affycoretools))
data(sample.ExpressionSet)
eset <- sample.ExpressionSet
eset

## ExpressionSet (storageMode: lockedEnvironment)
## assayData: 500 features, 26 samples
## element names: exprs, se.exprs
## protocolData: none
## phenoData
## sampleNames: A B ... Z (26 total)
## varLabels: sex type score
## varMetadata: labelDescription
## featureData: none</pre>
```

```
## experimentData: use 'experimentData(object)'
## Annotation: hgu95av2
```

This ExpressionSet object is a truncated data set, based on an Affymetrix HG-U95av2 array. There are 26 samples and 500 probesets. We will use the phenoData to fit a linear model using limma. comment: We will not cover any aspects of fitting a linear model here; the limma User's Guide covers this topic in depth. In addition, this analysis isn't meant to be correct in any sense; we are just doing this to get some data to annotate and output.

```
suppressMessages(library(limma))
pd <- pData(phenoData(eset))</pre>
design <- model.matrix(~0+type+sex, pd)</pre>
colnames(design) <- gsub("type|sex", "", colnames(design))</pre>
contrast \leftarrow matrix(c(1,-1,0))
colnames(contrast) <- "Case vs control"</pre>
fit <- lmFit(eset, design)</pre>
fit2 <- contrasts.fit(fit, contrast)</pre>
fit2 <- eBayes(fit)</pre>
topTable(fit2, 1)[,1:4]
##
                               logFC AveExpr
                                                   t
                                                               P. Value
## 31667_r_at
                            763.1646 735.1835 24.84768 1.429233e-18
## AFFX-HSAC07_X00351_M_st 193.4776 197.6336 23.57778 4.770893e-18
## 31375_at
                           368.1255 395.6733 22.26916 1.761763e-17
## 31466_at
                           198.3558 195.4360 20.62148 1.011904e-16
## 31597_r_at
                          1590.9543 1634.2481 20.58837 1.049424e-16
## 31440_at
                           597.5251 659.0559 20.51108 1.142742e-16
                           2344.2449 2504.3938 20.29175 1.457511e-16
## 31396_r_at
                        8005.2783 8681.4000 19.45532 3.768484e-16
## AFFX-hum_alu_at
## 31391_at
                            380.7353 401.1064 19.35000 4.258204e-16
## AFFX-HSAC07_X00351_3_at 4771.0705 4869.6635 19.34299 4.293033e-16
```

At this point we can generate a data.frame, but this data.frame has no annotation, such as gene names or symbols, etc, that say what each probeset is measuring. The *MArrayLM* object that we are calling 'fit2', is capable of containing these data, and will append those data to the topTable output.

```
suppressMessages(library(hgu95av2.db))
gns <- select(hgu95av2.db, featureNames(eset), c("ENTREZID", "SYMBOL", "GENENAME"))</pre>
## There are one-to many mappings here, so we just
## removed duplicates in a very naive way.
gns <- gns[!duplicated(gns[,1]),]</pre>
fit2$genes <- gns
topTable(fit2, 1)[,1:3]
##
                                           PROBEID ENTREZID
                                                              SYMBOL
## 31667_r_at
                                        31667_r_at 10002
                                                               NR2E3
## AFFX-HSAC07_X00351_M_st AFFX-HSAC07_X00351_M_st
                                                      <NA>
                                                                 <NA>
## 31375_at
                                          31375_at
                                                      <NA>
                                                                <NA>
## 31466_at
                                          31466_at
                                                       3128 HLA-DRB6
## 31597_r_at
                                        31597_r_at 1978 EIF4EBP1
```

```
## 31440_at
                                          31440_at
                                                       6932
                                                                TCF7
## 31396_r_at
                                        31396_r_at
                                                       4440
                                                                MSI1
## AFFX-hum_alu_at
                                   AFFX-hum_alu_at
                                                       <NA>
                                                                <NA>
## 31391_at
                                          31391_at
                                                       9001
                                                                HAP1
## AFFX-HSAC07_X00351_3_at AFFX-HSAC07_X00351_3_at
                                                                <NA>
                                                       <NA>
```

After adding the annotation data to the MArrayLM object, the topTable output now contains the appropriate annotation data for each probeset. At this point we can output an HTML table that contains these data.

```
suppressMessages(library(ReportingTools))
htab <- HTMLReport("afile", "My cool results")
publish(topTable(fit2, 1), htab)
finish(htab)
## [1] "./afile.html"</pre>
```

And now we have a HTML table called 'afile.html' in our working directory, that contains the data for our top 10 genes. This table is not particularly interesting, and the *ReportingTools* package already has functionality to just do something like

```
htab <- HTMLReport("afile2", "My cool results, ReportingTools style")
publish(fit2, htab, eset, factor = pd$type, coef = 1, n = 10)
finish(htab)
## [1] "./afile2.html"</pre>
```

and it will automatically generate an annotated table, with some extra plots that show the different groups, and we didn't even have to use topTable directly. However, the default plots in the HTML table are a combination of dotplot and boxplot, which I find weird (see afile2.html if you are running this code yourself). Since *ReportingTools* is easily extensible, we can make changes that are more pleasing.

Note that there are two differences in the way we did things. First, we create a data.frame, and then decorate it with the plots using the makeImages function. This will by default create dotplots (or you can specify boxplots). For the plots to fit in an HTML table, there are no axis labels. However, each plot is also a link, and if you click on it, a larger plot with axis labels will be presented. See 'afile3.html', if you are running this code yourself.

All the little files that get created can get pretty messy, so the default is to put everything into a 'reports' subdirectory, so your working directory stays clean. For this example we over-ride the defaults so we do not have to go searching in subdirectories for our tables.

An alternative parameterization that probably makes more sense is to fit coefficients for each sex/treatment combination.

With this parameterization we can look at intra-sex differences, as well as the interaction (looking for sex-specific changes). This now means that we have a total of three HTML tables to output, which makes things a bit more complex to present. Luckily, this is pretty simple to accomplish. For this step we will now use the default 'reports' subdirectory to keep everything straight. In addition, we will trim down the output a bit.

```
## get a list containing the output for each comparison
out <- lapply(1:3, function(x) topTable(fit2, x))</pre>
## process the output to add images
htab <- lapply(1:3, function(x){</pre>
    tmp <- HTMLReport(gsub("_", " ", colnames(contrast)[x]), colnames(contrast)[x], "./reports")</pre>
    tmp2 <- makeImages(out[[x]], eset, grps, design, contrast, x)</pre>
    publish(tmp2$df, tmp, .modifyDF = list(affyLinks, entrezLinks))
    finish(tmp)
    return(tmp)
})
## Now make an index.html file to contain links to the various comps
idx <- HTMLReport("index", "Links to our stuff")</pre>
publish(hwriter::hwrite("Univariate comparisons", br = TRUE, header = 2), idx)
publish(Link(htab), idx)
finish(idx)
## [1] "./index.html"
```

Now there will be an index.html file in the current directory that has individual links to each of the three comparisons we made. This is nice, as you only have to point a client or PI to a single link that they can use to explore all the results.

We are often asked to create a Venn diagram showing overlap between groups. This is pretty simple to do, but it would be nicer to have an HTML version with clickable links, so your PI or end user can see what genes are in each cell of the Venn diagram. As an example, we can generate a Venn diagram comparing overlapping genes between the male and female comparisons.

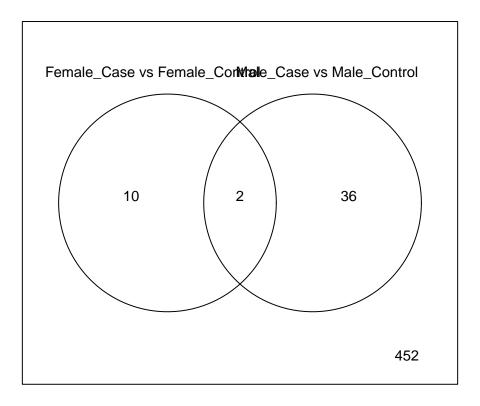


Figure 1: Venn diagram

```
collist <- list(1:2)
venn <- makeVenn(fit2, contrast, design, collist = collist, adj.meth = "none")
vennlnk <- vennPage(venn, "venn_diagram", "Venn diagram")</pre>
```

The makeVenn function also returns a vennCounts object that we can use in our *knitr* document to generate a Venn diagram there as well (1).

```
vennDiagram(venn[[1]]$venncounts, cex = 0.9)
```

And we can add a link to our index page quite easily.

```
idx <- HTMLReport("index","Links to our stuff")
publish(hwriter::hwrite("Univariate comparisons", br = TRUE, header = 2), idx)
publish(Link(htab), idx)
publish(hwriter::hwrite("Venn diagrams", br = TRUE, header = 2), idx)
publish(Link("Venn1", vennlnk), idx)
finish(idx)
## [1] "./index.html"</pre>
```

There is similar functionality for presenting the results of a GO hypergeometric analysis (makeGoTable), and GSEA analysis, based on the romer function in *limma* (runRomer and outputRomer).

4 Session information

The version of R and packages loaded when creating this vignette were:

toLatex(sessionInfo())

- R version 3.2.1 beta (2015-06-08 r68489), x86_64-unknown-linux-gnu
- Locale: LC_CTYPE=en_US.UTF-8, LC_NUMERIC=C, LC_TIME=en_US.UTF-8, LC_COLLATE=C, LC_MONETARY=en_US.UTF-8, LC_MESSAGES=en_US.UTF-8, LC_PAPER=en_US.UTF-8, LC_NAME=C, LC_ADDRESS=C, LC_TELEPHONE=C, LC_MEASUREMENT=en_US.UTF-8, LC_IDENTIFICATION=C
- Base packages: base, datasets, grDevices, graphics, methods, parallel, stats, stats4, utils
- Other packages: AnnotationDbi 1.30.1, Biobase 2.28.0, BiocGenerics 0.14.0, DBI 0.3.1,
 GenomeInfoDb 1.4.0, IRanges 2.2.4, RSQLite 1.0.0, ReportingTools 2.8.0, S4Vectors 0.6.0, affy 1.46.1,
 affycoretools 1.40.5, hgu95av2.db 3.1.3, knitr 1.10.5, limma 3.24.10, org.Hs.eg.db 3.1.2
- Loaded via a namespace (and not attached): AnnotationForge 1.10.1, BSgenome 1.36.0, BiocInstaller 1.18.3, BiocParallel 1.2.3, BiocStyle 1.6.0, Biostrings 2.36.1, Category 2.34.2, DESeq2 1.8.1, Formula 1.2-1, GGally 0.5.0, GO.db 3.1.2, GOstats 2.34.0, GSEABase 1.30.2, GenomicAlignments 1.4.1, GenomicFeatures 1.20.1, GenomicRanges 1.20.5, Hmisc 3.16-0, KernSmooth 2.23-14, MASS 7.3-40, Matrix 1.2-1, OrganismDbi 1.10.0, PFAM.db 3.1.2, R.methodsS3 1.7.0, R.oo 1.19.0, R.utils 2.1.0, RBGL 1.44.0, RColorBrewer 1.1-2, RCurl 1.95-4.6, Rcpp 0.11.6, RcppArmadillo 0.5.200.1.0, Rsamtools 1.20.4, VariantAnnotation 1.14.2, XML 3.98-1.2, XVector 0.8.0, acepack 1.3-3.3, affyio 1.36.0, annotate 1.46.0, biomaRt 2.24.0, biovizBase 1.16.0, bit 1.1-12, bitops 1.0-6, caTools 1.17.1, cluster 2.0.1, codetools 0.2-11, colorspace 1.2-6, dichromat 2.0-0, digest 0.6.8, edgeR 3.10.2, evaluate 0.7, ff 2.2-13, foreach 1.4.2, foreign 0.8-63, formatR 1.2, futile.logger 1.4.1, futile.options 1.0.0, gcrma 2.40.0, gdata 2.16.1, genefilter 1.50.0, geneplotter 1.46.0, ggbio 1.16.0, ggplot2 1.0.1, gplots 2.17.0, graph 1.46.0, grid 3.2.1, gridExtra 0.9.1, gtable 0.1.2, gtools 3.5.0, highr 0.5, hwriter 1.3.2, iterators 1.0.7, lambda.r 1.1.7, lattice 0.20-31, latticeExtra 0.6-26, locfit 1.5-9.1, magrittr 1.5, munsell 0.4.2, nnet 7.3-9, oligoClasses 1.30.0, plyr 1.8.2, preprocessCore 1.30.0, proto 0.3-10, reshape 0.8.5, reshape 2 1.4.1, rpart 4.1-9, rtracklayer 1.28.4, scales 0.2.4, splines 3.2.1, stringi 0.4-1, stringr 1.0.0, survival 2.38-1, tools 3.2.1, xtable 1.7-4, zlibbioc 1.14.0