Primer: Preparing NChannelSet objects with differential expression scores

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This document exemplifies the processing of differential expression data using small, simulated datasets shipped with the gCMAP package. To see real-life examples with data from available from public databases, please refer to the documentation of the gCMAPWeb companion package.

1 Differential expression analysis

The gCMAP package offers a the generate_gCMAP_NChannelSet function to process multiple instances differential expression experiments with two classes (e.g. cases vs controls). For microarray data, the limma package is used to calculate a moderated t-statistic (default). Optionally, a standard t-test can be computed instead. For RNAseq data, the DESeq package is used instead.

Data preprocessing differs considerably between different technologies and array platforms and needs to be performed beforehand. Normalized microarray data and accompanying annotation is passed to generate_gCMAP_NChannelSet as a list of ExpressionSet objects, RNAseq data can be passed as a list of CountDataSet objects instead.

To generate a set of 3 example CountDataSets , we use the makeExampleCountDataSet function from the DESeq package.

```
> library(gCMAP)
> library(DESeq)
> set.seed( 123 )
> cds.list <- lapply( 1:3, function(n) {</pre>
     cds <- makeExampleCountDataSet()</pre>
     featureNames(cds) <- paste("gene",1:10000, sep="_")</pre>
+ })
> names(cds.list) <- paste("Instance", 1:3, sep="")</pre>
> sapply(cds.list, dim)
          Instance1 Instance2 Instance3
Features
              10000
                         10000
                                    10000
Samples
                             5
> sapply(cds.list, function(n) pData(n)$condition )
     Instance1 Instance2 Instance3
[1,] "A"
                "A"
                           "A"
                           "A"
[2,] "A"
                "A"
[3.] "B"
                "B"
                           "B"
                "B"
                           "B"
[4,] "B"
[5,] "B"
                "B"
                           "B"
```

By default, each CountDataset object contains counts for 10000 genes from five samples. Each sample is assigned to one of two conditions, A or B, in the phenoData slot of the CountDataSet . The pData column containing group membership information (e.g. "condition") is provided as the control_perturb_col parameter. The levels associated with control and treatment groups are specified as "control" and "perturb" character strings.

Each of the three CountDataSet instances is analyzed individually by generate_gCMAP_NChannelSet. To assemble the results into a single NChannelSet, the input ExpressionSet or CountDataSet objects must contain measurements for the same features (e.g. the vectors returned by "featureNames" must be identical across all instances).

To include information about the instances in the NChannelSet, a 'sample.annotation' data.frame can be provided, containing exactly one row for each element of the input list of ExpressionSet / CountDataSet objects.

For array data, a NChannelSet with slots "exprs", "z", "p", and "log_fc" is returned, containing the average intenstity across all samples within the instance, z-scores, (raw) p-values and log2 fold changes, respectively. If count data is processed, an additional "mod_fc" channel is returned, providing the moderated fold change, calculated after performing variance-stabilising transformation across all input instances. (Please consult the DESeq vignette for details.)

1.1 Storing assayData as BigMatrix objects on disk

When large numbers of instances are processed, the resulting NChannelSet objects can require large amounts of memory. The bigmemory and bigmemoryExtras packages can be used to create BigMatrix objects, allowing methods to subset large datasets without having to load them fully into memory first.

Note: at the time of writing, the bigmemory package was only available for Unix and Mac OS X operating systems but not for Windows. Windows users can take advantage of gCMAP's functionality but datasets must be fully loaded into memory first.

If the bigmemory and bigmemoryExtras packages are available and a file name is provided via the "big.matrix" parameter, generate_gCMAP_NChannelSet uses the BigMatrix package to store data from each channel on disk. In the future, individual channels and / or subsets of the datasets can then be loaded without requiring the full object to be read into memory again.

To highlight this functionality, we derive three (arbitrary) instances from the sample. Expression Set object available from the Biobase package, process them and store the results in a temporary directory. Note: this section will only created the expected big.matrix files on disk if the bigmemory and bigmemory Extras packages can be loaded.

```
> ## list of ExpressionSets
> if (require(bigmemory)) {
+ data("sample.ExpressionSet") ## from Biobase
```

```
es.list <- list( sample.ExpressionSet[,1:4],
                     sample.ExpressionSet[,5:8],
+
                     sample.ExpressionSet[,9:12])
+
    ## three instances
    names(es.list) <- paste( "Instance", 1:3, sep=".")</pre>
    storage.file <- tempfile()</pre>
    storage.file ## filename prefix for BigMatrices
    de <- generate_gCMAP_NChannelSet(</pre>
      es.list,
      1:3,
+
     platform.annotation = annotation(es.list[[1]]),
      control_perturb_col="type",
      control="Control",
      perturb="Case",
      big.matrix=storage.file)
    channelNames(de)
   head( assayDataElement(de, "z") )
    dir(dirname( storage.file ))
[1] "file438c4ec62f9d.rdata"
                                    "file438c4ec62f9d_exprs"
[3] "file438c4ec62f9d_exprs.desc" "file438c4ec62f9d_log_fc"
[5] "file438c4ec62f9d_log_fc.desc" "file438c4ec62f9d_p"
[7] "file438c4ec62f9d_p.desc"
                                    "file438c4ec62f9d_z"
[9] "file438c4ec62f9d_z.desc"
```

If the bigmemoryExtras package is available, it generated a BigMatrix objects containing pointers to three files in the temporary directory, one for each channel (identified by their suffices). If the package is unavailable, a standard eSet is saved to disk, which will be read fully into memory upon reload.

To demonstrate the use of disk-based NChannelSet objects, we will first delete the object from the current R workspace and reload it from disk.

Accessing the complete matrix in the assayData slots, e.g. for the "z" channel, returns another BigMatrix object with assayData slot pointing to the associated file on disk. Upon subsetting, only the requested part of the dataset is loaded into memory.

The memorize function reads the complete NChannelSet into memory. In addition, one or more selected channels can be specified with the 'name' parameter.

```
> if (require(bigmemory)) {
    ## read z-score channel into memory
    dem <- memorize( de, name="z" )</pre>
    channelNames(dem)
    class( assayDataElement(dem, "z") ) ## matrix
    ## remove tempfiles
    unlink(paste(storage.file,"*", sep=""))
    sessionInfo()
R version 3.5.0 (2018-04-23)
Platform: x86_64-pc-linux-gnu (64-bit)
Running under: Ubuntu 16.04.4 LTS
Matrix products: default
BLAS: /home/biocbuild/bbs-3.7-bioc/R/lib/libRblas.so
LAPACK: /home/biocbuild/bbs-3.7-bioc/R/lib/libRlapack.so
locale:
 [1] LC_CTYPE=en_US.UTF-8
                                LC_NUMERIC=C
 [3] LC_TIME=en_US.UTF-8
                                LC_COLLATE=C
 [5] LC_MONETARY=en_US.UTF-8
                                LC_MESSAGES=en_US.UTF-8
 [7] LC_PAPER=en_US.UTF-8
                                LC_NAME=C
 [9] LC_ADDRESS=C
                                LC_TELEPHONE=C
[11] LC_MEASUREMENT=en_US.UTF-8 LC_IDENTIFICATION=C
attached base packages:
[1] stats4
              parallel stats
                                  graphics grDevices utils
                                                                 datasets
[8] methods
              base
other attached packages:
 [1] bigmemory_4.5.33
                          DESeq_1.32.0
                                                lattice_0.20-35
 [4] locfit_1.5-9.1
                          gCMAP_1.24.0
                                                limma_3.36.0
 [7] GSEABase_1.42.0
                                                annotate_1.58.0
                          graph_1.58.0
[10] XML_3.98-1.11
                          AnnotationDbi_1.42.0 IRanges_2.14.0
[13] S4Vectors_0.18.0
                                                BiocGenerics_0.26.0
                          Biobase_2.40.0
```

[1]	Rcpp_0.12.16	compiler_3.5.0	RColorBrewer_1.1-2
[4]	bigmemoryExtras_1.28.0	bitops_1.0-6	tools_3.5.0
[7]	digest_0.6.15	bit_1.1-12	RSQLite_2.1.0
[10]	memoise_1.1.0	Matrix_1.2-14	DBI_0.8
[13]	Category_2.46.0	GSEAlm_1.40.0	genefilter_1.62.0
[16]	bit64_0.9-7	grid_3.5.0	bigmemory.sri_0.1.3
[19]	RBGL_1.56.0	survival_2.42-3	geneplotter_1.58.0
[22]	blob_1.1.1	splines_3.5.0	xtable_1.8-2
[25]	RCurl_1.95-4.10		