## The pepStat user guide

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A step-by-step guide in the analysis of peptide microarray antibody binding

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#### 1 Introduction

The pepStat package offers a complete analytical framework for the analysis of peptide microarray data. It includes a novel normalization method to remove non-specific peptide binding activity of antibodies, a data smoothing reducing step to reduce background noise, and subject-specific positivity calls.

#### 1.1 Installing the package

The pepStat package requires GSL, an open source scientific computing library. This library is freely available at http://www.gnu.org/software/gsl/.

#### 1.2 Loading the package

As with any R package, it should first be loaded in the session.

```
library(pepStat)
## Loading required package:
                              Biobase
## Loading required package:
                              BiocGenerics
## Loading required package:
                              parallel
##
## Attaching package:
                      'BiocGenerics'
##
## The following objects are masked from 'package:parallel':
##
##
      clusterApply, clusterApplyLB, clusterCall, clusterEvalQ,
      clusterExport, clusterMap, parApply, parCapply, parLapply,
##
##
      parLapplyLB, parRapply, parSapply, parSapplyLB
##
## The following object is masked from 'package:stats':
##
##
      xtabs
##
## The following objects are masked from 'package:base':
##
##
      anyDuplicated, append, as.data.frame, as.vector, cbind,
      colnames, duplicated, eval, Filter, Find, get, intersect,
##
##
      lapply, Map, mapply, match, mget, order, paste, pmax,
##
      pmax.int, pmin, pmin.int, Position, rank, rbind, Reduce,
##
      rep.int, rownames, sapply, setdiff, sort, table, tapply,
##
      union, unique, unlist
##
## Welcome to Bioconductor
##
##
      Vignettes contain introductory material; view with
```

```
'browseVignettes()'. To cite Bioconductor, see
##
##
      'citation("Biobase")', and for packages 'citation("pkgname")'.
##
##
## Attaching package:
                       'Biobase'
##
  The following object is masked from 'package:base':
##
##
##
      anyMissing
##
## Loading required package:
                               IRanges
```

### 2 Generating a peptideSet

### 2.1 Reading in .gpr files

The reading function takes a path as its argument and parses all the .gpr files in the given directory. Alternatively, one may specify a character vector of paths to individual .gpr files.

Optionally, one may provide a path to a mapping file giving annotation data for each slide, such as treatment status or patient information. If provided, the data set **must** be a .csv file and **must** include columns labeled **filename**, **ptid**, and **visit**. Elements in column **filename** must correspond to the filenames of slides to be read in, without the .gpr extension. Column **ptid** is a subject or slide identifier. Column **visit** indicates a case or control condition, such as pre/post vaccination, pre/post infection, or healthy/infected status. Control conditions must be labelled *pre*, while case conditions must be labelled *post*. Alternatively, one may input a **data.frame** satisfying the same requirements.

By default channels F635 Median and B635 Median are collected, and the 'normexp' method of the backgroundCorrect function in the limma package corrects probe intensities for background fluorescence. Other methods may be selected, see documentation.

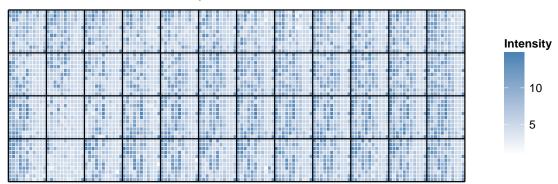
```
mapFile <- system.file("extdata/mapping.csv", package = "pepStat")</pre>
map <- read.csv(mapFile)</pre>
map
##
     filename ptid visit treatment
         f1_1
## 1
                  1
                      Pre
                             PLACEBO
         f1_2
## 2
                  1
                     Post
                             PLACEBO
         f2 1
## 3
                  2
                      Pre
                             PLACEBO
         f2 2
                  2 Post
## 4
                             PLACEBO
## 5
         f3_1
                  3
                     Pre
                             VACCINE
## 6
         f3_2
                  3 Post
                             VACCINE
         f4_1
                  4
## 7
                     Pre
                             VACCINE
         f4_2
                  4
                             VACCINE
## 8
                    Post
```

#### 2.2 Visualize slides

We include two plotting functions to detect possible spatial slide artifacts.

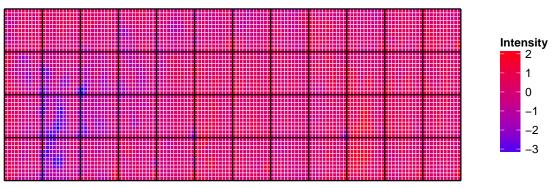
```
plotArrayImage(pSet, array.index = 1)
```





```
plotArrayResiduals(pSet, array.index = 1, smooth = TRUE)
```

### Smoothed Residuals for Sample Name f1\_1



#### 2.3 Accessing peptideSet elements

makePeptideSet returns an object of class peptideSet, the base structure used in pepStat. It contains the sequence and ID of the peptides, measured intensities, annotations added through a mapping file, and probe slide position information. Various accessor functions can extract these values.

```
# peptide intensities
exprs(pSet)[1:5, 1:4]
##
      f1 1
             f1_2
                    f2_1
                           f2 2
## 1 11.89 10.356 11.770 11.444
## 2 10.61 9.485 10.665 10.872
## 3 11.25 10.356 11.366 11.639
## 4 11.10 10.007 11.270 11.278
## 5 10.79 9.033 9.658 9.291
# probe information
head(values(ranges(pSet))[[1]])
## DataFrame with 6 rows and 5 columns
##
           featureID
                             peptide
                                        block
                                                          column
                                                    row
##
         <character>
                         <character> <factor> <factor> <factor>
## 1 WVTVYYGVPVWKDAE WVTVYYGVPVWKDAE
                                             1
                                                      1
                                                               1
## 2 VYYGVPVWKEATTTL VYYGVPVWKEATTTL
                                             1
                                                      1
                                                               2
## 3 VYYGVPVWRDAETTL VYYGVPVWRDAETTL
                                             1
                                                      1
                                                               3
## 4 GVPVWRDADTTLFCA GVPVWRDADTTLFCA
                                             1
                                                      1
                                                               4
## 5 VWKEAKTTLFCASDA VWKEAKTTLFCASDA
                                             1
                                                               5
## 6 DAETTLFCASDAKAY DAETTLFCASDAKAY
# same as 'peptide' and 'featureID' columns above
head(peptide(pSet), 4)
## [1] "WVTVYYGVPVWKDAE" "VYYGVPVWKEATTTL" "VYYGVPVWRDAETTL" "GVPVWRDADTTLFCA"
head(featureID(pSet), 4)
## [1] "WVTVYYGVPVWKDAE" "VYYGVPVWKEATTTL" "VYYGVPVWRDAETTL" "GVPVWRDADTTLFCA"
# mapping file slide annotations
head(pData(pSet))
##
        filename ptid visit treatment
## f1_1
            f1_1
                    1
                        Pre
                              PLACEBO
## f1_2
            f1_2
                    1 Post
                              PLACEBO
## f2_1
            f2_1
                    2
                       Pre
                              PLACEBO
## f2_2
            f2_2
                    2 Post
                              PLACEBO
            f3_1
                    3
## f3_1
                      Pre
                              VACCINE
## f3_2
            f3_2
                    3 Post
                              VACCINE
```

preproc(pSet) stores additional information such as slide layout, background correction methods, normalization, transformation, etc .

#### 2.4 Summarizing within-slide replicates

The function summarizePeptides summarizes within-slide replicates by either their mean or median. Additional peptide sequence and/or annotation information may be incorporated with a RangedData object from the IRanges package. In this example, we use pep\_hxb2 available in the PEP.db package.

```
library(PEP.db)
data(pep_hxb2)
psSet <- summarizePeptides(pSet, summary = "mean", position = pep_hxb2)
## Some peptides have no match in the RangedData object rownames and are removed from the peptideSet!</pre>
```

pep\_hxb2 gives information regarding the position of each peptide, their z-scores, the clades they belong to and the alignment with the reference sequence HXB2.

### 3 Normalizing the peptideSet

The primary goal of the data normalization step is to remove non-biological source of bias and increase the comparability of true positive signal intensities across slides. The method developped for this package uses physiochemical properties of individual peptides to model non-specific antibody binding to arrays.

```
pnSet <- normalizeArray(psSet)</pre>
```

An object of class peptideSet containing the corrected peptides intensities is returned.

## 4 Data smoothing

The optional data smoothing step takes advantage of the overlapping nature of the peptides on the array to remove background noise caused by experimental variation. It is likely that two overlapping peptides will share common binding signal, when present. pepStat use a sliding mean technique technique to borrow strength across neighboring peptides and to reduce signal variability. This statistic increases detection of binding *hotspots* that noisy signals might otherwise obscure. Peptides are smoothed according to their sequence alignment position, taken from position(psSet).

```
psmSet <- slidingMean(pnSet, width = 9)</pre>
```

### 5 Making calls

The final step is to make the positivity calls. The function makeCalls automatically uses information provided in the mapping file, accessed via pData(pSet). It detects whether samples are paired or not. If samples are paired, POST intensities are subtracted from PRE intensities, then thresholded. Otherwise, PRE samples are averaged, and then subtracted from POST intensities. These corrected POST intensities are thresholded.

The freq argument controls whether we return the percentage of responders against each peptide, or a matrix of subject specific call. When freq is TRUE, we may supply a group variable from pData(psmSet) on which we split the frequency calculation.

```
V_calls <- makeCalls(psmSet, freq = TRUE, group = "treatment", cutoff = 0.1,
    method = "FDR", verbose = TRUE)
## You have paired PRE/POST samples</pre>
```

```
sessionInfo()
## R Under development (unstable) (2013-04-23 r62650)
## Platform: x86_64-unknown-linux-gnu (64-bit)
##
## locale:
                                    LC_NUMERIC=C
   [1] LC_CTYPE=en_US.UTF-8
##
   [3] LC_TIME=en_US.UTF-8
                                    LC_COLLATE=en_US.UTF-8
##
   [5] LC_MONETARY=en_US.UTF-8
                                    LC_MESSAGES=en_US.UTF-8
   [7] LC_PAPER=C
##
                                    LC_NAME=C
##
   [9] LC_ADDRESS=C
                                    LC_TELEPHONE=C
  [11] LC_MEASUREMENT=en_US.UTF-8 LC_IDENTIFICATION=C
##
## attached base packages:
## [1] parallel stats
                           graphics grDevices utils
                                                          datasets methods
## [8] base
##
## other attached packages:
## [1] PEP.db_0.99.5
                            pepStat_0.99.6
                                                  IRanges_1.19.24
## [4] Biobase_2.21.6
                            BiocGenerics_0.7.4
                                                  knitr_1.4.1
## [7] BiocInstaller_1.11.4
##
## loaded via a namespace (and not attached):
##
   [1] colorspace_1.2-2
                           dichromat_2.0-0
                                               digest_0.6.3
   [4] evaluate_0.4.7
##
                           fields_6.8
                                               formatR_0.9
   [7] ggplot2_0.9.3.1.99 grid_3.1.0
                                               gtable_0.1.2
## [10] highr_0.2.1
                           labeling_0.2
                                               limma_3.17.21
## [13] MASS_7.3-28
                           munsell_0.4.2
                                               plyr_1.8
## [16] proto_0.3-10
                           RColorBrewer_1.0-5 RCurl_1.95-4.1
```

## [19] reshape2\_1.2.2 scales\_0.2.3 spam\_0.30-1 ## [22] stats4\_3.1.0 stringr\_0.6.2 tools\_3.1.0