Factor analysis with Adaptive Shrinkage using flashr

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Abstract

The *R* package *flashr* provides tools to perform factor analysis with adaptive shrinkage on the factor loadings and the factors and also provides means to visualization of the factor analysis results. The adaptive shrinkage is performed using the **ashr** package due to Stephens (2016).

The package provides generic functions to visualize loadings data and post processing functions to analyze the factors estimated with focus on sparsity and the proportion of variance in the data explained by each factor. It also provides a list of features that play the key role in distinguishing the factors.

flashr version: 0.1.1 ¹

¹This document used the vignette from Bioconductor package DESeq2, CountClust as knitr template

Contents

1 Introduction

FLASH (Factor Loadings with Adaptive Shrinkage) is an extension of the adaptive shrinkage methods in **ashr** package due to Stephens (2016) to the domain of factor analysis. An important consideration in any factor analysis scheme are shrinkage and sparsity. There are many algorithms that perform Sparse Factor Analysis (check Engelhardt and Stephens), however determining the level of shrinkage is a challenging task for the user. FLASH solves this problem by adaptively selecting the level of shrinkage for factor loadings and factors.

flashr offers 3 versions of FLASH (normal, greedy and backfitting) to perform factor analysis with adaptive shrinkage. Also, it provides generic visualization tools to view and analyze the factor loadings along with post processing tools to check the proportion of variance explained and sparsity level of the different factors. Finally, it offers functions to select a list of features that drive the factors or play the most key role in distinguishing the features.

2 flashr Installation

flashr requires the packages ashr, ggplot2, irlba, CountClust, cowplot, RColorBrewer, grid, gridExtra. Also one needs the packages devtools install the developmental version of the flashr which will be the most up-to-date version.

We install the Github version of the package.

```
devtools::install_github("kkdey/flashr")
library(flashr)
```

3 Data preparation

We install a single cell RNA-seq data across mouse embryo developmental stages due to Deng *et al* 2014 [?]. We load it as an ExpressionSet object.singleCellRNASeqMouseDeng2014 data package due to Deng *et al* is a processed version of the data publicly available at Gene Expression Omnibus (GEO:GSE45719: see http://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE45719).

```
library(devtools)

read.data1 = function() {
    x = tempfile()
    download.file('https://cdn.rawgit.com/kkdey/singleCellRNASeqMouseDeng2014/master/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/page/data/Dengangerickets/data/Dengangerickets/page
```

```
z = get(load((x)))
return(z)
}

Deng2014MouseESC <- read.data1()

## Alternatively
# install_github('kkdey/singleCellRNASeqMouseDeng2014')</pre>
```

We load the data

```
deng.counts <- Biobase::exprs(Deng2014MouseESC)
deng.meta_data <- Biobase::pData(Deng2014MouseESC)
deng.gene_names <- rownames(deng.counts)

voom_out <- limma::voom(deng.counts);
voom_weights <- t(voom_out$weights);
voom_data <- t(voom_out$E);</pre>
```

We apply FLASH on the data. We show here example of two versions of FLASH - the normal pooled FLASH on successive residuals and the greedy FLASH. We assume the number of factors to be K=10.

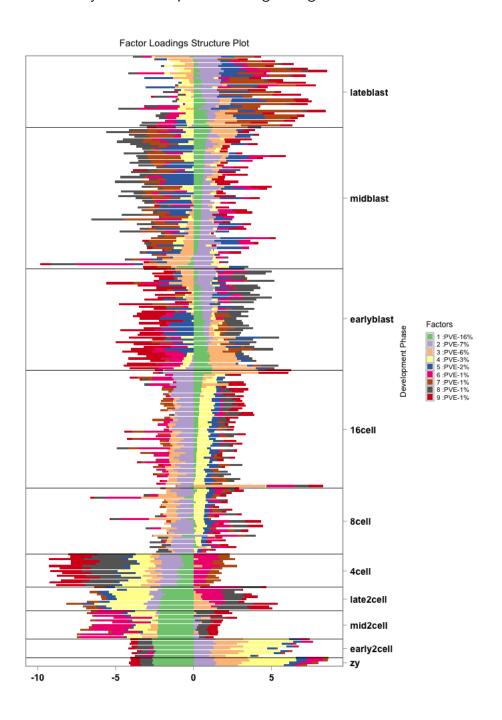
We load the FLASH output from the greedy implementation of FLASH and then we process the output.

```
library(flashr)
ll_deng <- get(data("flash_deng_ex"))</pre>
```

We postprocess the loadings and the factors from the FLASH output. We calculate the percentage of variance explained by each factor.

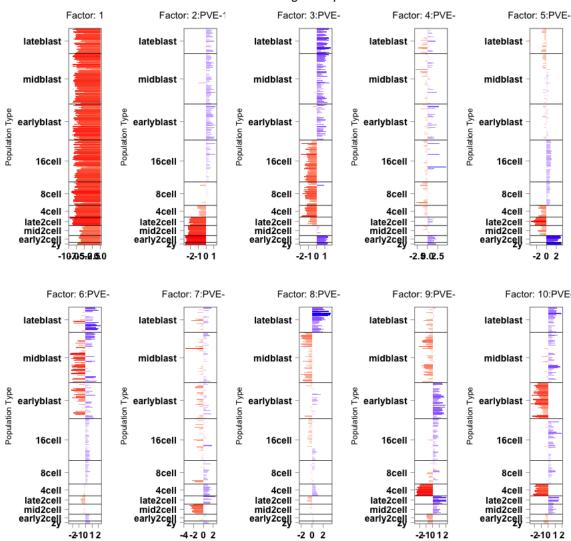
```
postprocess_ll <- flash_factor_postprocess(ll_deng$1,ll_deng$f, voom_data)
pve_percentage <- c("", paste0(":PVE-", round(postprocess_ll$PVE*100,0), "%"))</pre>
```

We represent the loadings by a stacked barchart of the loadings obtained from the FLASH output. We keep aside the first factor as that represents the mean factor.



We next look at a multi-grid bar chart representation of the factor loadings.

Factor Loadings Bar plot



Annotations of the top gene ontologies of the factors from the FLASH output.

```
ll_f_scale <- apply(ll_deng$f[,-1],2,function(x)</pre>
                                       if(sd(x)!=0) \{return (x/sd(x))\}
                                       else {return (x)}
   })
normalize <- function(x) { return (x/sum(x))}</pre>
abs_f_scale <- apply(ll_deng$f[,-1],2,function(x) normalize(abs(x)))
indices <- CountClust::ExtractTopFeatures(abs_f_scale, top_features=100, method="poisson",
imp_features <- apply(indices, c(1,2), function(x) deng.gene_names[x])</pre>
imp_features[1:6,1:6]
        [,1]
                  [,2]
                            [,3]
                                            [,4]
                                                             [,5]
                                                                      [,6]
##
## [1,] "Kit"
                  "Dgkk"
                            "Rnf141"
                                            "Reep2"
                                                             "Fam46b" "Ston2"
## [2,] "Arl6ip5" "Gm6981" "Aagab"
                                            "Clasp1"
                                                             "Top2b"
                                                                      "Hbs11"
## [3,] "Zfp772"
                  "Cd34"
                            "Herc6"
                                            "Atm"
                                                                      "Fam64a"
                                                             "Wee1"
## [4,] "Ankrd35" "H2-Q10" "1600025M17Rik" "Lox12"
                                                             "H2-T24" "Mmp19"
## [5,] "Lpcat4"
                  "Amotl2" "Smpdl3a"
                                            "Lrrc2"
                                                             "Dusp14" "Fam65a"
## [6,] "Smg1"
                            "Zfp706"
                                            "2310003F16Rik" "Srsf10" "Hdlbp"
                  "Eif5a"
```

We performed Enrichment Analysis on these important features extracted, the top annotations are reported below.

Factors — Gene Ontologies

factor 1 — cell development, neuron projection guidance, stem cell differentiation, mesenchymal cell development, anatomical structure development

factor 2 — bounding membrance organelle, protein C-terminus binding, lyosomal membrane

factor 3 — nucleolus, nuclear lumen, ribosome assembly, nuclear part, RNA binding, nucleotide binding

factor 4 — central nervous system development, cell surface receptor signaling, extracellular space

factor 5 — carbon oxygen lyase activity, chromosomal region, regulatory region nucleic acid binding

factor 6 — extracellular exosome, extracellular vesicle, extracellular organelle, vesicle

factor 7 —

factor 8 — protein localization to cell periphery, plasma membrane organization

factor 9 — sperm-egg recognition, cell-cell recognition, fertilization, regulation of cellular process, substrate-specific transporter activity