Drug repositioning for psoriasis based on cogena

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1 Introduction

This report enables user to reproduce all the results related with GSE13355 in the manuscript, **Drug repositioning and drug mode of action discovery based on co-expressed gene-set enrichment analysis**. The results related with GSE30999 are available at https://github.com/zhilongjia/psoriasis.

2 Data Preparation

2.1 Check package required

2.2 Download the raw data of GSE13355

2.3 Differential Expression Analysis

```
library(GEOquery)
library(affy)
# Download raw data of GSE13355
GSE13355raw <- ReadAffy(celfile.path="../data/GSE13355_RAW")</pre>
sampleNames(GSE13355raw) <- sub("(_|\\.).*CEL\\.gz","", sampleNames(GSE13355raw))</pre>
# Sample Label preprocessing
GSE13355series <- getGEO("GSE13355", destdir="../data")
GSE13355label <- pData(GSE13355series$GSE13355 series matrix.txt.gz)[,c("title", "geo accession")]
GSE13355label$title <- as.character(GSE13355label$title)</pre>
GSE13355label <- GSE13355label[grep("NN", GSE13355label$title, invert = T),]
GSE13355label[grep("PN", GSE13355label$title),"state"] = "ct"
GSE13355label[grep("PP", GSE13355label$title), "state"] = "Psoriasis"
GSE13355label$state <- as.factor(GSE13355label$state)</pre>
GSE13355label[,"gse_id"] = "GSE13355"
GSE13355label$rep <- sapply(strsplit(GSE13355label$title, "_"), "[", 2)
GSE13355raw <- GSE13355raw[,as.character(GSE13355label$geo_accession)]
vmd = data.frame(labelDescription = c("title", "geo_accession", "state", "gse_id", "rep"))
phenoData(GSE13355raw) = new("AnnotatedDataFrame", data = GSE13355label, varMetadata = vmd)
```

```
pData(protocolData(GSE13355raw)) <-</pre>
   pData(protocolData(GSE13355raw))[rownames(GSE13355label),,drop=FALSE]
# RMA normalization
GSE13355rma <- rma(GSE13355raw)
## Background correcting
## Normalizing
## Calculating Expression
# Filter the non-informative and non-expressed genes.
library(MetaDE)
library(annotate)
library(hgu133plus2.db)
GSE13355.Explist <- list(GSE13355=list(x = exprs(GSE13355rma),
                       y = ifelse (GSE13355label$state=="ct", 0, 1),
                       symbol = getSYMBOL(rownames(exprs(GSE13355rma)), "hgu133plus2") ))
GSE13355.Explist <- MetaDE.match(GSE13355.Explist, pool.replicate="IQR")</pre>
GSE13355.Explist.filtered <- MetaDE.filter(GSE13355.Explist, c(0.2,0.2))
colnames(GSE13355.Explist.filtered$GSE13355$x) <- colnames(exprs(GSE13355rma))</pre>
# DEG analysis via limma
DElimma <- function (Expdata, Explabel){</pre>
   library(limma)
   Expdesign <- model.matrix(~as.factor(Explabel$rep) + Explabel$state)</pre>
   Expfit1 <- lmFit(Expdata, Expdesign)</pre>
   Expfit2 <- eBayes(Expfit1)</pre>
   dif_Exp <- topTable(Expfit2, coef=tail(colnames(Expdesign), 1), number=Inf)</pre>
   return (dif_Exp)
}
GSE13355.limma <- DElimma(GSE13355.Explist.filtered$GSE13355$x, GSE13355label)
GSE13355.DE <- GSE13355.limma[GSE13355.limma$adj.P.Val<=0.05 & abs(GSE13355.limma$logFC)>=1,]
GSE13355.DEG <- rownames(GSE13355.DE)
GSE13355.DEG.expr <- GSE13355.Explist.filtered$GSE13355$x[GSE13355.DEG,]
```

3 Co-expression Analysis by cogena

```
# Install cogena if none
library(cogena)
if (packageVersion("cogena") < "1.2.0") {
    devtools::install_github("zhilongjia/cogena")
}
# Parameters for funtion coExp
nClust <- 10 # 10 clusters</pre>
```

4 Pathway Analysis by cogena

```
# Parameters for funtion clEnrich
annoGMT <- "c2.cp.kegg.v5.0.symbols.gmt.xz" # kegg pathway gene set
annofile <- system.file("extdata", annoGMT, package="cogena")</pre>
sampleLabel <- GSE13355label$state</pre>
names(sampleLabel) <- rownames(GSE13355label)</pre>
# cogena analysis (Pathway analysis)
cogena_result <- clEnrich(genecl_result, annofile=annofile, sampleLabel=sampleLabel)</pre>
# Summary the results obtained by cogena
summary(cogena result)
##
## Clustering Methods:
## pam
## The Number of Clusters:
##
## Metric of Distance Matrix:
## correlation
## Agglomeration method for hierarchical clustering (hclust and agnes):
## complete
##
## Gene set:
## c2.cp.kegg.v5.0.symbols.gmt.xz
```

4.1 Figure S1: Heatmap with co-expressed genes

```
# Figure 1
heatmapCluster(cogena_result, "pam", "10", maintitle="Psoriasis (GSE13355)")
```

```
## The number of genes in each cluster:
## upDownGene
## 1 2
## 470 238
## cluster_size
## 1 2 3 4 5 6 7 8 9 10
## 158 65 39 93 50 67 63 94 61 18
```

4.2 Table 1: Co-expressed genes are highly connected

```
# pPPI function: qet the PPI summary information about input genes
pPPI <- function(geneC, string_db){</pre>
    example1_mapped <- string_db$map(as.data.frame(geneC), "geneC",</pre>
                                       removeUnmappedRows = TRUE, quiet=TRUE)
    hits <- example1 mapped$STRING id
    net_summary <- string_db$get_summary(unique(hits))</pre>
    as.numeric( gsub("[^1:9]+\\: |\\)", "", strsplit(net_summary, "\n|\\(")[[1]] ) )
}
# Init table
cluster ppi <- data.frame(protein=numeric(13), interactions=numeric(13),
                              expected_interactions=numeric(13),
                              p_value=numeric(13), stringsAsFactors=FALSE)
rownames(cluster_ppi) <- c(1:10, "Up", "Down", "All_DE")</pre>
# Get PPI information for each cluster.
library(STRINGdb)
suppressWarnings(string_db <- STRINGdb$new(version="10", species=9606,</pre>
                                             score_threshold=400,
                                             input_directory="../tmp"))
for (i in 1:10) {
    i <- as.character(i)</pre>
    cluster_ppi[i,] <- pPPI(geneInCluster(cogena_result, "pam", "10", i), string_db)</pre>
}
cluster_ppi["Up",] <- pPPI(rownames(GSE13355.DE[GSE13355.DE$logFC>0,]), string_db)
cluster_ppi["Down",] <- pPPI(rownames(GSE13355.DE[GSE13355.DE$logFC<0,]), string_db)</pre>
cluster_ppi["All_DE",] <- pPPI(rownames(GSE13355.DE), string_db)</pre>
cluster_ppi$ratio <- cluster_ppi$interactions / cluster_ppi$expected_interactions</pre>
# Table 1
knitr::kable(cluster_ppi, caption="Summary of interactions within clusters")
```

Table 1: Summary of interactions within clusters

ratio	p_value	${\bf expected_interactions}$	interactions	protein	
2.595238	0.0000000	42	109	151	1
2.142857	0.0078911	7	15	62	2
9.250000	0.0000000	8	74	37	3
1.250000	0.2021140	16	20	92	4
23.181818	0.0000000	11	255	49	5
4.400000	0.0000002	5	22	65	6

	protein	interactions	${\tt expected_interactions}$	p_value	ratio
7	61	463	40	0.0000000	11.575000
8	90	19	11	0.0348286	1.727273
9	61	59	16	0.0000000	3.687500
10	18	3	0	0.0048405	Inf
Up	454	1620	635	0.0000000	2.551181
Down	231	235	112	0.0000000	2.098214
All_DE	685	2414	1278	0.0000000	1.888889

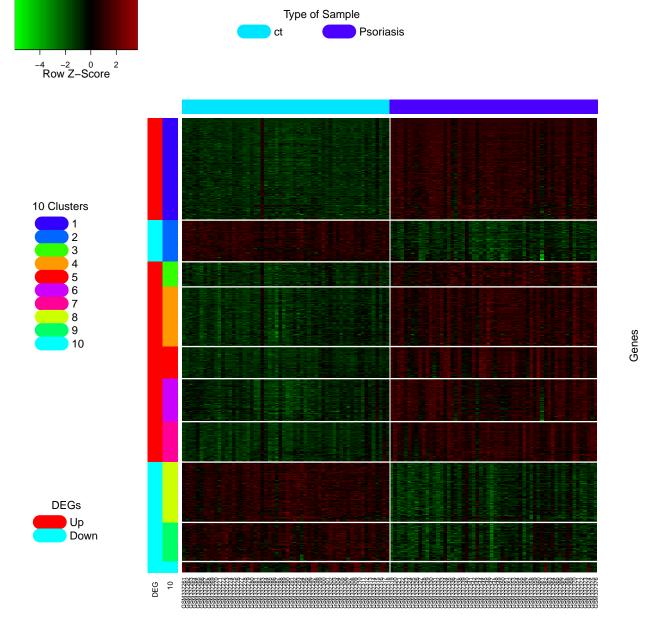
4.3 Figure 2: The result of pathway analysis

```
# Figure 2
heatmapPEI(cogena_result, "pam", "10", printGS=FALSE, maintitle="Psoriasis (GSE13355)")
```

4.4 Table 2 and S1: GSEA

This is to get the Table 2 and S1. The result can be obtained from ../result/GSEA_output, too. See gct and cls file format if needed.

```
# Prepare inputs for GSEA
expData <- as.data.frame(exprs(GSE13355rma))</pre>
expData$DESCRIPTION <- NA
expData <- expData[,c("DESCRIPTION", colnames(expData)[1:116])]</pre>
# Generate gct file
write.table(expData, file="../result/GSEA_input/GSE13355_exp.gct", sep="\t", quote=FALSE)
# Add the following 3 lines at the begining of GSE13355_exp.gct
fConn <- file('../result/GSEA_input/GSE13355_exp.gct', 'r+')</pre>
Lines <- sub("DESCRIPTION", "NAME\tDESCRIPTION", readLines(fConn))</pre>
writeLines(c("#1.2\n54675\t116", Lines), con = fConn)
close(fConn)
# Generate cls file
write.table(t(as.character(GSE13355label$state)),
   file="../result/GSEA_input/GSE13355.cls",quote=FALSE, col.names=FALSE,
   row.names=FALSE)
# Add cls format header
fConn1 <- file('../result/GSEA_input/GSE13355.cls', 'r+')</pre>
writeLines(c("116 2 1\n#ct Psoriasis", readLines(fConn1) ), con = fConn1)
close(fConn1)
# Check GSEA results in ../result/GSEA_output
```



Psoriasis (GSE13355) cogena: pam 10

Color Key

Figure 1: Heatmap with co-expression information

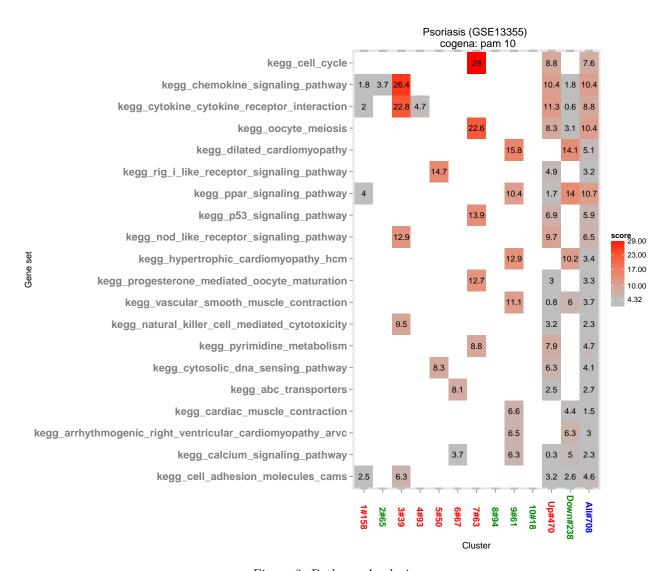


Figure 2: Pathway Analysis

```
# GSEA analysis
if (isTRUE(system("which java", intern=FALSE)==0) & file.exists("gsea2-2.1.0.jar")) {
    system(command="java -cp ./gsea2-2.1.0.jar -Xmx512m xtools.gsea.Gsea -res ../result/GSEA input/GSE1
    warning("Java is not found! GSEA was not run.")
}
# Show the gsea code here
# java -cp ./qsea2-2.1.0.jar -Xmx512m xtools.qsea.Gsea
# -res ../result/GSEA_input/GSE13355_exp.gct
# -cls ../result/GSEA_input/GSE13355.cls
# -gmx ../result/GSEA_input/c2.cp.kegg.v5.0.symbols.gmt
# -collapse true -mode Max_probe
# -norm meandiv -nperm 1000 -permute phenotype -rnd_type no_balance
# -scoring_scheme weighted -rpt_label GSE13355 -metric Signal2Noise
# -sort real -order descending -chip ../result/GSEA_input/HG_U133_Plus_2.chip
# -include_only_symbols true -make_sets false -median false -num 100
# -plot_top_x 20 -rnd_seed 149 -save_rnd_lists false -set_max 500
# -set_min 15 -zip_report false -out ../result/GSEA_output -gui false
```

5 Drug repositioning by cogena

5.1 Figure 3: Drug repositioning for cluster 5

5.2 Figure 4: Drug repositioning for cluster 7

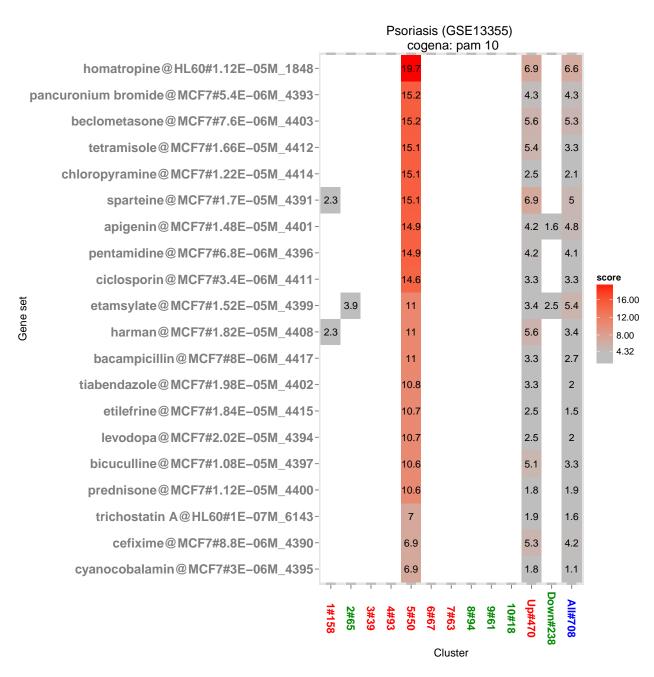


Figure 3: Drug Repositioning for cluster 5

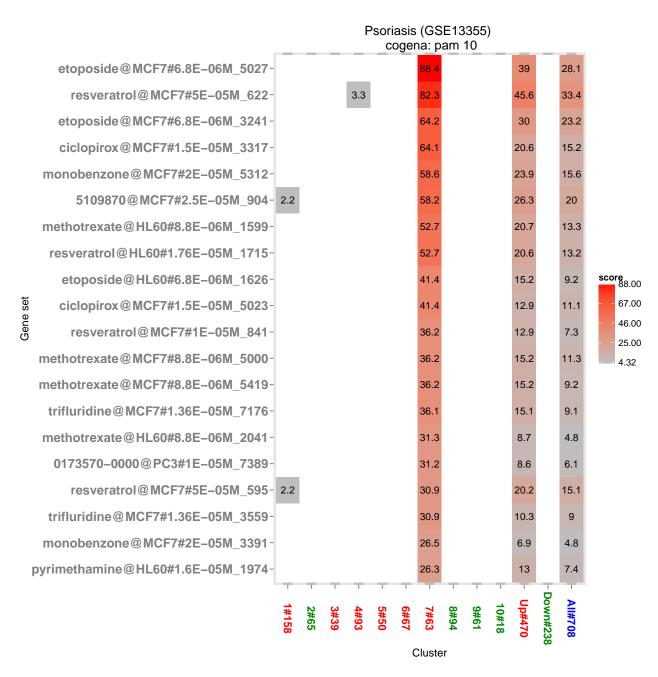


Figure 4: Drug Repositioning for cluster 7

5.3 Figure S2: Drug repositioning for cluster 3

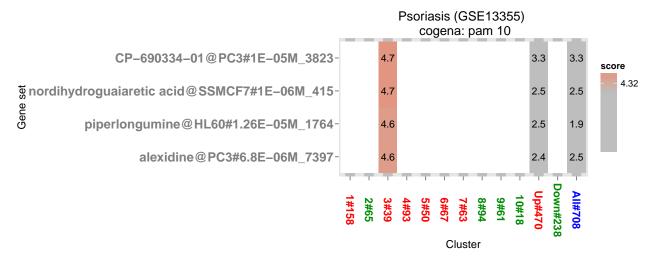


Figure 5: Drug Repositioning for cluster 3

5.4 Figure S3: Drug repositioning for cluster 9

5.5 Table S2: Output DEGs for CMAP and NFFinder Analysis

The input files for CMap and NFF inder , outputed by this chunk, are in $result/CMAP_input/$ and result/NFF inder_input/ respectively. Please visit CMap and NFF inder to get the final results ($Table\ S2$) by vourself.

```
# Convert gene symbols to probes in HGU133a.
symbol2Probe <- function(gs){
    library(hgu133a.db)
    p <- AnnotationDbi::select(hgu133a.db, gs, "PROBEID", "SYMBOL")$PROBEID
    p <- unique(p[which(!is.na(p))])
}

upGene <- rownames(GSE13355.limma[GSE13355.limma$logFC>= 1 & GSE13355.limma$adj.P.Val<=0.05,])
dnGene <- rownames(GSE13355.limma[GSE13355.limma$logFC<= -1 & GSE13355.limma$adj.P.Val<=0.05,])
upProbe <- symbol2Probe(upGene)
dnProbe <- symbol2Probe(dnGene)</pre>
```

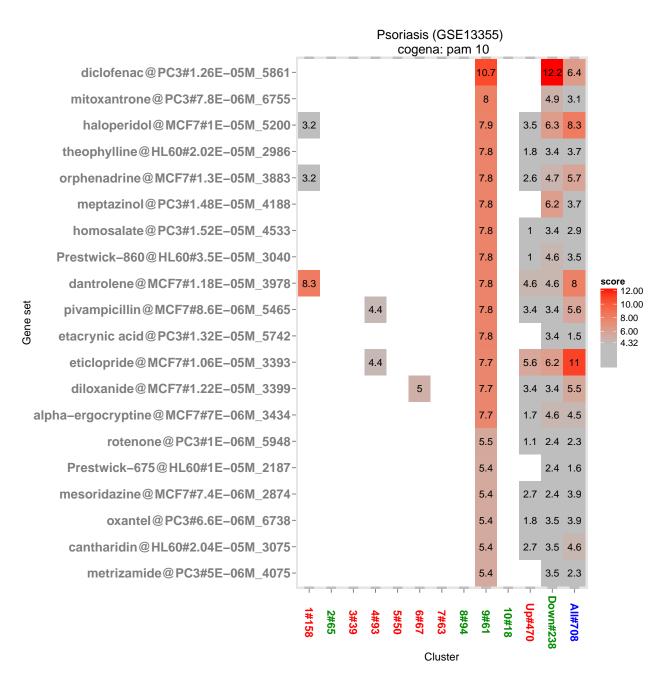


Figure 6: Drug Repositioning for cluster 9

6 Website, BugReports and System Info

- Website: https://github.com/zhilongjia/psoriasis
- BugReports: https://github.com/zhilongjia/psoriasis/issues

sessionInfo()

```
## R version 3.2.0 (2015-04-16)
## Platform: x86_64-pc-linux-gnu (64-bit)
## Running under: Debian GNU/Linux jessie/sid
## locale:
                                   LC_NUMERIC=C
## [1] LC CTYPE=en GB.UTF-8
   [3] LC TIME=en GB.UTF-8
                                   LC COLLATE=en GB.UTF-8
## [5] LC_MONETARY=en_GB.UTF-8
                                   LC_MESSAGES=en_GB.UTF-8
  [7] LC_PAPER=en_GB.UTF-8
                                   LC_NAME=C
  [9] LC_ADDRESS=C
                                   LC_TELEPHONE=C
## [11] LC_MEASUREMENT=en_GB.UTF-8 LC_IDENTIFICATION=C
##
## attached base packages:
  [1] stats4
                                                graphics grDevices utils
                  tools
                            parallel stats
   [8] datasets methods
##
##
## other attached packages:
## [1] hgu133a.db 3.2.2
                              STRINGdb 1.10.0
                                                    hash 2.2.6
## [4] gplots_2.17.0
                              RColorBrewer 1.1-2
                                                    plotrix_3.6
## [7] RCurl_1.95-4.7
                              bitops_1.0-6
                                                    igraph_1.0.1
## [10] plyr_1.8.3
                              sqldf_0.4-10
                                                    gsubfn_0.6-6
## [13] proto_0.3-10
                              png_0.1-7
                                                    cogena_1.5.1
## [16] kohonen_2.0.19
                              MASS 7.3-45
                                                    class 7.3-14
## [19] ggplot2_1.0.1
                              cluster_2.0.3
                                                    limma_3.26.3
## [22] hgu133plus2.db_3.2.2
                              org.Hs.eg.db_3.2.3
                                                    RSQLite_1.0.9000
## [25] DBI_0.3.1.9008
                              annotate_1.48.0
                                                    XML_3.98-1.3
## [28] AnnotationDbi_1.32.0
                              GenomeInfoDb_1.6.1
                                                    IRanges_2.4.4
## [31] S4Vectors_0.8.3
                              MetaDE_1.0.5
                                                    combinat_0.0-8
## [34] impute_1.44.0
                              survival_2.38-3
                                                    hgu133plus2cdf_2.18.0
```

##	[37]	affy_1.48.0	GEOquery_2.36.0	Biobase_2.30.0			
##	[40]	BiocGenerics_0.16.1					
##							
##	loaded via a namespace (and not attached):						
##	[1]	devtools_1.9.1	doParallel_1.0.10	R6_2.1.1			
##	[4]	affyio_1.40.0	KernSmooth_2.23-15	lazyeval_0.1.10.9000			
##	[7]	colorspace_1.2-6	compiler_3.2.0	preprocessCore_1.32.0			
##	[10]	chron_2.3-47	biwt_1.0	formatR_1.2.1			
##	[13]	caTools_1.17.1	scales_0.3.0	DEoptimR_1.0-4			
##	[16]	mvtnorm_1.0-3	robustbase_0.92-5	stringr_1.0.0			
##	[19]	apcluster_1.4.1	digest_0.6.8	rmarkdown_0.8.1			
##	[22]	rrcov_1.3-8	htmltools_0.2.6	highr_0.5.1			
##	[25]	BiocInstaller_1.20.1	mclust_5.1	gtools_3.5.0			
##	[28]	dplyr_0.4.3.9000	magrittr_1.5	Matrix_1.2-3			
##	[31]	Rcpp_0.12.2	munsell_0.4.2	stringi_1.0-1			
##	[34]	yaml_2.1.13	zlibbioc_1.16.0	grid_3.2.0			
##	[37]	gdata_2.17.0	lattice_0.20-33	splines_3.2.0			
##	[40]	knitr_1.11	tcltk_3.2.0	fastcluster_1.1.16			
##	[43]	reshape2_1.4.1	codetools_0.2-14	evaluate_0.8			
##	[46]	foreach_1.4.3	gtable_0.1.2	amap_0.8-14			
##	[49]	assertthat_0.1	xtable_1.8-0	pcaPP_1.9-60			
##	[52]	iterators_1.0.8	memoise_0.2.1	corrplot_0.73			

Thank you!