

Comparison of E(z) mutant brain expression profiles and age-correlated probesets indicates a “younger” brain

Jason Kennerdell

9/25/2017

Input the data and metadata:

```
inpath <- "~/Desktop/brain"
outpath <- "~/Desktop/brain/_8C_orderedList"
setwd(inpath)
files <- list.files()
htseq_files <- files[grepl("^JKL.*txt$", files)]
sampleNames <- read.csv("EzSampleNames.csv")
sampleTable <- data.frame(fileName = htseq_files,
                           stringsAsFactors=FALSE)
sampleTable$Library <- gsub("-counts.txt", "", sampleTable$fileName)
sampleTable$Library <- gsub("b", "", sampleTable$Library)
sampleTable$seq.batch <- ifelse(grepl("b", sampleTable$fileName), "B", "A")
sampleTable$seq.batch <- paste(sampleTable$Library, sampleTable$seq.batch)
sampleTable <- merge(sampleTable[,c(1,3)], sampleNames, by = "seq.batch")
sampleTable$genotype <- gsub("-.*$", "", sampleTable$Sample)
sampleTable$genotype <- factor(sampleTable$genotype, levels = c("w", "Ez"))
sampleTable$Temp <- gsub("[A-Z, a-z, -]", "", sampleTable$Sample)
sampleTable$Temp <- factor(sampleTable$Temp, levels = c("25", "29"))
sampleTable$rep <- gsub("^.*-", "", sampleTable$Sample)
sampleTable$condition <- paste(sampleTable$genotype, sampleTable$Temp, sep = "-")
sampleTable
```

##	seq.batch	fileName	Sample	batch	Library	genotype	Temp	rep
## 1	JKL10 A	JKL10-counts.txt	w-25-II	A	JKL10	w	25	II
## 2	JKL10 B	JKL10b-counts.txt	w-25-II	B	JKL10	w	25	II
## 3	JKL11 A	JKL11-counts.txt	w-25-III	A	JKL11	w	25	III
## 4	JKL11 B	JKL11b-counts.txt	w-25-III	B	JKL11	w	25	III
## 5	JKL12 A	JKL12-counts.txt	Ez-25-III	A	JKL12	Ez	25	III
## 6	JKL12 B	JKL12b-counts.txt	Ez-25-III	B	JKL12	Ez	25	III
## 7	JKL13 A	JKL13-counts.txt	w-29-I	A	JKL13	w	29	I
## 8	JKL13 B	JKL13b-counts.txt	w-29-I	B	JKL13	w	29	I
## 9	JKL14 A	JKL14-counts.txt	Ez-29-I	A	JKL14	Ez	29	I
## 10	JKL14 B	JKL14b-counts.txt	Ez-29-I	B	JKL14	Ez	29	I
## 11	JKL15 A	JKL15-counts.txt	w-29-II	A	JKL15	w	29	II
## 12	JKL15 B	JKL15b-counts.txt	w-29-II	B	JKL15	w	29	II
## 13	JKL16 A	JKL16-counts.txt	Ez-29-II	A	JKL16	Ez	29	II
## 14	JKL16 B	JKL16b-counts.txt	Ez-29-II	B	JKL16	Ez	29	II
## 15	JKL17 A	JKL17-counts.txt	w-29-III	A	JKL17	w	29	III
## 16	JKL17 B	JKL17b-counts.txt	w-29-III	B	JKL17	w	29	III
## 17	JKL18 A	JKL18-counts.txt	Ez-29-III	A	JKL18	Ez	29	III
## 18	JKL18 B	JKL18b-counts.txt	Ez-29-III	B	JKL18	Ez	29	III
## 19	JKL19 A	JKL19-counts.txt	w-29-IV	A	JKL19	w	29	IV
## 20	JKL19 B	JKL19b-counts.txt	w-29-IV	B	JKL19	w	29	IV

```

## 21   JKL20 A   JKL20-counts.txt Ez-29-IV   A   JKL20       Ez   29   IV
## 22   JKL20 B   JKL20b-counts.txt Ez-29-IV   B   JKL20       Ez   29   IV
## 23   JKL21 A   JKL21-counts.txt   w-29-V   A   JKL21       w   29   V
## 24   JKL21 B   JKL21b-counts.txt   w-29-V   B   JKL21       w   29   V
## 25   JKL22 A   JKL22-counts.txt   Ez-29-V   A   JKL22       Ez   29   V
## 26   JKL22 B   JKL22b-counts.txt   Ez-29-V   B   JKL22       Ez   29   V
## 27     JKL7 A     JKL7-counts.txt   w-25-I   A     JKL7       w   25   I
## 28     JKL7 B     JKL7b-counts.txt   w-25-I   B     JKL7       w   25   I
## 29     JKL8 A     JKL8-counts.txt   Ez-25-I   A     JKL8       Ez   25   I
## 30     JKL8 B     JKL8b-counts.txt   Ez-25-I   B     JKL8       Ez   25   I
## 31     JKL9 A     JKL9-counts.txt   Ez-25-II  A     JKL9       Ez   25   II
## 32     JKL9 B     JKL9b-counts.txt   Ez-25-II  B     JKL9       Ez   25   II
##      condition
## 1      w-25
## 2      w-25
## 3      w-25
## 4      w-25
## 5      Ez-25
## 6      Ez-25
## 7      w-29
## 8      w-29
## 9      Ez-29
## 10     Ez-29
## 11     w-29
## 12     w-29
## 13     Ez-29
## 14     Ez-29
## 15     w-29
## 16     w-29
## 17     Ez-29
## 18     Ez-29
## 19     w-29
## 20     w-29
## 21     Ez-29
## 22     Ez-29
## 23     w-29
## 24     w-29
## 25     Ez-29
## 26     Ez-29
## 27     w-25
## 28     w-25
## 29     Ez-25
## 30     Ez-25
## 31     Ez-25
## 32     Ez-25

```

Set up the statistical model

```
design <- formula(~ Temp + genotype)
```

DESeq2 Statistics

```
dds <- DESeqDataSetFromHTSeqCount(sampleTable = sampleTable, directory = inpath, design = design)
# Combine the technical replicates (different runs) by adding the count
# totals for each gene across the two runs:
dds <- collapseReplicates(dds, groupby=dds$Library, run = dds$batch)
dds <- DESeq(dds, betaPrior=T)
# What does the data look like?
head(assay(dds)) # This is the sum of the two runs HTSeq-count output!
```

```
##           JKL10 JKL11 JKL12 JKL13 JKL14 JKL15 JKL16 JKL17 JKL18 JKL19
## FBgn0000003      0      0      0      0      0      0      0      0      0      0
## FBgn0000008  1444  1874  1687  1305  1453  1725  1529  1856  1500  1558
## FBgn0000014      0      0      0      0      0      0      1      0      1      0
## FBgn0000015      1      6      1      3      0      1      1      1      2      0
## FBgn0000017  9186 10798  9189  6877  8808  9121  9834 11313  9272  8564
## FBgn0000018   262   306   286   336   288   317   272   346   274   285
##           JKL20 JKL21 JKL22 JKL7  JKL8  JKL9
## FBgn0000003      1      0      0      0      0      1
## FBgn0000008  1353  1589  1367  1980  1861  1884
## FBgn0000014      1      1      1      0      3      0
## FBgn0000015      0      1      0      0      0      2
## FBgn0000017  8697  8196  8612 11453 11360 11139
## FBgn0000018   285   238   284   286   333   333
```

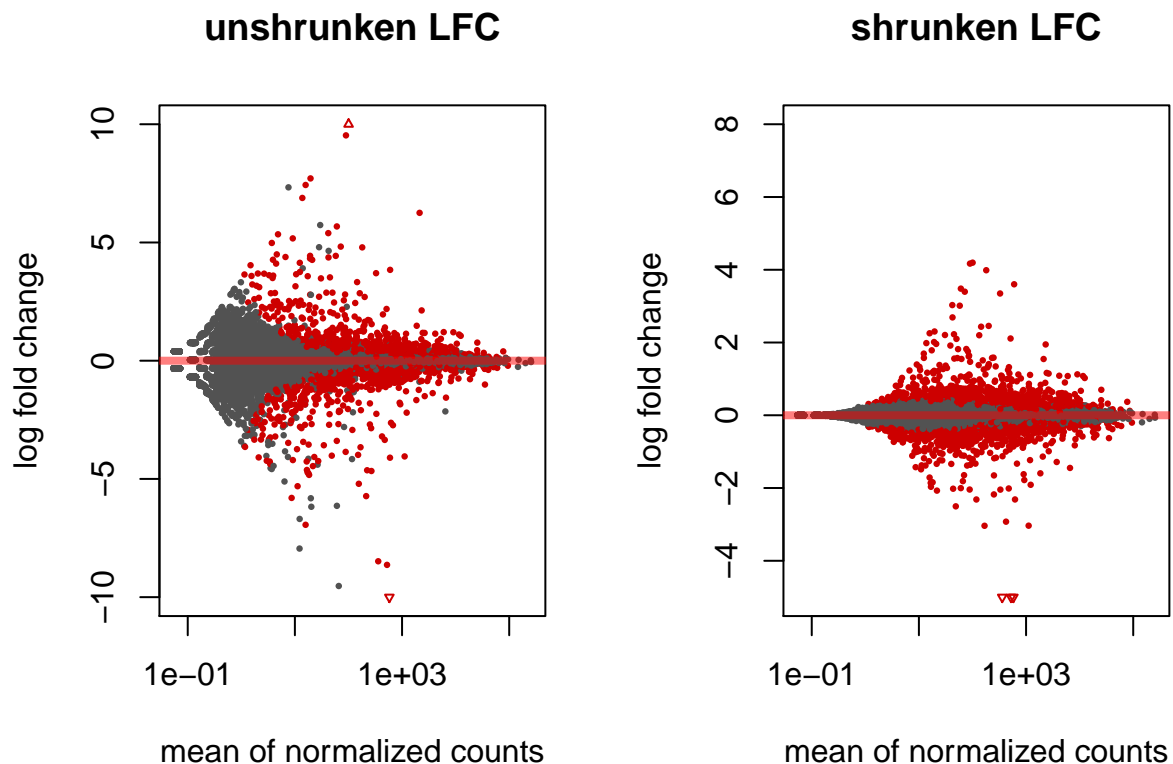
```
# What are the columns?
colData(dds)
```

```
## DataFrame with 16 rows and 9 columns
##           Sample  batch Library genotype  Temp  rep
##           <factor> <factor> <factor> <factor> <factor> <character>
## JKL10    w-25-II      A    JKL10      w      25      II
## JKL11    w-25-III      A    JKL11      w      25      III
## JKL12    Ez-25-III      A    JKL12      Ez      25      III
## JKL13      w-29-I      A    JKL13      w      29      I
## JKL14      Ez-29-I      A    JKL14      Ez      29      I
## ...      ...      ...      ...      ...      ...      ...
## JKL21      w-29-V      A    JKL21      w      29      V
## JKL22      Ez-29-V      A    JKL22      Ez      29      V
## JKL7      w-25-I      A    JKL7      w      25      I
## JKL8      Ez-25-I      A    JKL8      Ez      25      I
## JKL9      Ez-25-II      A    JKL9      Ez      25      II
##           condition runsCollapsed sizeFactor
##           <character> <character> <numeric>
## JKL10      w-25      A,B  0.9090574
## JKL11      w-25      A,B  1.1403424
## JKL12      Ez-25      A,B  1.0027251
## JKL13      w-29      A,B  1.0220507
## JKL14      Ez-29      A,B  0.9570594
## ...      ...      ...      ...
## JKL21      w-29      A,B  0.8750556
## JKL22      Ez-29      A,B  0.9618390
## JKL7      w-25      A,B  1.1341278
## JKL8      Ez-25      A,B  1.1093046
```

```
## JKL9          Ez-25          A,B  1.1303920
results <- results(dds, alpha=0.05)
results$ensembl <- rownames(results)
```

Prepare MA plots:

```
# For Maximum likelihood estimates:
resultsMLE <- results(dds, addMLE=TRUE, alpha = 0.05)
par(mfrow=c(1,2))
plotMA(resultsMLE, MLE=TRUE, alpha = 0.05, main="unshrunk LFC", ylim=c(-10,10))
plotMA(results, alpha = 0.05, main="shrunk LFC", ylim=c(-5,8))
```



Add Annotation

```
# Add usefull gene names:
library(biomaRt)

# Add FBgn Names
mart = useMart("ENSEMBL_MART_ENSEMBL", host="aug2017.archive.ensembl.org")
#listDatasets(mart)
mart = useMart("ENSEMBL_MART_ENSEMBL", host="aug2017.archive.ensembl.org",
              dataset = "dmelanogaster_gene_ensembl")
genemap <- getBM(attributes = c("ensembl_gene_id", "entrezgene", "external_gene_name", "flybasegid_gene"),
                from = "ensembl", to = "ensembl", keys = results$ensembl)
idx <- match(results$ensembl, genemap$ensembl_gene_id)
results$entrez <- genemap$entrezgene[idx]
```

```
results$geneSymbol <- genemap$external_gene_name[idx]
results$cg <- genemap$flybasecgid_gene[idx]
```

Prepare Ranked lists of genes for OrderedList analysis

```
# Using age correlated genes by p value and beta coefficient:
correl_lst <- read.csv("~/Desktop/brain/GSE25007/ranked_aging_correlated_genes_pvalue_only.csv")
# How many genes are there in this list?
dim(correl_lst)[1]
```

```
## [1] 11651
```

```
results_lst <- as.data.frame(results)
results_lst <- results_lst[!is.na(results_lst$pvalue),]
results_lst$product <- -log10(results_lst$pvalue)*results_lst$log2FoldChange
results_lst <- results_lst[order(results_lst$product),]
# How many genes are there in this list?
dim(results_lst)[1]
```

```
## [1] 15167
```

```
# Select only members of lists that are present in the opposite list
results_lst <- results_lst[results_lst$ensembl %in% correl_lst$ensembl,]
head(results_lst)
```

```
##          baseMean log2FoldChange      lfcSE      stat      pvalue
## FBgn0015037  422.7991      -2.925188 0.08788261 -33.28517 6.328883e-243
## FBgn0039321  926.5579      -1.780766 0.05706890 -31.20380 9.462272e-214
## FBgn0033395 1117.2706      -3.035148 0.13523096 -22.44418 1.458786e-111
## FBgn0031489  621.9415      -2.314285 0.10254531 -22.56841 8.856526e-113
## FBgn0003227  331.5243      -2.043967 0.09020525 -22.65906 1.135676e-113
## FBgn0036790 1848.6685      -1.596376 0.06316438 -25.27335 6.272728e-141
##          padj      ensembl entrez geneSymbol      cg      product
## FBgn0015037 8.401241e-240 FBgn0015037  45524      Cyp4p1 CG10842 -708.4766
## FBgn0039321 1.130458e-210 FBgn0039321  43061      CG10550 CG10550 -379.3460
## FBgn0033395 7.577443e-109 FBgn0033395  35946      Cyp4p2  CG1944 -336.4036
## FBgn0031489 4.809496e-110 FBgn0031489  33510      CG17224 CG17224 -259.3220
## FBgn0003227 6.783959e-111 FBgn0003227  49241      rec  CG31293 -230.8553
## FBgn0036790 4.683768e-138 FBgn0036790  40020      AstC-R1 CG7285 -223.8159
```

```
tail(results_lst)
```

```
##          baseMean log2FoldChange      lfcSE      stat      pvalue
## FBgn0052365 200.54575      2.456391 0.08867186 27.70203 6.599880e-169
## FBgn0029114 513.57093      2.110276 0.05834028 36.17186 1.686935e-286
## FBgn0054031 100.54137      4.194846 0.14435159 29.05992 1.152960e-185
## FBgn0052437  89.80844      4.169186 0.14265941 29.22476 9.401423e-188
## FBgn0029924 327.41872      3.347829 0.09073609 36.89633 5.291929e-298
## FBgn0035999 599.31421      3.600854 0.07547901 47.70670 0.000000e+00
##          padj      ensembl entrez geneSymbol      cg      product
## FBgn0052365 6.065289e-166 FBgn0052365  38876      CG32365 CG32365 413.1169
## FBgn0029114 3.358969e-283 FBgn0029114  44497      Tollo  CG6890 603.0597
## FBgn0054031 1.147868e-182 FBgn0054031 3885665      CG34031 CG34031 775.7872
## FBgn0052437 1.021080e-184 FBgn0052437  40366      CG32437 CG32437 779.7496
## FBgn0029924 1.264454e-294 FBgn0029924  31641      CG4586  CG4586 995.2304
```

```
## FBgn0035999 0.000000e+00 FBgn0035999 39097 CG3552 CG3552 Inf
```

```
# How many genes are there in this list?
```

```
dim(results_lst)[1]
```

```
## [1] 11096
```

```
correl_lst <- correl_lst[correl_lst$ensembl %in% results_lst$ensembl,]
```

```
head(correl_lst)
```

```
##      X      ensembl      beta      p.value      prod
## 1 3946 FBgn0031775 0.09066047 9.907021e-08 0.6349911
## 2 8941 FBgn0040735 0.02357414 5.324881e-07 0.1478968
## 3 2651 FBgn0029507 0.08291106 5.737657e-07 0.5174699
## 4 8939 FBgn0040733 0.02273366 4.631188e-06 0.1212683
## 5 3366 FBgn0030808 0.02275069 6.806960e-06 0.1175539
## 6 3046 FBgn0030310 0.04808252 7.391507e-06 0.2467242
```

```
tail(correl_lst)
```

```
##      X      ensembl      beta      p.value      prod
## 11646 6124 FBgn0035649 -0.018821410 4.377408e-05 -0.08203844
## 11647 692 FBgn0004242 -0.005928273 3.225493e-05 -0.02662627
## 11648 10122 FBgn0053202 -0.015904473 2.073152e-05 -0.07448651
## 11649 5636 FBgn0034731 -0.015560635 1.480967e-05 -0.07514938
## 11650 8719 FBgn0039937 -0.030295639 1.348605e-05 -0.14754326
## 11651 8700 FBgn0039900 -0.018599584 4.715846e-06 -0.09906957
```

```
# How many genes are there in this list?
```

```
dim(correl_lst)[1]
```

```
## [1] 11096
```

OrderedList analysis

Test for overlap of Positively age correlated genes and down regulated in $E(z)$ genes, one sided:

```
library(OrderedList)
```

```
## Loading required package: twilight
```

```
## Loading required package: splines
```

```
# Test for overlap of age correlated genes and down regulated in E(z) genes, one sided:
```

```
myComp <- compareLists(ID.List1=results_lst$ensembl,
                       ID.List2=correl_lst$ensembl,
                       mapping=NULL, two.sided = F)
```

```
## Simulating random scores...
```

```
## 0%.....:.....:.....:.....100%
```

```
## -----
```

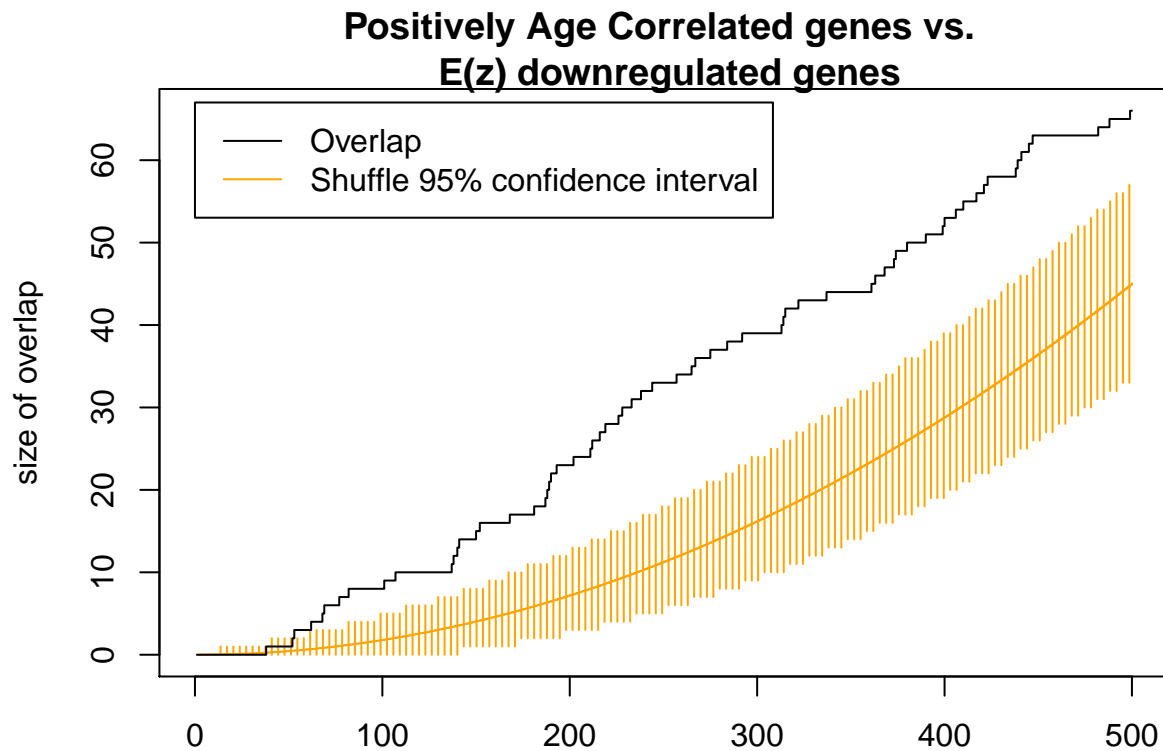
```
myComp
```

```
## List comparison
```

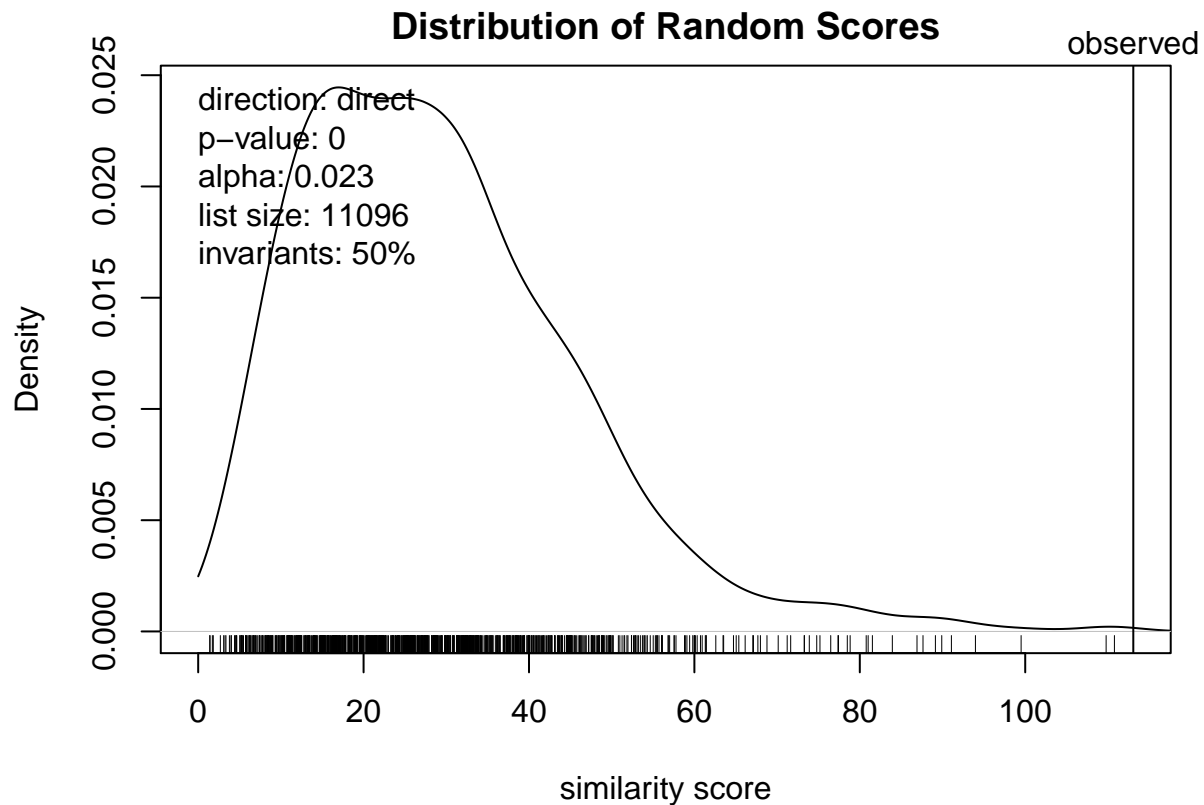
```
## Assessing similarity of : top ranks
```

```
## Length of lists : 11096
## Quantile of invariant genes : 0.5
## Number of random samples : 1000
## -----
## Genes Scores p.values Rev.Scores Rev.p.values
## 0.115 100 0.1750498 0.175 1.000000e-05 0.837
## 0.077 150 1.5421404 0.117 6.755207e-03 0.860
## 0.058 200 5.4372546 0.080 7.045099e-02 0.874
## 0.038 300 23.9592517 0.017 8.581160e-01 0.886
## 0.029 400 59.2457191 0.004 3.450717e+00 0.910
## 0.023 500 113.0714173 0.000 8.752541e+00 0.925
## 0.015 750 336.3657095 0.000 3.921233e+01 0.971
## 0.012 1000 694.8006617 0.000 1.066578e+02 0.988
## 0.008 1500 1845.9043047 0.000 4.269078e+02 0.997
## 0.006 2000 3635.2856954 0.000 1.118397e+03 1.000
## 0.005 2500 6147.6877328 0.000 2.310002e+03 1.000
```

```
myOverlap <- getOverlap(myComp, max.rank=500, percent=.95)
#par(mfrow=c(1,2))
plot(myOverlap, no.title=TRUE, no.legend=TRUE,
     main = "Positively Age Correlated genes vs. \n E(z) downregulated genes")
legend(0,67,legend=c("Overlap", "Shuffle 95% confidence interval"),
     col=c("black", "orange"), lty=c(1,1))
```



```
plot(myOverlap, "scores")
```



```
myOverlap
```

```
## List comparison
##   Assessing similarity of           : top ranks
##   Length of lists                 : 11096
##   Number of random samples        : 1000
## -----
##   Lists are more alike in direct order
##   Chosen regularization parameter   : alpha = 0.023 ( 500 genes)
##   Weighted overlap score            : 113.0714
##   Significance of similarity         : p-value = 0
##   Score percentage for common entries : 95
##   Entries contributing score percentage : 33

# Make a publication plot:
#####
jpeg(file=paste(outpath, "Fig_8c.jpg", sep="/"),
     quality=100,
     res=300,
     width=1440,
     height=960)
par(cex.axis=1.25, cex.lab=2, mar=c(4,5,4,2), col.lab="white")
plot(myOverlap, no.title=TRUE, no.legend=TRUE, cex.lab=2,
     #xlab = NA, #Overlapping gene count",
     #ylab = NA, #Rank of genes in the two lists",
     main = "", las=1)
legend(0,67,legend=c("Overlap", "Shuffle 95% confidence interval"),
      col=c("black", "orange"), lty=c(1,1), cex=.75)
dev.off()
```



```
## pdf
## 2
```

```
#####
```

Test for overlap of age correlated genes and down regulated in E(z) genes, two sided:

```
myComp2 <- compareLists(ID.List1=results_lst$ensembl,
                        ID.List2=correl_lst$ensembl,
                        mapping=NULL, two.sided = T)
```

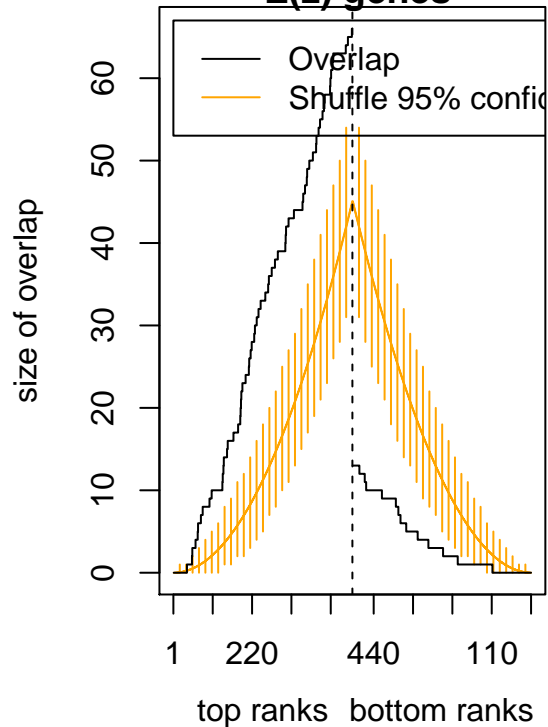
```
## Simulating random scores...
## 0%.....:.....:.....:.....100%
## -----
```

```
myComp2
```

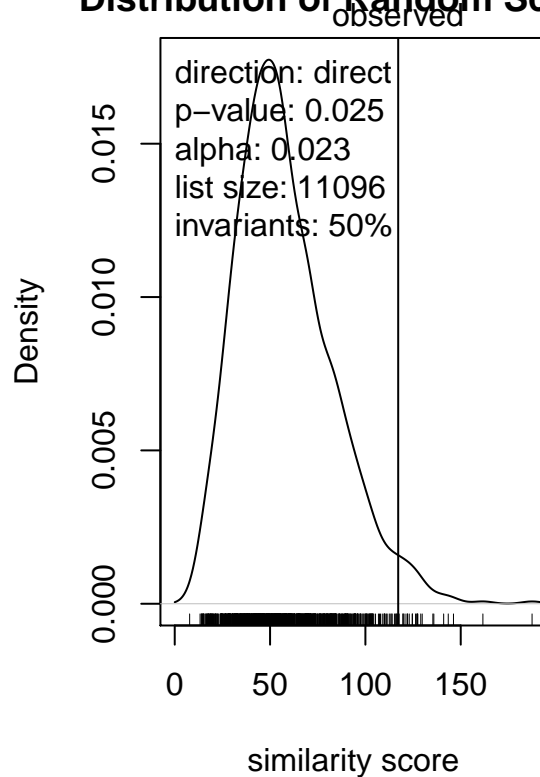
```
## List comparison
## Assessing similarity of      : top and bottom ranks
## Length of lists             : 11096
## Quantile of invariant genes : 0.5
## Number of random samples    : 1000
## -----
##      Genes      Scores p.values  Rev.Scores Rev.p.values
## 0.115   100    0.1750498    0.365 5.793481e-05    0.962
## 0.077   150    1.5449312    0.316 1.541563e-02    0.952
## 0.058   200    5.4688754    0.239 1.561663e-01    0.953
## 0.038   300   24.3598726    0.107 1.889960e+00    0.958
## 0.029   400   60.8640411    0.048 7.645187e+00    0.963
## 0.023   500  117.2033568    0.025 1.948915e+01    0.969
## 0.015   750  355.2388347    0.006 8.626226e+01    0.995
## 0.012  1000  746.3071778    0.005 2.257384e+02    1.000
## 0.008  1500 2048.4685764    0.020 8.266810e+02    1.000
## 0.006  2000 4165.2537622    0.146 2.026616e+03    1.000
## 0.005  2500 7269.9349495    0.549 4.011702e+03    1.000
```

```
myOverlap2 <- getOverlap(myComp2, max.rank=500, percent=.95)
par(mfrow=c(1,2))
plot(myOverlap2, no.title=TRUE, no.legend=TRUE,
     main = "Age Correlated genes vs. \n E(z) genes")
legend(0,67,legend=c("Overlap", "Shuffle 95% confidence interval"),
     col=c("black", "orange"), lty=c(1,1))
plot(myOverlap2, "scores")
```

Age Correlated genes vs. E(z) genes



Distribution of Random Scores



```
myOverlap2
```

```
## List comparison
##   Assessing similarity of           : top and bottom ranks
##   Length of lists                  : 11096
##   Number of random samples         : 1000
## -----
##   Lists are more alike in direct order
##   Chosen regularization parameter   : alpha = 0.023 ( 500 genes)
##   Weighted overlap score            : 117.2034
##   Significance of similarity         : p-value = 0.025
##   Score percentage for common entries : 95
##   Entries contributing score percentage : 36
```

Test for overlap of Negatively age correlated genes and up regulated in E(z) genes, one sided:

```
# Test for overlap of age correlated genes and down regulated in E(z) genes, one sided:
myComp3 <- compareLists(ID.List1=rev(results_lst$ensembl),
                        ID.List2=rev(correl_lst$ensembl),
                        mapping=NULL, two.sided = F)
```

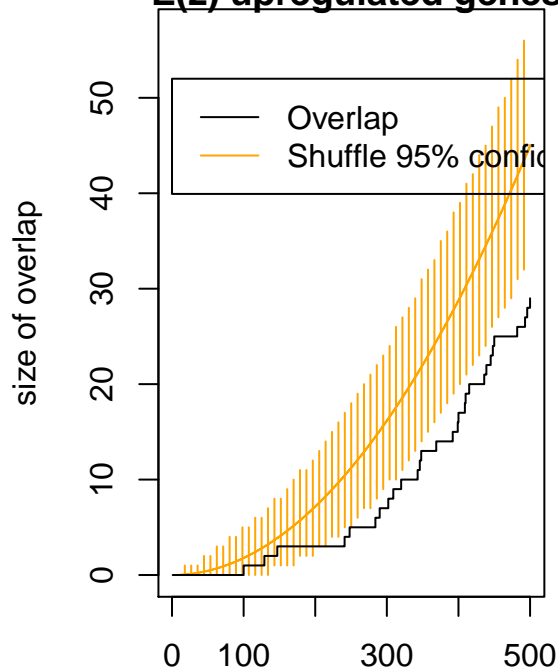
```
##   Simulating random scores...
##   0%.....:.....:.....:.....100%
##   -----
```

```
myComp3
```

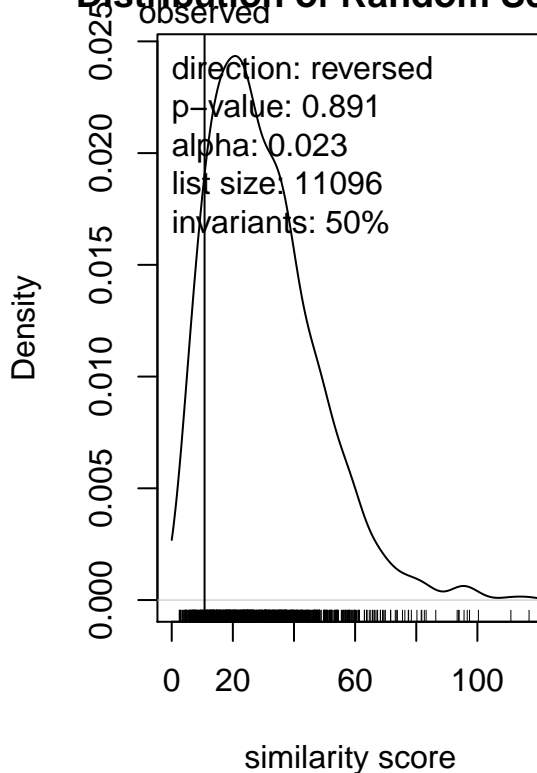
```
## List comparison
## Assessing similarity of      : top ranks
## Length of lists            : 11096
## Quantile of invariant genes : 0.5
## Number of random samples    : 1000
## -----
##      Genes      Scores p.values  Rev.Scores Rev.p.values
## 0.115   100 0.000000e+00    0.846 4.793481e-05    0.821
## 0.077   150 2.790725e-03    0.914 8.660425e-03    0.838
## 0.058   200 3.162078e-02    0.930 8.571532e-02    0.846
## 0.038   300 4.006210e-01    0.949 1.031844e+00    0.857
## 0.029   400 1.618322e+00    0.972 4.194470e+00    0.870
## 0.023   500 4.131939e+00    0.991 1.073661e+01    0.891
## 0.015   750 1.887313e+01    1.000 4.704993e+01    0.926
## 0.012  1000 5.150652e+01    1.000 1.190806e+02    0.974
## 0.008  1500 2.025643e+02    1.000 3.997732e+02    0.999
## 0.006  2000 5.299681e+02    1.000 9.082185e+02    1.000
## 0.005  2500 1.122247e+03    1.000 1.701700e+03    1.000
```

```
myOverlap3 <- getOverlap(myComp3, max.rank=500, percent=.95)
par(mfrow=c(1,2))
plot(myOverlap3, no.title=TRUE, no.legend=TRUE,
     main = "Negatively Age Correlated genes vs. \n E(z) upregulated genes")
legend(0,52,legend=c("Overlap", "Shuffle 95% confidence interval"),
     col=c("black", "orange"), lty=c(1,1))
plot(myOverlap3, "scores")
```

Negatively Age Correlated genes vs. E(z) upregulated genes



Distribution of Random Scores



```
myOverlap3
```

```
## List comparison
##   Assessing similarity of           : top ranks
##   Length of lists                  : 11096
##   Number of random samples         : 1000
## -----
##   Lists are more alike in reversed order
##   Chosen regularization parameter   : alpha = 0.023 ( 500 genes)
##   Weighted overlap score            : 10.73661
##   Significance of similarity         : p-value = 0.891
##   Score percentage for common entries : 95
##   Entries contributing score percentage : 8
# Make a publication plot:
jpeg(file=paste(outpath, "Fig_S5f.jpg", sep="/"), width=360, height=240)
par(cex.axis=1.25, cex.lab=2, mar=c(4,5,4,2), col.lab="white")
plot(myOverlap3, no.title=TRUE, no.legend=TRUE, cex.lab=2,
     main = "", las=1)
legend(0,52,legend=c("Overlap", "Shuffle 95% confidence interval"),
      col=c("black", "orange"), lty=c(1,1), cex=.75)
dev.off()

## pdf
## 2
```

Test for overlap of age correlated genes and up regulated in E(z) genes, two sided:

```
# Test for overlap of age correlated genes and down regulated in E(z) genes, one sided:
myComp4 <- compareLists(ID.List1=rev(results_lst$ensembl),
                        ID.List2=rev(correl_lst$ensembl),
                        mapping=NULL, two.sided = T)

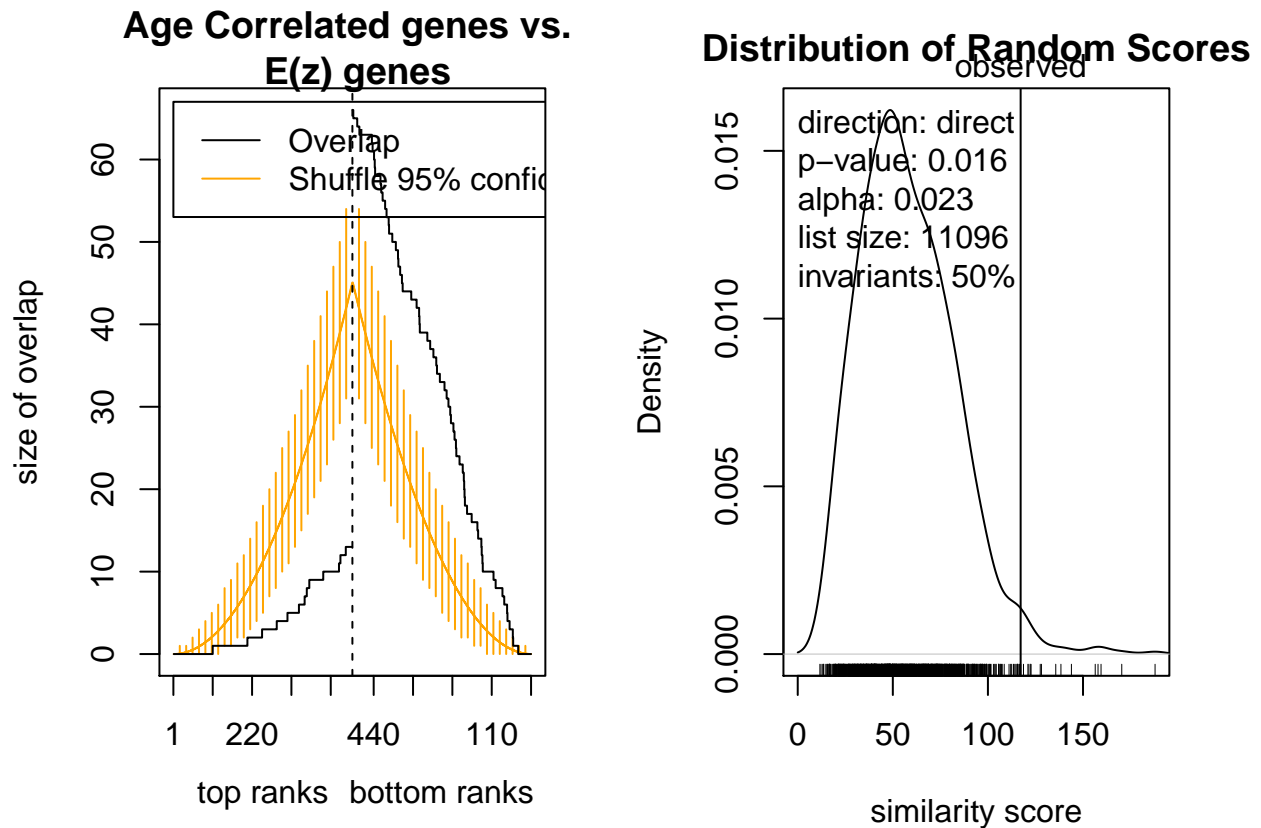
##   Simulating random scores...
##   0%.....:.....100%
##   -----

myComp4
```

```
## List comparison
##   Assessing similarity of           : top and bottom ranks
##   Length of lists                  : 11096
##   Quantile of invariant genes      : 0.5
##   Number of random samples         : 1000
## -----
##           Genes      Scores p.values   Rev.Scores Rev.p.values
## 0.115    100      0.1750498      0.361 5.793481e-05      0.962
## 0.077    150      1.5449312      0.305 1.541563e-02      0.954
## 0.058    200      5.4688754      0.240 1.561663e-01      0.948
## 0.038    300     24.3598726      0.099 1.889960e+00      0.950
## 0.029    400     60.8640411      0.037 7.645187e+00      0.959
## 0.023    500    117.2033568      0.016 1.948915e+01      0.975
## 0.015    750   355.2388347      0.006 8.626226e+01      0.993
```

```
## 0.012 1000 746.3071778 0.004 2.257384e+02 0.999
## 0.008 1500 2048.4685764 0.009 8.266810e+02 1.000
## 0.006 2000 4165.2537622 0.162 2.026616e+03 1.000
## 0.005 2500 7269.9349495 0.547 4.011702e+03 1.000
```

```
myOverlap4 <- getOverlap(myComp4, max.rank=500, percent=.95)
par(mfrow=c(1,2))
plot(myOverlap4, no.title=TRUE, no.legend=TRUE,
     main = "Age Correlated genes vs. \n E(z) genes")
legend(0,67,legend=c("Overlap", "Shuffle 95% confidence interval"),
     col=c("black", "orange"), lty=c(1,1))
plot(myOverlap4, "scores")
```



```
myOverlap4
```

```
## List comparison
## Assessing similarity of : top and bottom ranks
## Length of lists : 11096
## Number of random samples : 1000
## -----
## Lists are more alike in direct order
## Chosen regularization parameter : alpha = 0.023 ( 500 genes)
## Weighted overlap score : 117.2034
## Significance of similarity : p-value = 0.016
## Score percentage for common entries : 95
## Entries contributing score percentage : 36
```

Prepare a ranked list for publication: (Supplemental Data 2)

```
ranked_list <- correl_lst[,c(2,4)]  
rownames(ranked_list) <- 1:dim(ranked_list)[1]  
write.csv(file=file.path(outpath, "rankedAgingGenes.csv"), ranked_list)
```

Save image

```
save.image(file=file.path(outpath, "OrderedList.RData"))
```