

analysis_polyA_vs_pseudo

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Analyzing differential expression results

We have calculated differential expression genes in DE_ribo_vs_polyA.Rmd, now we will try to make sense of them. Our main workhorse will be Gene Set Enrichment Analysis (GSEA) across several reference sets.

```
suppressPackageStartupMessages({
  library(DESeq2)
  library(WebGestaltR)
  library(ggplot2)
  library(yaml)
})

params <- read_yaml("../config.yml")
data_path <- params$data_path
local_data_path <- params$local_data_path
samples <- params$samples

# Load the DESeq2 object with the original count matrix
deseq_path <- paste(local_data_path, "deseq2_output", sep = "/")
dds <- readRDS(paste(deseq_path, "polyA_vs_pseudo_data.rds", sep = "/"))

# Load the DESeqResults object with differentially expressed genes, at FDR 0.1 and 0.05
res1 <- readRDS(paste(deseq_path, "polyA_vs_pseudo_FDR_0.1.rds", sep = "/"))
res05 <- readRDS(paste(deseq_path, "polyA_vs_pseudo_FDR_0.05.rds", sep = "/"))
```

Top genes

Let's look at the top 20 most upregulated and downregulated genes and see if we can find a pattern.

```
res1 <- subset(res1, res1$padj < 0.1)
res1 <- res1[order(res1$log2FoldChange), ]
as.data.frame(head(res1, n=20))
```

##		baseMean	log2FoldChange	lfcSE	stat	pvalue
##	AC000123.1	104.04962	-10.812481	0.8005574	-13.506190	1.437723e-41
##	LAMC1-AS1	146.00598	-9.835680	0.7034012	-13.983030	1.978809e-44
##	LIMS1-AS1	98.41865	-9.668729	0.7090164	-13.636819	2.418826e-42
##	AC005104.1	53.02654	-9.415706	0.7845085	-12.002045	3.466253e-33
##	AC011510.1	55.93759	-9.395642	0.7933071	-11.843638	2.321605e-32
##	AC021683.1	74.80835	-9.049879	0.7903612	-11.450307	2.343137e-30
##	NADK2-AS1	56.18247	-8.946004	0.7076319	-12.642172	1.235924e-36
##	LINC02637	31.00691	-8.891071	0.8310229	-10.698948	1.029402e-26
##	AL355075.2	92.56400	-8.792439	0.5993409	-14.670181	1.000778e-48
##	AP002340.1	34.21704	-8.690400	0.7941363	-10.943209	7.161895e-28
##	AP001793.1	35.96543	-8.605386	0.8157615	-10.548900	5.139547e-26
##	AF235103.1	40.69350	-8.536563	0.7995084	-10.677265	1.300469e-26
##	AC018695.4	53.63662	-8.536246	0.7998947	-10.671712	1.380594e-26
##	AC005884.2	45.77090	-8.437622	0.7716925	-10.933918	7.934707e-28
##	SPDYE2	50.93918	-8.355601	0.7641029	-10.935178	7.825257e-28
##	ETV5-AS1	36.10791	-8.352391	0.7857165	-10.630285	2.154538e-26
##	AL138921.2	67.98953	-8.324670	0.5971517	-13.940628	3.587702e-44
##	AL139349.1	39.62756	-8.303501	0.7869751	-10.551161	5.017322e-26
##	IGKV1-33	36.84540	-8.225531	0.8373767	-9.822975	8.965780e-23
##	AL133406.2	119.17882	-8.192827	0.4586779	-17.861833	2.338310e-71
##		padj				
##	AC000123.1	3.194682e-40				
##	LAMC1-AS1	4.884720e-43				
##	LIMS1-AS1	5.527571e-41				
##	AC005104.1	5.374178e-32				
##	AC011510.1	3.501896e-31				
##	AC021683.1	3.221607e-29				
##	NADK2-AS1	2.230600e-35				
##	LINC02637	1.186519e-25				
##	AL355075.2	2.827490e-47				
##	AP002340.1	8.716903e-27				
##	AP001793.1	5.710147e-25				
##	AF235103.1	1.491236e-25				
##	AC018695.4	1.579983e-25				
##	AC005884.2	9.641285e-27				
##	SPDYE2	9.512291e-27				
##	ETV5-AS1	2.444429e-25				
##	AL138921.2	8.729671e-43				
##	AL139349.1	5.580782e-25				
##	IGKV1-33	8.449153e-22				
##	AL133406.2	1.308504e-69				

Reminder, these ^ are the ones that are much more expressed in true bulk than in pseudobulk. This looks like a lot of lncRNAs. None of the genes jump out as significant (except for maybe the immunoglobulin gene IHGJ3P).

```
as.data.frame(tail(res1, n=20))
```

##	baseMean	log2FoldChange	lfcSE	stat	pvalue
## ICAM4	71.61889	8.931028	0.8025448	11.12838	9.127426e-29
## AP005329.3	73.00113	9.193496	0.7837153	11.73066	8.876524e-32
## AC005523.2	64.39783	9.218436	0.7812961	11.79890	3.954423e-32
## CIDEA	68.12531	9.320185	0.7825104	11.91062	1.041985e-32
## AL391121.1	104.98773	9.466573	0.7703664	12.28840	1.045498e-34
## LINC02591	153.19954	9.518911	0.7586782	12.54670	4.144025e-36
## MAFIP	74.03265	9.749916	0.7826940	12.45687	1.283120e-35
## RPS10-NUDT3	80.45463	9.790623	0.7955178	12.30723	8.281369e-35
## AC092069.1	185.49030	9.876999	0.7676609	12.86636	6.960854e-38
## AL133453.1	198.81626	10.155562	0.7585382	13.38833	7.075115e-41
## HIST1H2AK	190.71820	10.391822	0.7689105	13.51500	1.275629e-41
## DDT	5426.62151	10.578877	0.3302520	32.03274	3.818486e-225
## AC087190.1	258.54147	10.594199	0.7694527	13.76848	3.943974e-43
## SFT2D3	189.19177	10.660077	0.7623820	13.98259	1.991004e-44
## AC020656.1	562.90434	11.351167	0.7730433	14.68374	8.194215e-49
## EVA1B	772.85972	11.691198	0.7500108	15.58804	8.778475e-55
## HYI	508.34796	12.111364	0.7697252	15.73466	8.752764e-56
## ZNF593	914.83142	12.617987	0.7581973	16.64209	3.453547e-62
## C19orf33	3002.60065	13.981387	0.7680679	18.20332	4.857007e-74
## MIF	12368.57330	14.850642	1.0777692	13.77906	3.406942e-43
##	padj				
## ICAM4	1.168949e-27				
## AP005329.3	1.300946e-30				
## AC005523.2	5.900228e-31				
## CIDEA	1.589124e-31				
## AL391121.1	1.727430e-33				
## LINC02591	7.310414e-35				
## MAFIP	2.204391e-34				
## RPS10-NUDT3	1.372999e-33				
## AC092069.1	1.329270e-36				
## AL133453.1	1.515101e-39				
## HIST1H2AK	2.843238e-40				
## DDT	2.569130e-222				
## AC087190.1	9.284224e-42				
## SFT2D3	4.910633e-43				
## AC020656.1	2.337937e-47				
## EVA1B	3.056198e-53				
## HYI	3.126249e-54				
## ZNF593	1.520770e-60				
## C19orf33	2.867716e-72				
## MIF	8.046225e-42				

These ^ are the genes that are much more expressed in pseudobulk than true bulk. There are two histone genes, which I find interesting. I'm also interested that there's more MIF and DDT, which are apparently closely related inflammatory cytokines. It could be a coincidence though if there just happened to be more macrophages in that set.

GeneCards says both EVA1B and SFT2D3 are "predicted to be an integral component of membrane"... is that anything?

GSEA

WebGestaltR expects a data frame with two columns, gene name and fold change.

```
res1$gene <- rownames(res1); rownames(res1) <- NULL
res1 <- subset(res1, select=c("gene","log2FoldChange"))
res1 <- as.data.frame(res1)
nrow(res1)
```

```
## [1] 19820
```

```
res05 <- subset(res05, res05$padj < 0.05)
res05$gene <- rownames(res05); rownames(res05) <- NULL
res05 <- subset(res05, select=c("gene","log2FoldChange"))
res05 <- as.data.frame(res05)
nrow(res05)
```

```
## [1] 18433
```

GO Biological process

Our first try at GSEA will use the same reference set we used for overrepresentation analysis in the single-cell data, GO Biological process.

```
GO_bp <- suppressWarnings(WebGestaltR(enrichMethod = "GSEA",
                                     enrichDatabase = "geneontology_Biological_Process_noRedundant",
                                     interestGene = res1,
                                     interestGeneType = "genesymbol",
                                     isOutput = FALSE))
```

```
## Loading the functional categories...
## Loading the ID list...
## Performing the enrichment analysis...
## 1000 permutations of score complete...
```

```
nrow(GO_bp)
```

```
## [1] 31
```

```
GO_bp <- GO_bp[order(GO_bp$normalizedEnrichmentScore, decreasing = TRUE),]
head(subset(GO_bp, select=c("geneSet","description","normalizedEnrichmentScore","pValue","FDR","size")))
```

```
##      geneSet                                     description
## 1 GO:0033108 mitochondrial respiratory chain complex assembly
## 2 GO:0010257                                NADH dehydrogenase complex assembly
## 3 GO:0017004                                cytochrome complex assembly
## 4 GO:0140053                                mitochondrial gene expression
## 5 GO:0006414                                translational elongation
## 6 GO:0009141                                nucleoside triphosphate metabolic process
##      normalizedEnrichmentScore pValue          FDR size
```

```
## 1          3.180506      0 0.0000000000    83
## 2          2.862794      0 0.0000000000    52
## 3          2.818635      0 0.0000000000    32
## 4          2.559436      0 0.0000000000   137
## 5          2.520298      0 0.0000000000   112
## 6          2.336421      0 0.0001532636   231
```

```
tail(subset(GO_bp, select=c("geneSet", "description", "normalizedEnrichmentScore", "pValue", "FDR", "size")))
```

```
##          geneSet                                description
## 22 GO:0014074          response to purine-containing compound
## 13 GO:0035249          synaptic transmission, glutamatergic
## 12 GO:0007606          sensory perception of chemical stimulus
## 10 GO:0051932          synaptic transmission, GABAergic
## 14 GO:0098742 cell-cell adhesion via plasma-membrane adhesion molecules
## 15 GO:0043062          extracellular structure organization
##      normalizedEnrichmentScore pValue          FDR size
## 22              -1.980920      0 0.022446001   104
## 13              -2.036757      0 0.013361557    53
## 12              -2.076490      0 0.011134631    95
## 10              -2.109470      0 0.009543969    22
## 14              -2.111255      0 0.014315954   183
## 15              -2.148563      0 0.015906615   294
```

```
GO_bp_05 <- suppressWarnings(WebGestaltR(enrichMethod = "GSEA",
                                         enrichDatabase = "geneontology_Biological_Process_noRedundant",
                                         interestGene = res05,
                                         interestGeneType = "genesymbol",
                                         isOutput = FALSE))
```

```
## Loading the functional categories...
## Loading the ID list...
## Performing the enrichment analysis...
## 1000 permutations of score complete...
```

```
nrow(GO_bp_05)
```

```
## [1] 32
```

```
GO_bp_05 <- GO_bp_05[order(GO_bp_05$normalizedEnrichmentScore, decreasing = TRUE),]
head(subset(GO_bp_05, select=c("geneSet", "description", "normalizedEnrichmentScore", "pValue", "FDR", "size")))
```

```
##          geneSet                                description
## 1 GO:0033108 mitochondrial respiratory chain complex assembly
## 2 GO:0010257          NADH dehydrogenase complex assembly
## 3 GO:0017004          cytochrome complex assembly
## 4 GO:0006414          translational elongation
## 5 GO:0140053          mitochondrial gene expression
## 7 GO:0009141          nucleoside triphosphate metabolic process
##      normalizedEnrichmentScore pValue          FDR size
## 1              3.208698      0 0.0000000000    83
```

```
## 2          2.968032      0 0.0000000000    52
## 3          2.770416      0 0.0000000000    32
## 4          2.479687      0 0.0000000000   111
## 5          2.478813      0 0.0000000000   136
## 7          2.347948      0 0.0004560144   222
```

```
tail(subset(GO_bp_05, select=c("geneSet","description","normalizedEnrichmentScore","pValue","FDR","size")))
```

```
##          geneSet          description
## 20 GO:0014074      response to purine-containing compound
## 12 GO:0098742 cell-cell adhesion via plasma-membrane adhesion molecules
## 11 GO:0007606      sensory perception of chemical stimulus
## 13 GO:0035249      synaptic transmission, glutamatergic
## 10 GO:0051932      synaptic transmission, GABAergic
## 14 GO:0043062      extracellular structure organization
##      normalizedEnrichmentScore pValue      FDR size
## 20          -1.983351      0 0.025626879    97
## 12          -2.097999      0 0.004911818   169
## 11          -2.120663      0 0.004271146    84
## 13          -2.121953      0 0.005338933    49
## 10          -2.149639      0 0.003737253    21
## 14          -2.159137      0 0.006406720   277
```

GO Cellular Component

```
GO_cc <- suppressWarnings(WebGestaltR(enrichMethod = "GSEA",
                                     enrichDatabase = "geneontology_Cellular_Component_noRedundant",
                                     interestGene = res05,
                                     interestGeneType = "genesymbol",
                                     isOutput = FALSE))
```

```
## Loading the functional categories...
## Loading the ID list...
## Performing the enrichment analysis...
## 1000 permutations of score complete...
```

```
nrow(GO_cc)
```

```
## [1] 21
```

```
GO_cc <- GO_cc[order(GO_cc$normalizedEnrichmentScore, decreasing = TRUE),]
head(subset(GO_cc, select=c("geneSet","description","normalizedEnrichmentScore","pValue","FDR","size")))
```

```
##          geneSet          description normalizedEnrichmentScore pValue FDR
## 1 GO:0098798 mitochondrial protein complex      3.338596      0 0
## 2 GO:0070469      respiratory chain      3.191677      0 0
## 3 GO:0005743 mitochondrial inner membrane      3.084723      0 0
## 4 GO:0030964      NADH dehydrogenase complex      3.080469      0 0
## 5 GO:0005840      ribosome      3.058882      0 0
```

```
## 6 GO:0044455 mitochondrial membrane part 3.006808 0 0
## size
## 1 219
## 2 74
## 3 337
## 4 41
## 5 193
## 6 177
```

```
tail(subset(GO_cc, select=c("geneSet", "description", "normalizedEnrichmentScore", "pValue", "FDR", "size")))
```

```
##      geneSet      description normalizedEnrichmentScore      pValue
## 19 GO:0005844      polysome      1.720138 0.008948546
## 21 GO:0030684      preribosome      1.641510 0.023310023
## 20 GO:0045177      apical part of cell      -1.859651 0.000000000
## 18 GO:0016323 basolateral plasma membrane      -1.951507 0.000000000
## 14 GO:0005581      collagen trimer      -2.198581 0.000000000
## 15 GO:0031012      extracellular matrix      -2.238606 0.000000000
##      FDR size
## 19 0.0281917885 57
## 21 0.0499569907 52
## 20 0.0282576159 235
## 18 0.0126594119 129
## 14 0.0009042437 56
## 15 0.0009042437 305
```

```
GO_cc_05 <- suppressWarnings(WebGestaltR(enrichMethod = "GSEA",
  enrichDatabase = "geneontology_Cellular_Component_noRedundant",
  interestGene = res05,
  interestGeneType = "genesymbol",
  isOutput = FALSE))
```

```
## Loading the functional categories...
## Loading the ID list...
## Performing the enrichment analysis...
## 1000 permutations of score complete...
```

```
nrow(GO_cc_05)
```

```
## [1] 21
```

```
GO_cc_05 <- GO_cc_05[order(GO_cc_05$normalizedEnrichmentScore, decreasing = TRUE),]
head(subset(GO_cc_05, select=c("geneSet", "description", "normalizedEnrichmentScore", "pValue", "FDR", "size")))
```

```
##      geneSet      description normalizedEnrichmentScore pValue FDR
## 1 GO:0098798 mitochondrial protein complex      3.391830 0 0
## 2 GO:0070469      respiratory chain      3.207390 0 0
## 3 GO:0030964      NADH dehydrogenase complex      3.126552 0 0
## 4 GO:0005743 mitochondrial inner membrane      3.103100 0 0
## 5 GO:0005840      ribosome      3.047249 0 0
## 6 GO:0044455 mitochondrial membrane part      3.017820 0 0
```

```
##      size
## 1    219
## 2     74
## 3     41
## 4    337
## 5    193
## 6    177
```

```
tail(subset(GO_cc_05, select=c("geneSet", "description", "normalizedEnrichmentScore", "pValue", "FDR", "size")))
```

```
##      geneSet      description normalizedEnrichmentScore      pValue
## 20 GO:0005844      polysome          1.692982 0.011709602
## 21 GO:0030684      preribosome        1.648616 0.008988764
## 19 GO:0045177      apical part of cell -1.848681 0.000000000
## 18 GO:0016323 basolateral plasma membrane -1.931832 0.000000000
## 12 GO:0031012      extracellular matrix -2.244989 0.000000000
## 13 GO:0005581      collagen trimer    -2.282803 0.000000000
##      FDR size
## 20 0.03809260   57
## 21 0.04950971   52
## 19 0.02865283  235
## 18 0.01576664  129
## 12 0.00000000  305
## 13 0.00000000   56
```

Cell types

Let's try a custom set for cell types, as curated by the folks at <http://www.gsea-msigdb.org/>

```
C8 <- suppressWarnings(WebGestaltR(enrichMethod = "GSEA",
  enrichDatabaseFile = "GSEA_custom_sets/c8.all.v7.5.1.symbols.gmt",
  enrichDatabaseType = "genesymbol",
  interestGene = res1,
  interestGeneType = "genesymbol",
  isOutput = FALSE))
```

```
## Loading the functional categories...
## Loading the ID list...
## Performing the enrichment analysis...
## 1000 permutations of score complete...
```

```
nrow(C8)
```

```
## [1] 175
```

```
C8 <- C8[order(C8$normalizedEnrichmentScore, decreasing = TRUE),]
head(subset(C8, select=c("geneSet", "normalizedEnrichmentScore", "pValue", "FDR", "size")))
```

```
##      geneSet
## 1      DESCARTES_FETAL_LUNG_LYMPHOID_CELLS
```



```
## 2          DESCARTES_FETAL_PANCREAS_LYMPHOID_CELLS
## 3          DESCARTES_FETAL_HEART_LYMPHOID_CELLS
## 4 FAN_OVARY_CL10_PUTATIVE_EARLY_ATRESIA_GRANULOSA_CELL
## 5          DESCARTES_FETAL_ADRENAL_LYMPHOID_CELLS
## 6          DESCARTES_FETAL_INTESTINE_LYMPHOID_CELLS
##   normalizedEnrichmentScore pValue FDR size
## 1          3.614286      0 0 115
## 2          3.588763      0 0 108
## 3          3.527898      0 0 81
## 4          3.517152      0 0 233
## 5          3.517135      0 0 116
## 6          3.433727      0 0 119
```

```
tail(subset(C8, select=c("geneSet", "normalizedEnrichmentScore", "pValue", "FDR", "size")))
```

```
##                                     geneSet
## 89          DESCARTES_FETAL_EYE_VASCULAR_ENDOTHELIAL_CELLS
## 81          DESCARTES_FETAL_MUSCLE_SMOOTH_MUSCLE_CELLS
## 75          AIZARANI_LIVER_C13_LSECS_2
## 64          GAO_LARGE_INTESTINE_ADULT_CJ_IMMUNE_CELLS
## 65 DESCARTES_FETAL_PLACENTA_VASCULAR_ENDOTHELIAL_CELLS
## 66          AIZARANI_LIVER_C10_MVECS_1
##   normalizedEnrichmentScore pValue      FDR size
## 89          -2.214887      0 0.0005080715 72
## 81          -2.235256      0 0.0003048429 50
## 75          -2.275383      0 0.0001905268 212
## 64          -2.350739      0 0.0000000000 356
## 65          -2.379345      0 0.0000000000 75
## 66          -2.411041      0 0.0000000000 209
```

```
C8_05 <- suppressWarnings(WebGestaltR(enrichMethod = "GSEA",
                                     enrichDatabaseFile = "GSEA_custom_sets/c8.all.v7.5.1.symbols.gmt",
                                     enrichDatabaseType = "genesymbol",
                                     interestGene = res05,
                                     interestGeneType = "genesymbol",
                                     isOutput = FALSE))
```

```
## Loading the functional categories...
## Loading the ID list...
## Performing the enrichment analysis...
## 1000 permutations of score complete...
```

```
nrow(C8_05)
```

```
## [1] 179
```

```
C8_05 <- C8_05[order(C8_05$normalizedEnrichmentScore, decreasing = TRUE),]
head(subset(C8_05, select=c("geneSet", "normalizedEnrichmentScore", "pValue", "FDR", "size")))
```

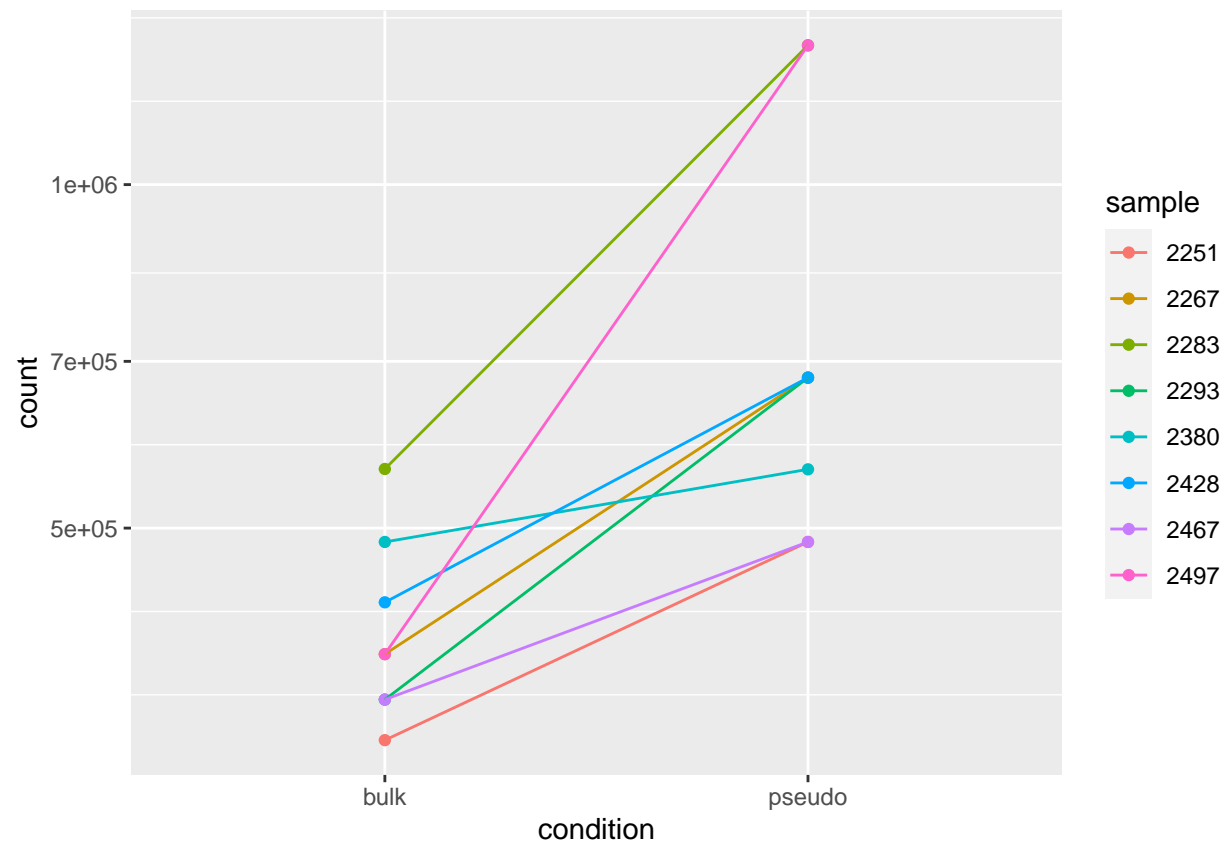
```
##                                     geneSet
## 1          DESCARTES_FETAL_LUNG_LYMPHOID_CELLS
```

```
## 2          DESCARTES_FETAL_HEART_LYMPHOID_CELLS
## 3          DESCARTES_FETAL_ADRENAL_LYMPHOID_CELLS
## 4 FAN_OVARY_CL10_PUTATIVE_EARLY_ATRESIA_GRANULOSA_CELL
## 5          DESCARTES_FETAL_PANCREAS_LYMPHOID_CELLS
## 6          DESCARTES_FETAL_INTESTINE_LYMPHOID_CELLS
##   normalizedEnrichmentScore pValue FDR size
## 1          3.617819         0 0 109
## 2          3.560492         0 0 77
## 3          3.545725         0 0 108
## 4          3.529760         0 0 227
## 5          3.510799         0 0 104
## 6          3.472424         0 0 115
```

```
tail(subset(C8_05, select=c("geneSet", "normalizedEnrichmentScore", "pValue", "FDR", "size")))
```

```
##                                     geneSet
## 64          DESCARTES_FETAL_MUSCLE_SMOOTH_MUSCLE_CELLS
## 65    DESCARTES_FETAL_LIVER_VASCULAR_ENDOTHELIAL_CELLS
## 66          AIZARANI_LIVER_C13_LSECS_2
## 67          GAO_LARGE_INTESTINE_ADULT_CJ_IMMUNE_CELLS
## 68 DESCARTES_FETAL_PLACENTA_VASCULAR_ENDOTHELIAL_CELLS
## 69          AIZARANI_LIVER_C10_MVECS_1
##   normalizedEnrichmentScore pValue FDR size
## 64          -2.277272         0 0 46
## 65          -2.302453         0 0 85
## 66          -2.356499         0 0 196
## 67          -2.410707         0 0 338
## 68          -2.423104         0 0 71
## 69          -2.434836         0 0 202
```

```
gene <- "MT-ATP6"
d <- plotCounts(dds, gene=gene, intgroup=c("condition", "sample"), returnData=TRUE)
ggplot(d, aes(x=condition, y=count, group=sample, color=sample)) +
  geom_point() + scale_y_log10() + geom_line()
```



Conclusions