

Supplementary methods S4

limma-voom analysis

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Introduction

The limma-voom analysis is performed 6 times, one for each tissue/statistical approach. Here, as an example, we show how it is performed with leaf samples against reference. For the other analysis, we follow the same pipeline, adjusting the variables to each wanted condition.

Librarys

List of libraries used in this pipeline:

```
library(Biobase)
library(knitr)
library(limma)
library(edgeR)
library(gplots)
library(ggplot2)
library(RColorBrewer)
library(ggrepel)
library(dplyr)
library(metaMA)
library(xlsx)
library(SummarizedExperiment)
library(reshape)
```

Pipeline - Leaf samples (against reference)

Data input:

```
#1-. Load the SE objects:  
load("/home/albert/Desktop/Lolium/P226/se_P226.RData")  
#2-. Extract leaf samples:  
se.f <- se_P226[, se_P226$Tissue == "Leaf"]
```

Extracting the count matrices and the phenotype:

```
#1-. We extract the phenotype contained in the columns  
pheno <- colData(se.f)  
#2-. We set 35% as our first level:  
pheno$Soil_Humidity <- relevel(pheno$Soil_Humidity, "35%")  
#3-. We extract the count matrices  
counts <- assay(se.f)  
#4-. We rename the count matrices according to the phenotype  
colnames(counts) <- pheno$SampleName
```

Independent Filtering:

```
#1-. We establish our independent filtering threshold  
keep <- rowSums(counts)>0  
#2-. We apply the threshold to our counting matrices  
counts.f <- counts[keep,]
```

Data normalization:

```
#1-. We create a dge object:  
dge <- DGEList(counts=counts.f, group=pheno$Soil_Humidity)  
#2-. We normalize our data using the TMM method:  
dge.n <- calcNormFactors(dge, method = "TMM")
```

Model Building:

```
#1-. We set the design for our model:  
design <- model.matrix(~0+dge.n$samples$group)  
#2-. We change the rownames of our design:  
rownames(design) <- colnames(dge.n)  
#3-. We change the colnames of our design:  
colnames(design) <- c("Leaf_35", "Leaf_01", "Leaf_05", "Leaf_15")  
#4-. We apply the voom transformation:  
dge.v <- voom(dge.n, design)  
#5-. We fit our model:  
fit <- lmFit(dge.v, design)
```

Data testing:

35% vs 15% analysis:

```
#1-. We contrast the conditions we want to assess:
contrast.matrix.1 <-makeContrasts(Leaf_15-Leaf_35,levels=design)
#2-. We fit the contrast matrix to the model:
fitc.1 <-contrasts.fit(fit,contrast.matrix.1)
#3-. We apply the eBayes test:
fite.1 <-eBayes(fitc.1)
#4-. We obtain our genes results:
res_35vs15 <-topTable(fite.1,number=Inf,adjust="BH")
#5-. We transform our data into a data.frame:
res.df.1 <- as.data.frame(res_35vs15)
#6-. We get rid off those genes with NA values:
res.df.1 <- res.df.1[complete.cases(res.df.1),]
#7-. We order the results by
res.ord.1 <- res.df.1[order(res.df.1$adj.P.Val, decreasing = TRUE), ]
#8-. We save the results in a .csv file:
write.csv(file="P226_limma_Results_Leaf_AR_1.csv", x=res.ord.1)
#9-. We save the results in an excel file:
write.xlsx(res.ord.1, file="P226_limma_Results_Leaf_AR_1.xlsx", row.names = TRUE)
```

15% vs 05% analysis:

```
#1-. We contrast the conditions we want to assess:
contrast.matrix.2 <-makeContrasts(Leaf_05-Leaf_35,levels=design)
#2-. We fit the contrast matrix to the model:
fitc.2 <-contrasts.fit(fit,contrast.matrix.2)
#3-. We apply the eBayes test:
fite.2 <-eBayes(fitc.2)
#4-. We obtain our genes results:
res_35vs05 <-topTable(fite.2,number=Inf,adjust="BH")
#5-. We transform our data into a data.frame:
res.df.2 <- as.data.frame(res_35vs05)
#6-. We get rid off those genes with NA values:
res.df.2 <- res.df.2[complete.cases(res.df.2),]
#7-. We order the results by
res.ord.2 <- res.df.2[order(res.df.2$adj.P.Val, decreasing = TRUE), ]
#8-. We save the results in a .csv file:
write.csv(file="P226_limma_Results_Leaf_AR_2.csv", x=res.ord.2)
#9-. We save the results in an excel file:
write.xlsx(res.ord.2, file="P226_limma_Results_Leaf_AR_2.xlsx", row.names = TRUE)
```

05% vs 01% analysis:

```
#1-. We contrast the conditions we want to assess:
contrast.matrix.3 <-makeContrasts(Leaf_01-Leaf_35,levels=design)
#2-. We fit the contrast matrix to the model:
fitc.3 <-contrasts.fit(fit,contrast.matrix.3)
#3-. We apply the eBayes test:
```

```

fite.3 <-eBayes(fitc.3)
#4-. We obtain our genes results:
res_35vs01 <-topTable(fite.3,number=Inf,adjust="BH")
#5-. We transform our data into a data.frame::
res.df.3 <- as.data.frame(res_35vs01)
#6-. We get rid off those genes with NA values:
res.df.3 <- res.df.3[complete.cases(res.df.3),]
#7-. We order the results by
res.ord.3 <- res.df.3[order(res.df.3$adj.P.Val, decreasing = TRUE), ]
#8-. We save the results in a .csv file:
write.csv(file="P226_limma_Results_Leaf_AR_3.csv", x=res.ord.3)
#9-. We save the results in an excel file:
write.xlsx(res.ord.3, file="P226_limma_Results_Leaf_AR_3.xlsx", row.names = TRUE)

```

Data saving:

```

#1-. We create an .RData object containing all data generated:
save.image(file= "P226_limma_Leaf_AR.RData")

```