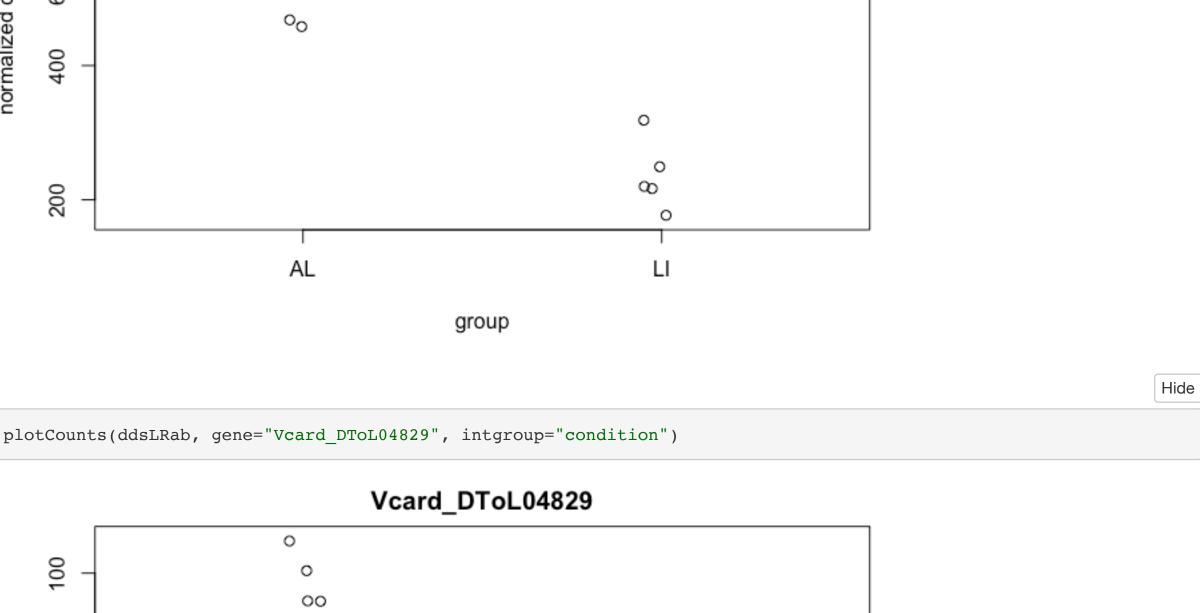
```
Code ▼
Vanessa cardui DE
Differntial expression in Vanessa cardui adults
Experimental set up was prepared with adult females in two food availability conditions: ad libitum and no resources, to analyze the influence that
environmental conditions may have on investment in migration or reproduction in imagines
Abdomen tissue
In this section we analyse differential expression in the abdomen tissue
Preparing data We upload count data from the DESeq2 object crated within the nf-core rnasq pipeline. Pipeline was run for the entire
experimental dataset therefore needs to be split into different sets/tissues:
Pipeline was run for the entire experimental dataset therefore needs to be split into different sets/tissues:
                                                                                                                          Hide
 colDataLRab<-colDataLR[colDataLR$tissue == "A", ] #selecting abdomen samples only</pre>
 #selecting samples with abdomen tissue based on ID
 countLRab<-countLR[,c(1,2,3,4,5,10,11,12,13,14)]
 names(colDataLRab)[names(colDataLRab)=="Group3"]<-"tissue"</pre>
 names(colDataLRab)[names(colDataLRab)=="Group1"]<-"condition"</pre>
 head(countLR)
                  AL_ADF_A_11 AL_ADF_A_19 AL_ADF_A_3 AL_ADF_A_7 AL_ADF_A_9 AL_ADF_H_11 AL_ADF_H_3 AL_ADF_H_7
 Vcard_DToL00001
                                                                              0
 Vcard_DToL00002
                                                                                                                    0
 Vcard_DToL00003
                                                                              0
                                                                                                                    0
 Vcard_DToL00004
                                                                              0
 Vcard_DToL00005
 Vcard_DToL00006
                  AL_ADF_H_9 LI_ADF_A_10 LI_ADF_A_12 LI_ADF_A_14 LI_ADF_A_16 LI_ADF_A_8 LI_ADF_H_10 LI_ADF_H_12
                                                                   0
 Vcard_DToL00001
 Vcard_DToL00002
                                                                                                          0
                                                                                                                       0
 Vcard_DToL00003
                                                                                                                       0
                                       0
                                                                                                                       0
 Vcard_DToL00004
                                                                                                                       0
 Vcard_DToL00005
                                         0
                            0
 Vcard_DToL00006
                  LI_ADF_H_14 LI_ADF_H_16 LI_ADF_H_8
 Vcard DToL00001
                             0
                                          0
 Vcard_DToL00002
 Vcard_DToL00003
 Vcard_DToL00004
 Vcard_DToL00005
 Vcard_DToL00006
                                                                                                                          Hide
 head(colDataLR)
 DataFrame with 6 rows and 5 columns
                    sample condition
                                              tissue
                                                           Group4 sizeFactor
              <character> <character> <character> <character> <numeric>
AL_ADF_A_11 AL_ADF_A_11 AL A 11 1.117485
AL_ADF_A_19 AL_ADF_A_19 AL A 19 1.035097
AL_ADF_A_3 AL_ADF_A_3 AL A 3 1.056977
AL_ADF_A_7 AL_ADF_A_7 AL A 7 0.883751
AL_ADF_A_9 AL_ADF_A_9 AL A 9 1.579290
AL_ADF_H_11 AL_ADF_H_11 AL A H 11 0.826602
Constructing DESeq2 object, at this point counts are transformed according to the DESeq specific algorithm. Current hypothsis includes split into
treatment groups only (~condition), filtering out genes with less then 10 total counts
                                                                                                                          Hide
 ddsLRab <- DESeqDataSetFromMatrix(countData=countLRab, colData =colDataLRab, design = ~condition)
 some variables in design formula are characters, converting to factors
                                                                                                                          Hide
 #Basic filtering
 keep <- rowSums(counts(ddsLRab)) >= 10
 ddsLRab[keep,]
 head(ddsLRab)
 class: DESeqDataSet
 dim: 6 10
 metadata(1): version
 assays(1): counts
 rownames(6): Vcard DToL00007 Vcard DToL00010 ... Vcard DToL00013 Vcard DToL00014
 rowData names(0):
 colnames(10): AL_ADF_A_11 AL_ADF_A_19 ... LI_ADF_A_16 LI_ADF_A_8
 colData names(5): sample condition tissue Group4 sizeFactor
DESeq analysis Main valuation, calculation of p-values
                                                                                                                          Hide
 #ddsLRab <- DESeq(ddsLRab) #running the main function</pre>
 resLRab <- results(ddsLRab)</pre>
 resLRab <- resLRab[order(resLRab$padj),]</pre>
 head(resLRab)
 log2 fold change (MLE): condition LI vs AL
 Wald test p-value: condition LI vs AL
 DataFrame with 6 rows and 6 columns
                                                                         pvalue
                    baseMean log2FoldChange
                                                  lfcSE
                                                              stat
                                                                                        padj
                                   <numeric> <numeric> <numeric> <numeric>
                   <numeric>
                                                                                  <numeric>
 Vcard_DToL13592 541.427
                                    -1.89555 0.313908 -6.03855 1.55504e-09 1.56422e-05
 Vcard DToL04829
                    52.077
                                    -2.30664 0.401831 -5.74031 9.45037e-09 4.75306e-05
                    274.921 1.23667 0.269397 4.59051 4.42161e-06 1.48257e-02
478.949 -2.46656 0.567757 -4.34440 1.39658e-05 3.27025e-02
 Vcard DToL13539
                    478.949
                                    -2.46656 0.567757 -4.34440 1.39658e-05 3.27025e-02
 Vcard_DToL02600
                               -1.37088 0.323629 -4.23596 2.27575e-05 3.27025e-02
 Vcard DToL10357
                    334.966
                                   4.34707 1.023610 4.24681 2.16838e-05 3.27025e-02
 Vcard_DToL13220 160.525
                                                                                                                          Hide
 summary(resLRab)
 out of 11237 with nonzero total read count
 adjusted p-value < 0.1</pre>
 LFC > 0 (up)
                   : 10, 0.089%
 LFC < 0 \text{ (down)} : 12, 0.11%
 outliers [1] : 112, 1%
 low counts [2] : 1066, 9.5%
 (mean count < 5)
 [1] see 'cooksCutoff' argument of ?results
 [2] see 'independentFiltering' argument of ?results
Setting up p-value to 0.05
                                                                                                                          Hide
 res05 <- results(ddsLRab, alpha=0.05)</pre>
 summary(res05)
 out of 11237 with nonzero total read count
 adjusted p-value < 0.05
 LFC > 0 (up)
                   : 2, 0.018%
 LFC < 0 \text{ (down)} : 5, 0.044%
 outliers [1] : 112, 1%
 low counts [2]
                   : 0, 0%
 (mean count < 1)
 [1] see 'cooksCutoff' argument of ?results
 [2] see 'independentFiltering' argument of ?results
Log fold change shrinkage is recommended for visualization and ranking, her we follow main DESEq2 vignette exactly
"Shrinkage of effect size (LFC estimates) is useful for visualization and ranking of genes. To shrink the LFC, we pass the dds object to the function
IfcShrink. Below we specify to use the apeglm method for effect size shrinkage (Zhu, Ibrahim, and Love 2018), which improves on the previous
estimator.
We provide the dds object and the name or number of the coefficient we want to shrink, where the number refers to the order of the coefficient as
it appears in resultsNames(dds)."
                                                                                                                          Hide
 head(resultsNames(ddsLRab))
 [1] "Intercept"
                             "condition_LI_vs_AL"
                                                                                                                          Hide
 resLFC <- lfcShrink(ddsLRab, coef="condition_LI_vs_AL", type="apeglm")</pre>
 Error in lfcShrink(ddsLRab, coef = "condition_LI_vs_AL", type = "apeglm") :
   type='apeglm' requires installing the Bioconductor package 'apeglm'
Visual exploration of the results
We us first 100 significant genes to build a PCA plot:
                                                                                                                          Hide
 vsdata <- vst(ddsLRab, blind=FALSE)</pre>
 plotPCA(vsdata, intgroup=c("condition"))
    20 -
 PC2: 17% variance
                                                                                         group
   -10 -
             -25
                                                             25
                                                                                     50
                                      PC1: 43% variance
We use function plotMA, that shows the log2 fold changes attributable to a given variable over the mean of normalized counts for all the samples
in the DESeqDataSet. Points are colored blue if the adjusted p-value is less than 0.1.
                                                                                                                          Hide
 plotMA(resLRab, ylim=c(-5,5))
      7
log fold change
      ņ
                                                   1e+03
                                                                            1e+05
             1e+00
                         1e+01
                                      1e+02
                                                               1e+04
                                      mean of normalized counts
Checking significant gene candidates:
                                                                                                                          Hide
 res <- resLRab[order(resLRab$padj),]</pre>
 head(res, 10)
 log2 fold change (MLE): condition LI vs AL
 Wald test p-value: condition LI vs AL
 DataFrame with 10 rows and 6 columns
                    baseMean log2FoldChange
                                                  lfcSE
                                                              stat
                                                                         pvalue
                                                                                        padj
                                   <numeric> <numeric> <numeric>
                                                                      <numeric>
                                                                                   <numeric>
                   <numeric>
 Vcard_DToL04829 52.0770
                                    -2.30664 0.401831 -5.74031 9.45037e-09 4.75306e-05
 Vcard DToL13539 274.9208
                                   1.23667 0.269397 4.59051 4.42161e-06 1.48257e-02
 Vcard DToL02600 478.9490
                                    -2.46656 0.567757 -4.34440 1.39658e-05 3.27025e-02
                                    -1.37088 0.323629 -4.23596 2.27575e-05 3.27025e-02
 Vcard_DToL10357 334.9660
                                     4.34707 1.023610 4.24681 2.16838e-05 3.27025e-02
 Vcard_DToL13220 160.5249
 Vcard_DToL16018
                   24.6975
                                    -2.76963 0.650293 -4.25905 2.05300e-05 3.27025e-02
 Vcard_DToL01847 664.3799
                                    1.67992 0.429905 3.90765 9.31980e-05 8.13203e-02
                                    -1.32700 0.334322 -3.96925 7.21005e-05 8.13203e-02
 Vcard_DToL04245 4685.0017
 Vcard_DToL05076 100.0051
                                    -2.60856 0.677126 -3.85240 1.16968e-04 8.13203e-02
                                                                                                                          Hide
 #Plotting these genes
 plotCounts(ddsLRab, gene="Vcard_DToL13592", intgroup="condition")
                                       Vcard_DToL13592
      1200
                               oo
                               0
      800
normalized count
      9
                              ^{\circ}
                                                                    0
                                                                     0
                                                                    ^{\circ}
      200
                                                                      0
                               ΑL
                                                                     LI
                                                group
```



8

0

0

0

LI

Hide

Hide

Hide

Hide

normalized count

50

20

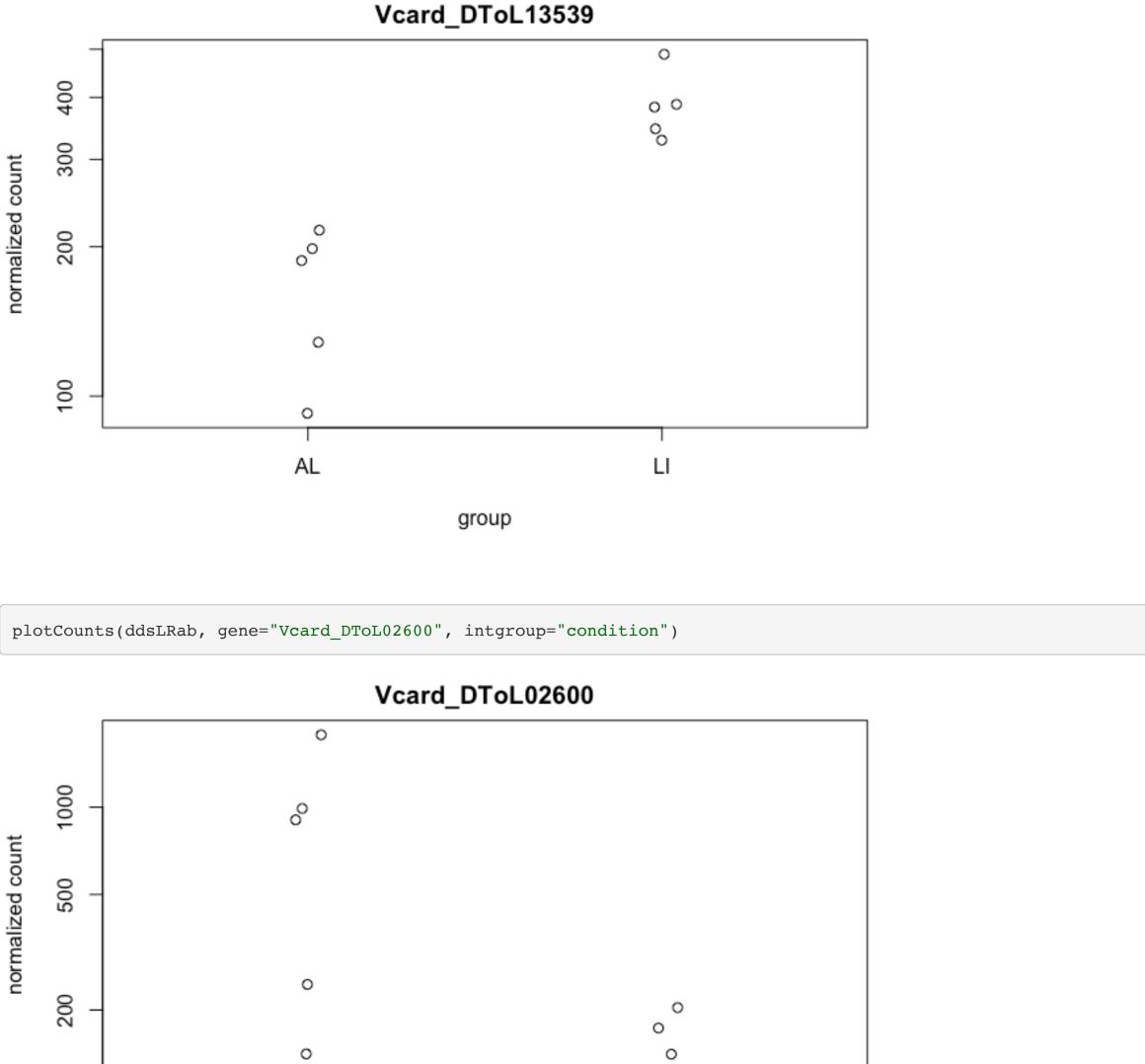
10

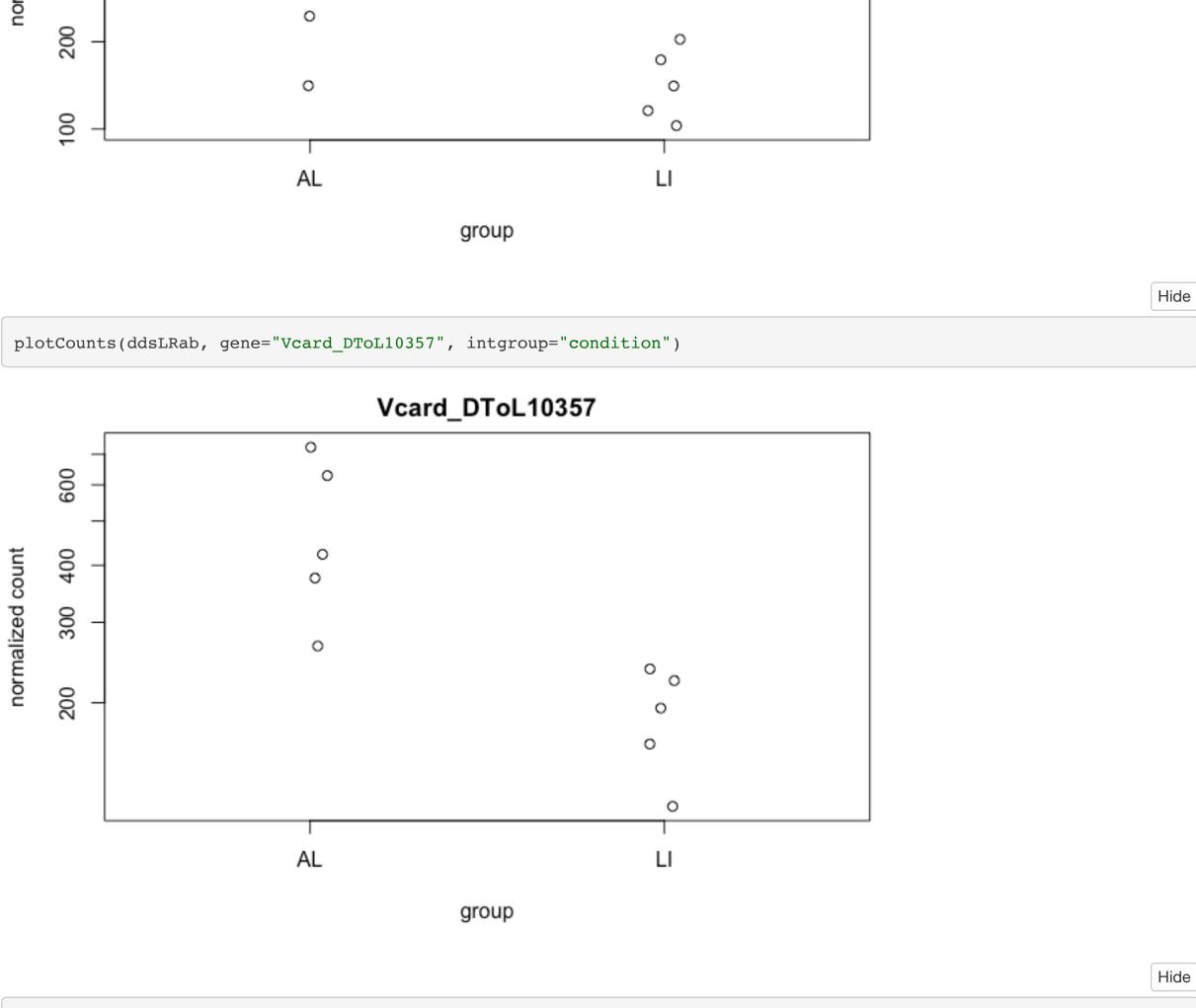
0

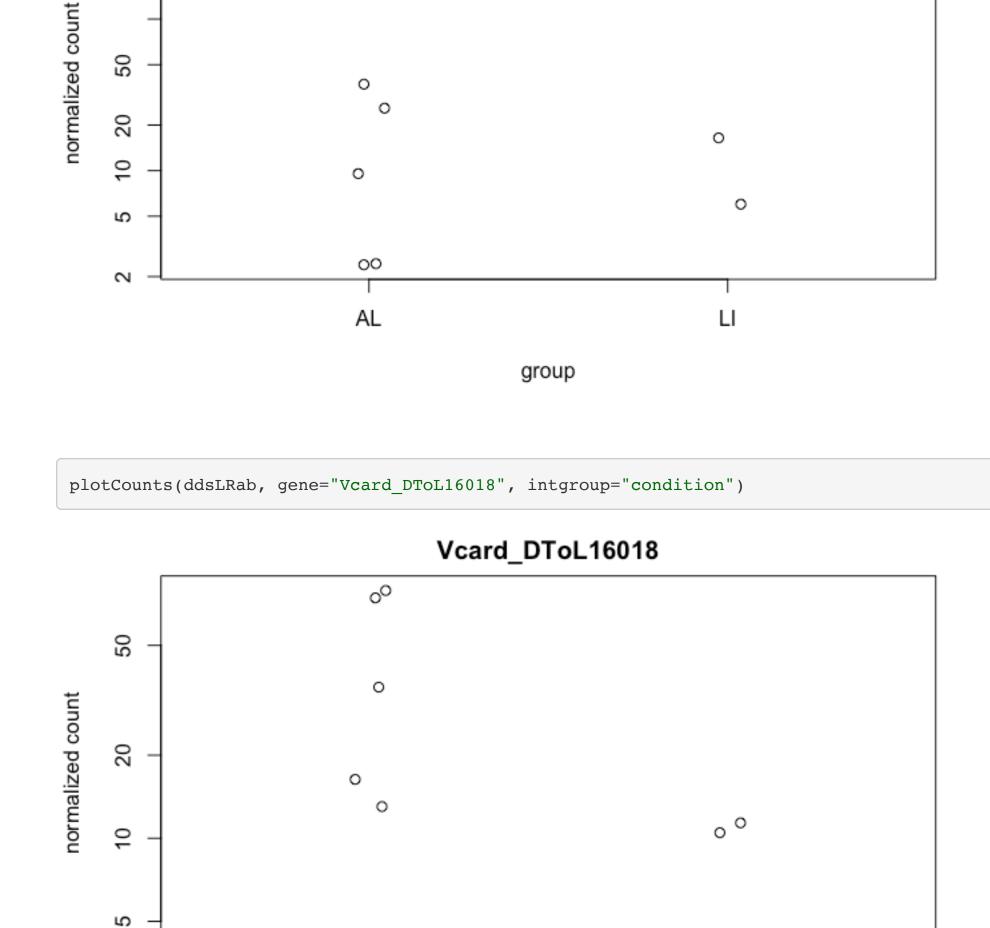
ΑL

plotCounts(ddsLRab, gene="Vcard_DToL13539", intgroup="condition")

group







plotCounts(ddsLRab, gene="Vcard DToL13220", intgroup="condition")

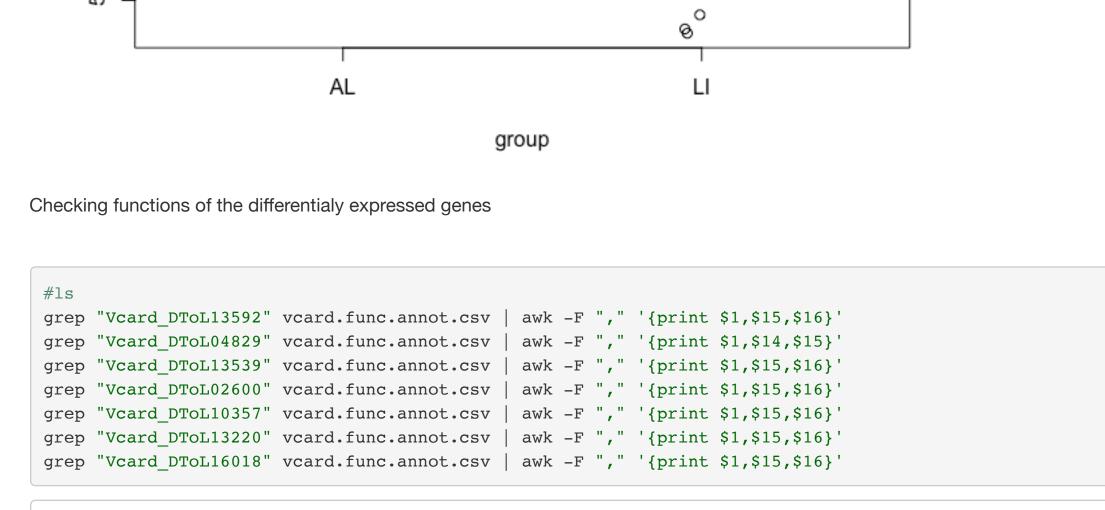
500

200

Vcard_DToL13220

၀

0



Vcard_DToL13592-RA - -Vcard_DToL04829-RA "Zinc-binding domain present in Lin-11 Isl-1 Vcard_DToL13539-RA "Insect pheromone-binding family A10/OS-D" Vcard_DToL02600-RA SCP / Tpx-1 / Ag5 / PR-1 / Sc7 family of extracellular domains. glipr2 Vcard DToL10357-RA Leucine Rich repeat chp Vcard_DToL13220-RA Insect cuticle protein -Vcard_DToL16018-RA Protein of unknown function (DUF1676) Osi9