

RWC23_ELT2_ChIP_Boeck_Time_Resolved_RNA

Note: Ensure BioConductor is version 3.10 or above

Install libraries

```
# fill this in  
# install.packages("fpc")  
# install.packages("vcd")
```

Note: you must load biomaRt before loading tidyverse

Load libraries

```
library(biomaRt)  
library(tidyverse)
```

```
## -- Attaching packages -----  
  
## v ggplot2 3.3.0      v purrr  0.3.4  
## v tibble  3.0.1      v dplyr  0.8.5  
## v tidyr   1.0.3      v stringr 1.4.0  
## v readr   1.3.1      v forcats 0.5.0  
  
## -- Conflicts -----  
## x dplyr::filter() masks stats::filter()  
## x dplyr::lag()     masks stats::lag()  
## x dplyr::select() masks biomaRt::select()
```

```
library(ComplexHeatmap)
```

```
## Loading required package: grid
```

```
## =====  
## ComplexHeatmap version 2.2.0  
## Bioconductor page: http://bioconductor.org/packages/ComplexHeatmap/  
## Github page: https://github.com/jokergoo/ComplexHeatmap  
## Documentation: http://jokergoo.github.io/ComplexHeatmap-reference  
##  
## If you use it in published research, please cite:  
## Gu, Z. Complex heatmaps reveal patterns and correlations in multidimensional  
## genomic data. Bioinformatics 2016.  
## =====
```

Load custom functions

```
source("../RWC23_Functions.R")
```

Pseudocode: - Bring in Boeck Data - Translate to WBGeneID - Filter for ELT-2 ChIP bound genes, make heatmap - Filter for intestine expressed genes (spencer data), make heatmap, add row annotation for binding cluster

Import Time-resolved RNA

```
time_resolved_rna <-  
  read.delim(  
    "../02_Public_Intesine_RNA/01_input/9_Boeck_et_al_2016_time-resolved_transcriptome/Unified_dcpm_per  
    quote = "",  
    stringsAsFactors = FALSE  
  )  
  
paramart <-  
  useMart("parasite_mart",  
    dataset = "wbps_gene",  
    host = "https://parasite.wormbase.org",  
    port = 443)  
  
time_resolved_rna <- getBM(  
  mart = paramart,  
  filter = c("wormbase_gseqname"),  
  value = time_resolved_rna$WormbaseName,  
  attributes = c("wormbase_gseq", "wbps_gene_id", "wikigene_name")  
) %>% right_join(time_resolved_rna, by = c("wormbase_gseq" = "WormbaseName"))  
  
## Cache found  
time_resolved_rna <- time_resolved_rna %>% drop_na(wbps_gene_id)  
  
intestine_gene_list <-  
  read_csv("../02_Public_Intesine_RNA/02_output/RWC23_Public_Intestine_RNA_Data.csv")  
  
## Parsed with column specification:  
## cols(  
##   WBGeneID = col_character()  
## )
```

Import wTF3.0 worm transcription factor database

```
wTF3.0 <-  
  read.csv(  
    "../01_ChIPseq_RNAseq_Integration/01_input/TF3-0_namesonly.txt",  
    sep = "\t",  
    header = TRUE  
  ) %>% select(WBGeneID)
```

Filter time-resolved RNA-seq based on intestine expression

```
time_resolved_rna_intestine_df <- time_resolved_rna %>%  
  remove_rownames() %>%  
  arrange(wbps_gene_id) %>%  
  filter(wbps_gene_id %in% intestine_gene_list$WBGeneID) %>%  
  select(-(emb_4cell:emb_471min), -DE, -D, -DX, -Soma, -Male, -AdultSPE9, -gonad, -LENGTH)  
head(time_resolved_rna_intestine_df)  
  
##   wormbase_gseq   wbps_gene_id wikigene_name emb_510min emb_548min emb_587min
```

```
## 1      T13A10.10 WBGene000000005      aat-4      0.1841      0.1632      0.1776
## 2      T11F9.4 WBGene000000007      aat-6      0.1513      0.1482      0.1586
## 3      ZK455.1 WBGene000000040      aco-1      2.3243      1.9498      1.8170
## 4      T25C8.2 WBGene000000067      act-5      15.2874      16.6729      16.8900
## 5      F57F5.4 WBGene000000073      add-2      0.7871      0.7277      0.6445
## 6      D2030.10 WBGene000000084      aex-1      0.1429      0.1878      0.1805
##      emb_626min emb_665min      L1      L2      L3      L4      YA
## 1      0.1677      0.1630 0.0436931 0.2184170 0.265660 0.3224440 0.408817
## 2      0.1554      0.1584 0.1681510 0.2751570 0.349014 0.3264440 0.271406
## 3      1.8299      1.9978 5.0249900 5.9824800 8.917410 2.3600200 4.554760
## 4      16.6729      18.0843 29.1548000 49.1039000 71.569300 29.3725000 34.417200
## 5      0.4962      0.3919 0.5606450 0.3947570 0.335628 0.1979400 0.387552
## 6      0.1832      0.1716 0.1049800 0.0941584 0.122310 0.0752852 0.155656
```

```
time_resolved_rna_intestine_matrix <-
  time_resolved_rna_intestine_df %>%
  select(-wormbase_gseq, -wikigene_name) %>%
  remove_rownames() %>%
  arrange(wbps_gene_id) %>%
  column_to_rownames(var = "wbps_gene_id") %>%
  as.matrix()
head(time_resolved_rna_intestine_matrix)
```

```
##      emb_510min emb_548min emb_587min emb_626min emb_665min
## WBGene000000005 0.1841      0.1632      0.1776      0.1677      0.1630
## WBGene000000007 0.1513      0.1482      0.1586      0.1554      0.1584
## WBGene000000040 2.3243      1.9498      1.8170      1.8299      1.9978
## WBGene000000067 15.2874      16.6729      16.8900      16.6729      18.0843
## WBGene000000073 0.7871      0.7277      0.6445      0.4962      0.3919
## WBGene000000084 0.1429      0.1878      0.1805      0.1832      0.1716
##      L1      L2      L3      L4      YA
## WBGene000000005 0.0436931 0.2184170 0.265660 0.3224440 0.408817
## WBGene000000007 0.1681510 0.2751570 0.349014 0.3264440 0.271406
## WBGene000000040 5.0249900 5.9824800 8.917410 2.3600200 4.554760
## WBGene000000067 29.1548000 49.1039000 71.569300 29.3725000 34.417200
## WBGene000000073 0.5606450 0.3947570 0.335628 0.1979400 0.387552
## WBGene000000084 0.1049800 0.0941584 0.122310 0.0752852 0.155656
```

Perform row normalization

```
time_resolved_rna_intestine_matrix_scaled <-
  t(apply(unlist(time_resolved_rna_intestine_matrix), 1, scale))
colnames(time_resolved_rna_intestine_matrix_scaled) <-
  colnames(time_resolved_rna_intestine_matrix)
```

Store index of relevant genes for row annotations. Use custom function

```
gene_names <-
  c("elt-2", "elt-7", "elt-4", "pqm-1", "mtl-2", "ets-4", "aat-6")
GOI_df <-
  GOI_annotate_heatmap(gene_names, time_resolved_rna_intestine_df$wikigene_name)
GOI_df
```

```
##      name index
## 1 elt-2    159
## 2 elt-7   2115
## 3 elt-4    161
```

```
## 4 pqm-1 457
## 5 mtl-2 350
## 6 ets-4 2459
## 7 aat-6 2
```

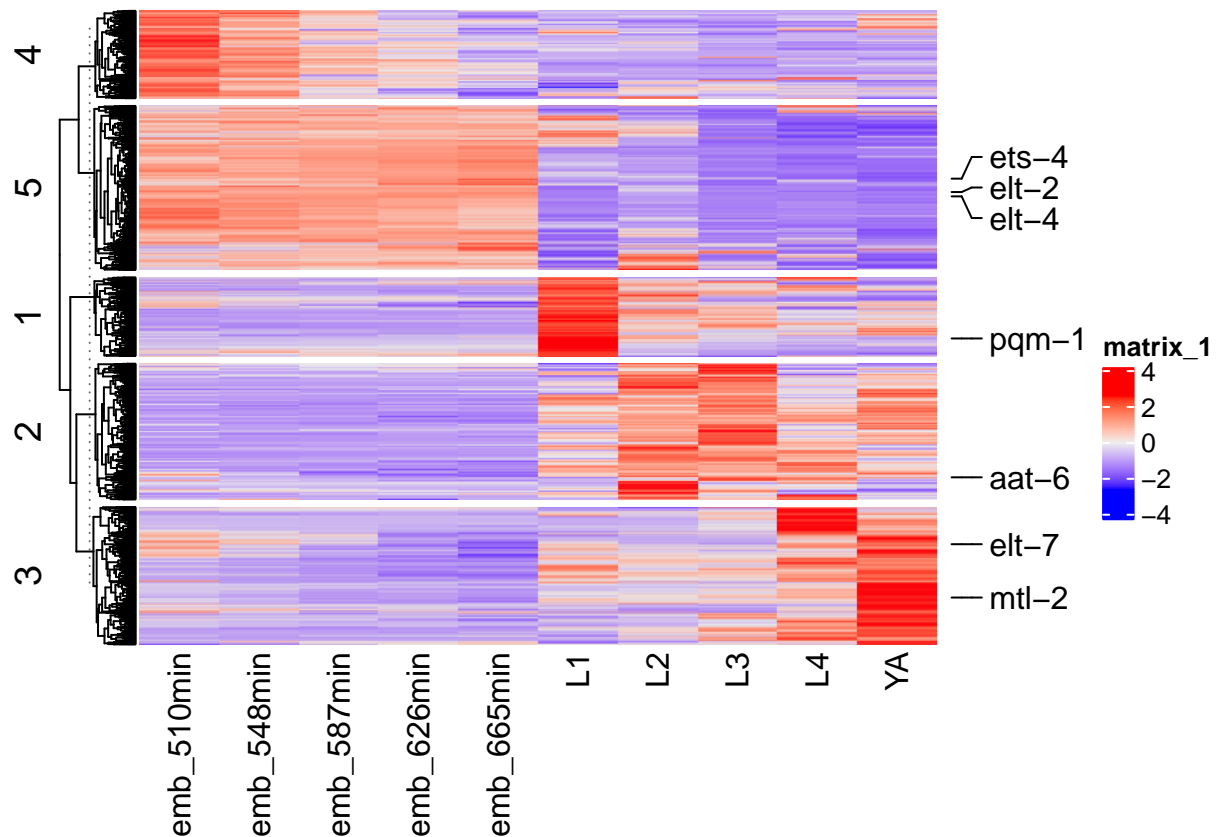
```
time_resolved_rna_intestine_df %>% filter(wikigene_name %in% GOI_df$name)
```

```
## wormbase_gseq wbps_gene_id wikigene_name emb_510min emb_548min emb_587min
## 1 T11F9.4 WBGene000000007 aat-6 0.1513 0.1482 0.1586
## 2 C33D3.1 WBGene00001250 elt-2 1.0771 0.7690 0.7890
## 3 C39B10.6 WBGene00001252 elt-4 0.4558 0.5370 0.5394
## 4 T08G5.10 WBGene00003474 mtl-2 0.0855 0.0869 0.0960
## 5 F40F8.7 WBGene00004096 pqm-1 1.1130 0.9787 0.9506
## 6 C18G1.2 WBGene00015981 elt-7 0.3221 0.3533 0.3047
## 7 F22A3.1 WBGene00017687 ets-4 1.6876 1.9241 2.0529
## emb_626min emb_665min L1 L2 L3 L4 YA
## 1 0.1554 0.1584 0.1681510 0.275157 0.3490140 0.3264440 0.2714060
## 2 0.8991 0.8958 0.2715040 0.531793 0.5176460 0.3497100 0.3836130
## 3 0.5519 0.5908 0.0731321 0.093782 0.0656069 0.0929745 0.0279745
## 4 0.1147 0.1162 2.6928700 3.669790 6.0893500 6.8993700 11.9476000
## 5 0.9540 1.0173 2.4094400 1.813160 1.0956700 1.0476000 0.8723370
## 6 0.1257 0.0675 0.3254970 0.393028 0.2046660 0.4542380 0.2526850
## 7 2.0319 2.0283 0.4815530 1.165750 1.3026800 0.4953370 0.5412020
```

```
Boeck_intestine_RNA <-
```

```
Heatmap(
  time_resolved_rna_intestine_matrix_scaled,
  cluster_columns = FALSE,
  show_row_names = FALSE,
  row_km = 5
) +
  rowAnnotation(foo = anno_mark(GOI_df$index, labels = GOI_df$name))
```

```
Boeck_intestine_RNA
```



```
# pdf(file = "./03_plots/200915_Boeck_RNA_Intestine.pdf", width = 7, height = 7)
# Boeck_intestine_RNA
# dev.off()
```

Filter heatmap for only transcription factors. This is very ugly, fix later.

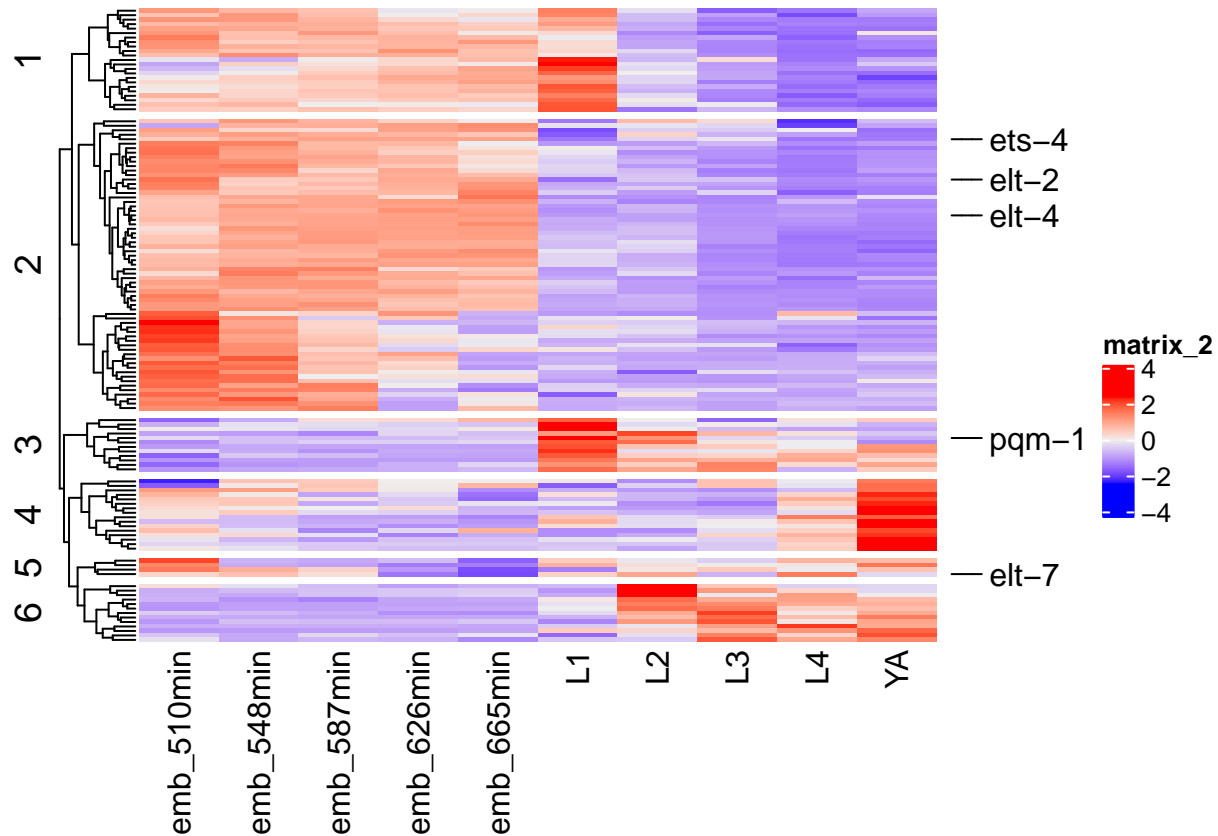
```
time_resolved_rna_intestine_matrix_scaled_TFONLY <-
  matrix_select(time_resolved_rna_intestine_matrix_scaled, wTF3.0$WBGeneID)

tf_GOI_df <-
  GOI_df %>%
  left_join(time_resolved_rna_intestine_df, by = c("name" = "wikigene_name")) %>%
  select(name:wbps_gene_id, -index) %>% filter(wbps_gene_id %in% wTF3.0$WBGeneID)
tf_GOI_df
```

```
##   name wormbase_gseq  wbps_gene_id
## 1 elt-2      C33D3.1 WBGene00001250
## 2 elt-7      C18G1.2 WBGene00015981
## 3 elt-4      C39B10.6 WBGene00001252
## 4 pqm-1      F40F8.7 WBGene00004096
## 5 ets-4      F22A3.1 WBGene00017687
```

```
tf_GOI_df <-
  GOI_annotate_heatmap(
    tf_GOI_df$wbps_gene_id,
    rownames(time_resolved_rna_intestine_matrix_scaled_TFONLY)
  ) %>% full_join(tf_GOI_df, by = c("name" = "wbps_gene_id"))
```

```
Heatmap(
  time_resolved_rna_intestine_matrix_scaled_TFONLY,
  cluster_columns = FALSE,
  show_row_names = FALSE,
  row_split=6
) +
  rowAnnotation(foo = anno_mark(at = tf_GOI_df$index,
                                labels = tf_GOI_df$name.y))
```



Import ELT-2 ChIP-seq binding data

```
chip_df <-
  read_csv(file = "../01_ChIPseq_RNAseq_Integration/01_input/200719_annotatedPeaks.csv")

## Parsed with column specification:
## cols(
##   .default = col_double(),
##   name = col_character(),
##   cluster.description = col_character(),
##   peak = col_character(),
##   WBGeneID = col_character(),
##   feature_strand = col_character(),
##   insideFeature = col_character(),
##   fromOverlappingOrNearest = col_character()
## )
```

```
## See spec(...) for full column specifications.
```

```
head(chip_df)
```

```
## # A tibble: 6 x 32
##   LE_1  LE_2  L1_1  L1_2  L3_1  L3_2  LE_IDR  L1_IDR  L3_IDR  summit_agreement
##   <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1  1.93  1.60  4.25  3.77  4.88  5.01     0       1       1       27.4
## 2  2.11  1.94  4.05  4.46  4.95  5.94     0       1       1       12.4
## 3  1.22  1.53  2.61  2.85  2.45  2.86     0       0       1       137
## 4  1.81  1.42  2.74  3.28  4.18  4.49     0       0       1       2.5
## 5  2.22  2.17  2.24  2.13  4.02  4.10     1       1       1       10
## 6  1.89  2.10  3.43  2.85  3.42  3.53     0       0       1       124.
## # ... with 22 more variables: k4cluster <dbl>, k11cluster <dbl>,
## #   k4weights <dbl>, k11weights <dbl>, LE_nonNormed <dbl>, L1_nonNormed <dbl>,
## #   L3_nonNormed <dbl>, LE_std <dbl>, L1_std <dbl>, L3_std <dbl>, name <chr>,
## #   cluster.description <chr>, variance <dbl>, peak <chr>, WBGeneID <chr>,
## #   start_position <dbl>, end_position <dbl>, feature_strand <chr>,
## #   insideFeature <chr>, distancetoFeature <dbl>, shortestDistance <dbl>,
## #   fromOverlappingOrNearest <chr>
```

Subset ELT-2 ChIP with literature Intestine Expression

Do this earlier in the code to have k4labels stored in the time_resolved_rna dataframe and subsequent subsetting

```
chip_rna_df <- chip_df %>%
  select(name, cluster.description, WBGeneID) %>%
  right_join(time_resolved_rna_intestine_df,
    by = c("WBGeneID" = "wbps_gene_id")) %>%
  replace_na(list("cluster.description" = "Not_Bound", "name" = "Not_Bound"))

chip_rna_df$cluster.description <-
  factor(
    chip_rna_df$cluster.description,
    levels = c(
      "Embryo_Specific",
      "Larval",
      "Increasing",
      "L3_High",
      "Not_Changing",
      "Not_Bound"
    )
  )
```

Subset heatmap based on ELT-2 binding pattern

```
#### Handle duplicate rows created by 1:many gene:peak mapping

# match will return the first index of each non-redundant gene
nr_gene_name_ixs = match(unique(chip_rna_df$wikigene_name), chip_rna_df$wikigene_name)
#length(nr_gene_name_ixs)
#[1] 3286
```

```

chip_rna_df = chip_rna_df[nr_gene_name_ixs,]

chip_rna_matrix <-
  chip_rna_df %>% select(emb_510min:YA) %>% as.matrix()
  #chip_rna_df %>% select(emb_548min,emb_626min,L1,L2,L3,L4) %>% as.matrix()

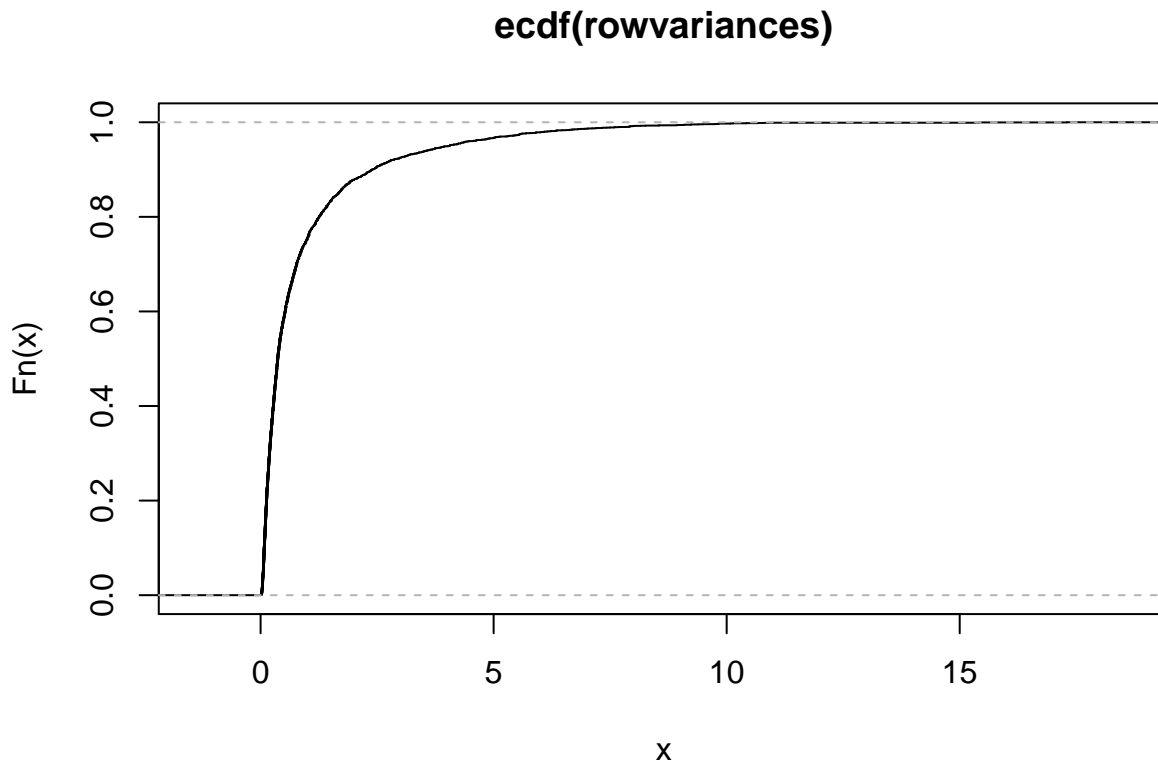
#### Handle 0's and take the log
# 1. Just replace 0's as NAs so we can apply log(). Alternatively, we could do log(x + .01), but there
chip_rna_matrix_na = chip_rna_matrix;
chip_rna_matrix_na[0 == chip_rna_matrix_na] <- NA
# 2. Apply log()
chip_rna_matrix_log = log( chip_rna_matrix_na )
# 3. Do variances row-wise, make sure to set na.rm=T
rowvariances = apply(chip_rna_matrix_log, 1, var, na.rm=T)

range(rowvariances) # no NaNs

## [1] 0.001551605 17.297836019

# 4. Plot distribution of row variances of the log data...
# You can do hist with 10,100,1000 breaks, and there will always be
# a dominant spike all the way on the lowest value. This is because the data elicit no natural lowest b
plot(ecdf(rowvariances)) # no "steps" anywhere, just a smooth curve

```



```

# therefore, we will choose to exclude the lowest 5% of the rows by their variance

chip_rna_matrix = chip_rna_matrix_log
rownames(chip_rna_matrix) <- chip_rna_df$wikigene_name

```



```
chip_rna_matrix_scaled <- row_scale(chip_rna_matrix) # calls base::scale() via RWC23_Functions.R
```

```
for (name in gene_names) {
  index <- which(rownames(chip_rna_matrix_scaled) == name)
  for (i in 1:length(index)) {
    print(c(name, index[i]))
  }
}
```

```
## [1] "elt-2" "159"
## [1] "elt-7" "2092"
## [1] "elt-4" "161"
## [1] "pqm-1" "457"
## [1] "mtl-2" "350"
## [1] "ets-4" "2431"
## [1] "aat-6" "2"
```

```
BoeckRNA_ELT2_chip_Heatmap <-
```

```
function(subsetrows,
        column_title,
        name = "Boeck Time Resolved RNA",
        row_split=chip_rna_df$cluster.description[subsetrows],
        clustering_distance_rows = "euclidean",
        clustering_method_rows = "complete",
        ...)
```

```
{
```

```
ix=which(rownames(chip_rna_matrix_scaled)[subsetrows] %in% gene_names)
chip_GOI_df = data.frame(name=rownames(chip_rna_matrix_scaled)[subsetrows][ix], index=ix)
```

```
BoeckRNA_ELT2_chip <- Heatmap(
  chip_rna_matrix_scaled[subsetrows,],
  name=name,
  row_split = row_split,
  column_title = column_title,
  row_title = NULL,
  cluster_columns = FALSE,
  clustering_distance_rows = clustering_distance_rows,
  clustering_method_rows = clustering_method_rows,
  ...)
```

```
) +
```

```
rowAnnotation(
  ELT2_cluster = chip_rna_df$cluster.description[subsetrows],
  col = list(
    ELT2_cluster = c(
      "Embryo_Specific" = "#7570B3",
      "Larval" = "#1B9E77",
      "Increasing" = "#E7298A",
      "L3_High" = "#D95F02",
      "Not_Changing" = "#505050",
      "Not_Bound" = "yellow"
    )
  ),
)
```

```

border = TRUE
) + rowAnnotation(foo = anno_mark(at = chip_GOI_df$index,
                                labels = chip_GOI_df$name))
BoeckRNA_ELT2_chip
}

library(fpc) # for bootstrapping
library(vcd) # for mosaic

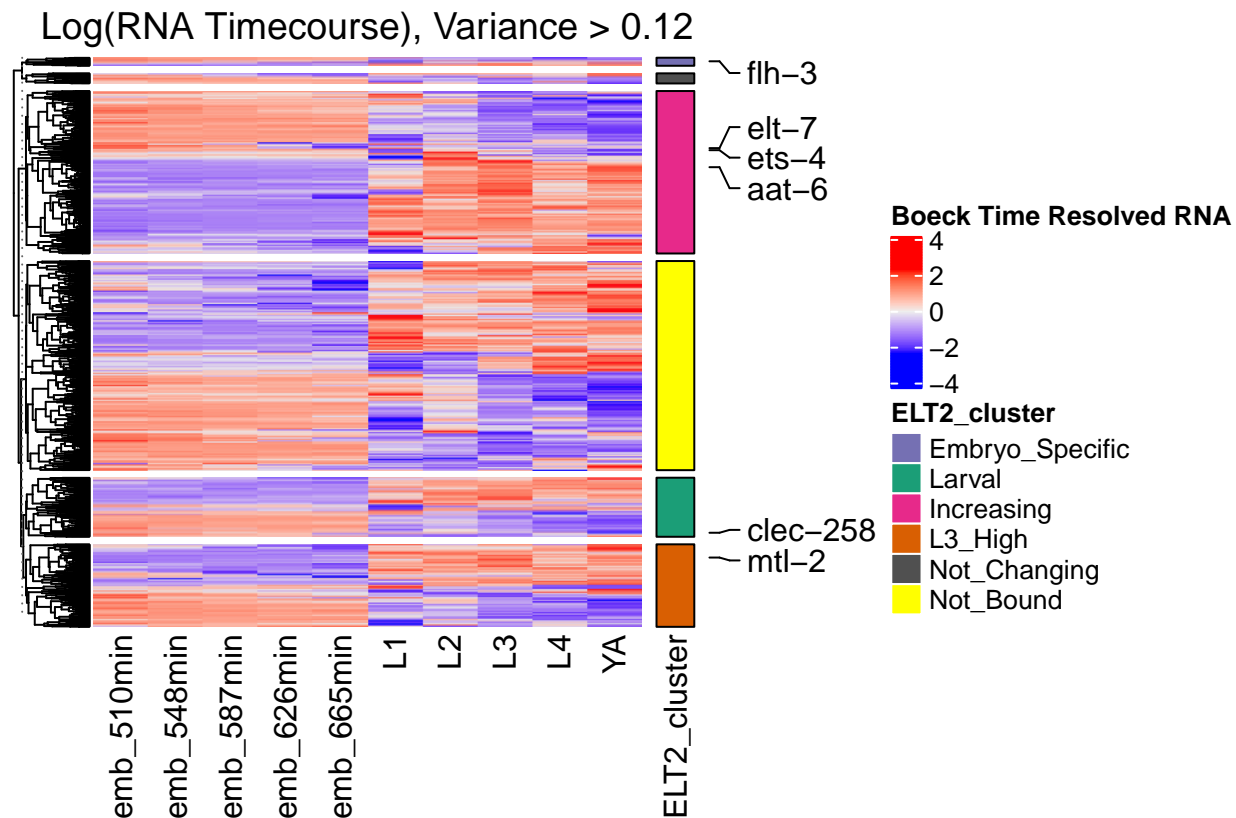
gene_names <-
  c("flh-3", "elt-7", "clec-258", "pqm-1", "mtl-2", "ets-4", "aat-6")

real = apply(chip_rna_matrix_scaled, 1, function(x) { ! any(is.na(x)) }) # NA's introduced by log trans.
embryo_specific = chip_rna_df$cluster.description == "Embryo_Specific"
larval = chip_rna_df$cluster.description == "Larval"
l3_high = chip_rna_df$cluster.description == "L3_High"
increasing = chip_rna_df$cluster.description == "Increasing"

#changing = rowvariances > 0.1355294 # .05 thresh from chipseq

### ALL ###
qthreshold = quantile(rowvariances,.2)
changing = rowvariances > qthreshold
BoeckRNA_ELT2_chip_Heatmap(real & changing, sprintf("Log(RNA Timecourse), Variance > %.2f", qthreshold))

```



```

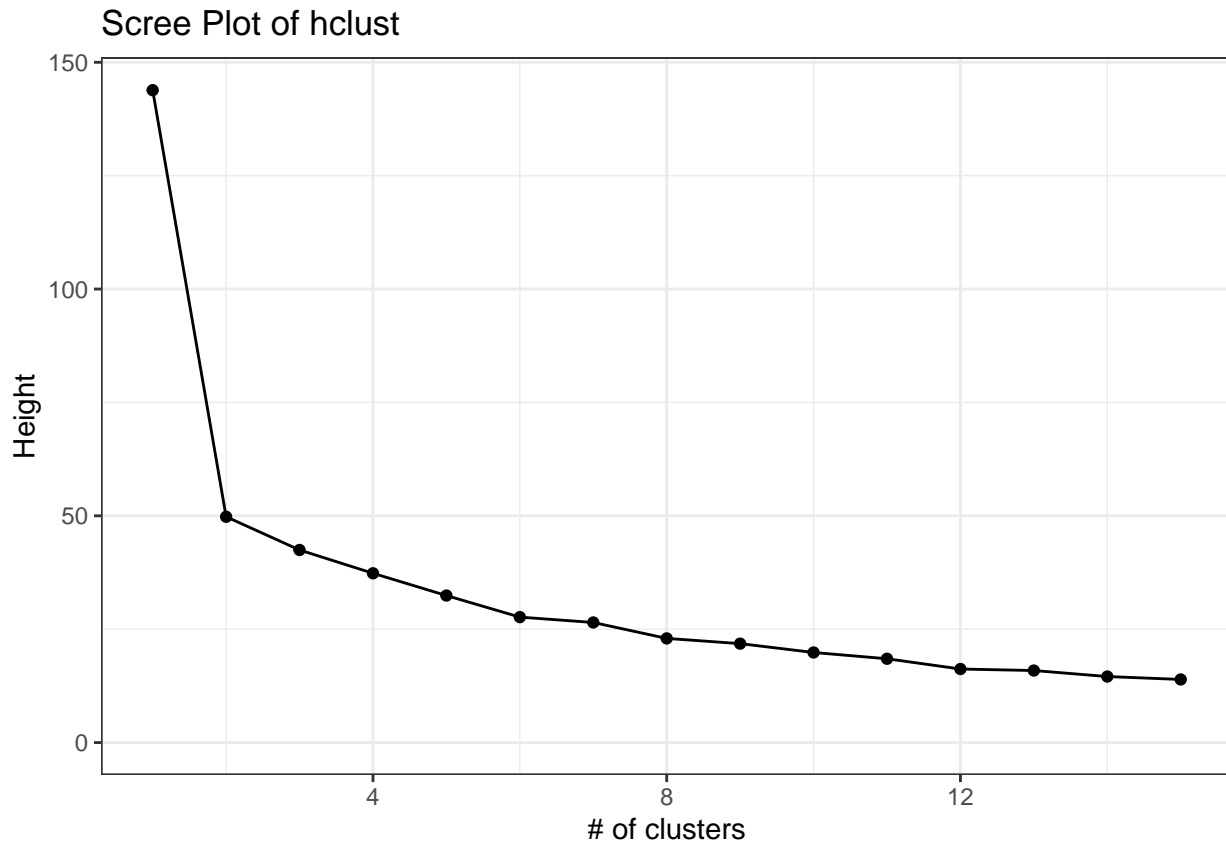
data.whole = chip_rna_matrix_scaled[real&changing,]
agglom='ward.D2'
hc = hclust(dist(data.whole), method =agglom )

```

```
# plot the hc height to get the screeplot
ggplot(NULL, aes(x=length(hc$height):1, y=hc$height)) +
  geom_point() + geom_line() +
  theme_bw() + labs(title="Scree Plot of hclust", x = "# of clusters", y="Height") + scale_x_continuous
```

```
## Warning: Removed 2296 rows containing missing values (geom_point).
```

```
## Warning: Removed 2296 row(s) containing missing values (geom_path).
```



```
nclust=3
# main 2 clusters are stable, 3rd unstable
set.seed(5)
clusterboot( dist(data.whole), clustermethod=disthclustCBI,method=agglom,k=nclust)
```

```
## boot 1
## boot 2
## boot 3
## boot 4
## boot 5
## boot 6
## boot 7
## boot 8
## boot 9
## boot 10
## boot 11
## boot 12
## boot 13
## boot 14
```

boot 15
boot 16
boot 17
boot 18
boot 19
boot 20
boot 21
boot 22
boot 23
boot 24
boot 25
boot 26
boot 27
boot 28
boot 29
boot 30
boot 31
boot 32
boot 33
boot 34
boot 35
boot 36
boot 37
boot 38
boot 39
boot 40
boot 41
boot 42
boot 43
boot 44
boot 45
boot 46
boot 47
boot 48
boot 49
boot 50
boot 51
boot 52
boot 53
boot 54
boot 55
boot 56
boot 57
boot 58
boot 59
boot 60
boot 61
boot 62
boot 63
boot 64
boot 65
boot 66
boot 67
boot 68

```

## boot 69
## boot 70
## boot 71
## boot 72
## boot 73
## boot 74
## boot 75
## boot 76
## boot 77
## boot 78
## boot 79
## boot 80
## boot 81
## boot 82
## boot 83
## boot 84
## boot 85
## boot 86
## boot 87
## boot 88
## boot 89
## boot 90
## boot 91
## boot 92
## boot 93
## boot 94
## boot 95
## boot 96
## boot 97
## boot 98
## boot 99
## boot 100

## * Cluster stability assessment *
## Cluster method: hclust
## Full clustering results are given as parameter result
## of the clusterboot object, which also provides further statistics
## of the resampling results.
## Number of resampling runs: 100
##
## Number of clusters found in data: 3
##
## Clusterwise Jaccard bootstrap (omitting multiple points) mean:
## [1] 0.4438836 0.7915816 0.8700212
## dissolved:
## [1] 64 0 0
## recovered:
## [1] 2 78 93

subsetrows=real&changing
ix=which(rownames(chip_rna_matrix_scaled)[subsetrows] %in% gene_names)
chip_GOI_df = data.frame(name=rownames(chip_rna_matrix_scaled)[subsetrows][ix], index=ix)

Heatmap(data.whole,
        column_title=sprintf("Log(RNA Timecourse), Variance > %.2f", qthreshold),

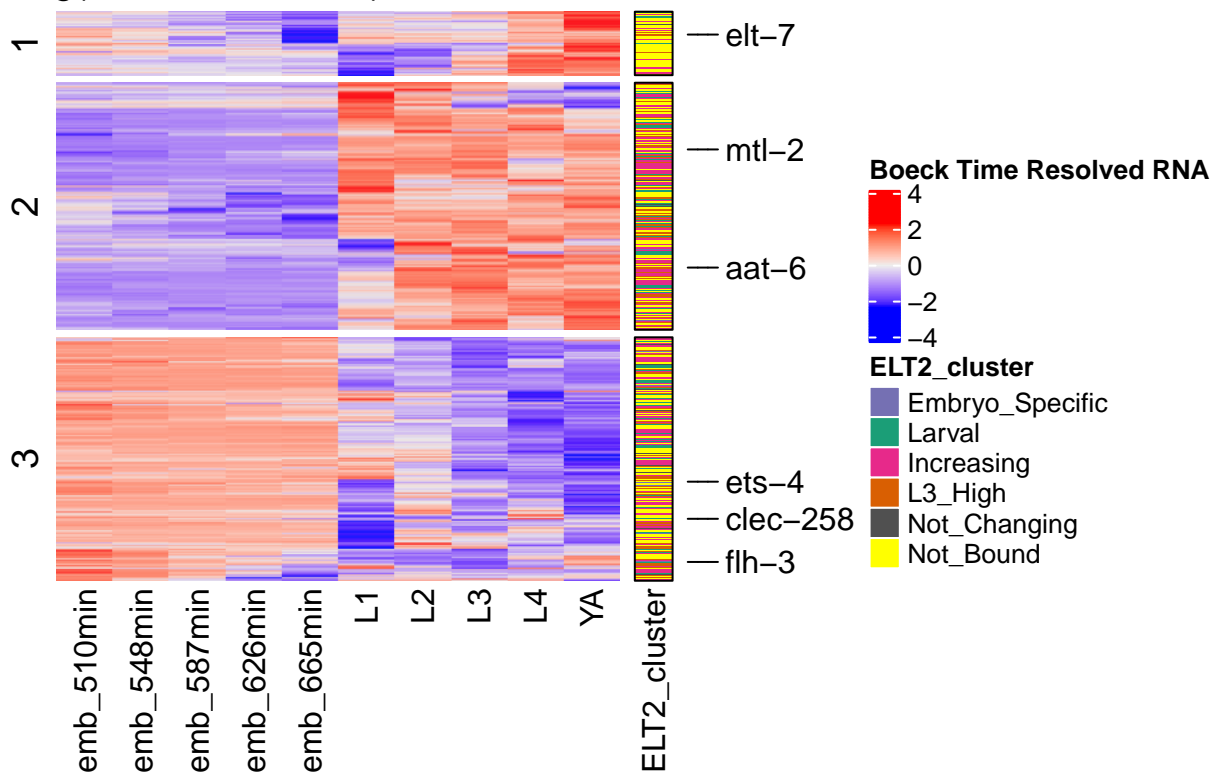
```

```

name = "Boeck Time Resolved RNA",
cluster_columns = F,
row_order=hc$order,row_split=cutree(hc, nclust),show_row_names = F) +
rowAnnotation(
  ELT2_cluster = chip_rna_df$cluster.description[real&changing],
  col = list(
    ELT2_cluster = c(
      "Embryo_Specific" = "#7570B3",
      "Larval" = "#1B9E77",
      "Increasing" = "#E7298A",
      "L3_High" = "#D95F02",
      "Not_Changing" = "#505050",
      "Not_Bound" = "yellow"
    )
  ),
  border = TRUE
)+ rowAnnotation(foo = anno_mark(at = chip_GOI_df$index,
                                labels = chip_GOI_df$name))

```

Log(RNA Timecourse), Variance > 0.12

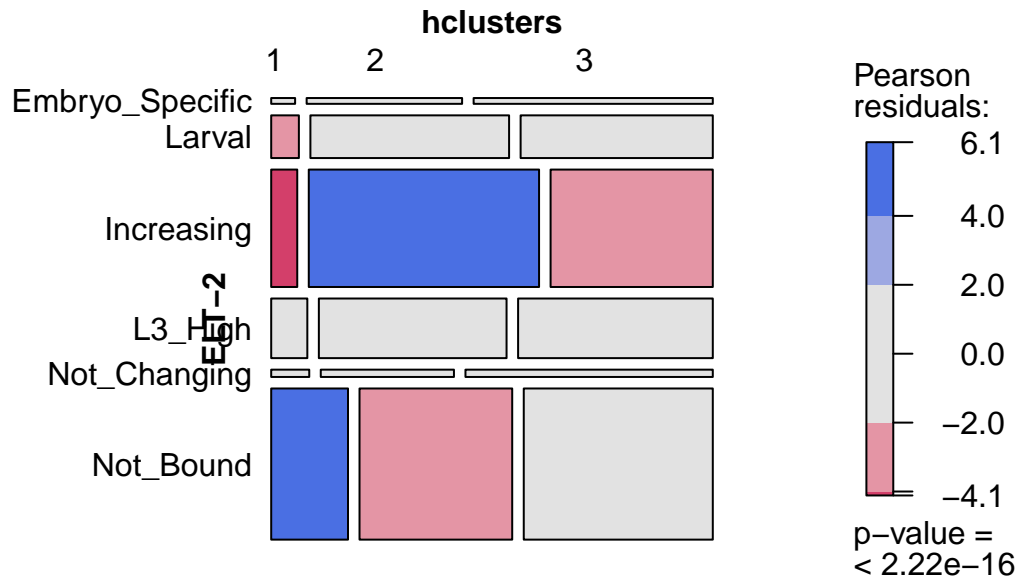


```

hclusters = cutree(hc, nclust)
tab=table(hclusters, chip_rna_df$cluster.description[real&changing])
dimnames(tab) <- list("hclusters"=1:nclust,
                      "ELT-2"=levels(chip_rna_df$cluster.description[real&changing]))
mosaic(t(tab),shade=T,
       just_labels="right",
       rot_labels=c(0,0,0,0),
       offset_label=c(0,0,0,-.5),
       main = "Time Resolved RNA associates with some \nELT-2 peak clusters")

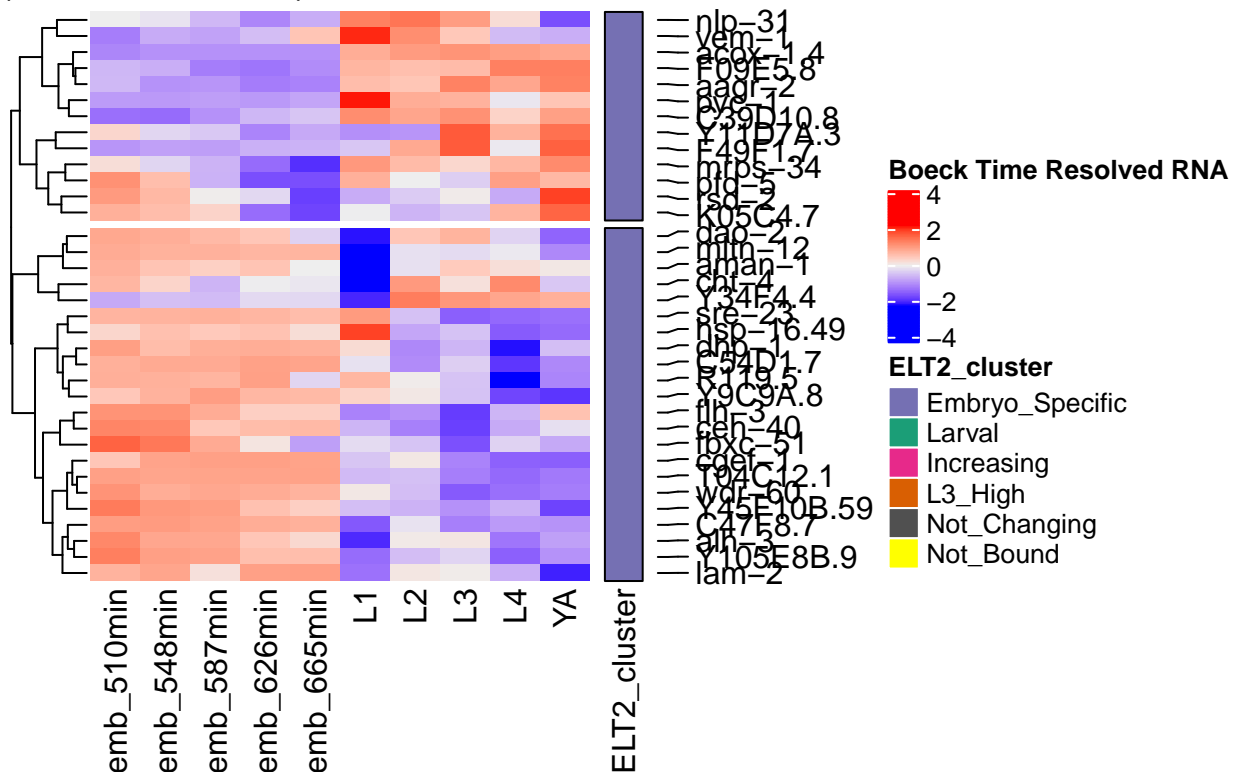
```

Time Resolved RNA associates with some ELT-2 peak clusters



```
gene_names=rownames(chip_rna_matrix_scaled)[real & changing & embryo_specific]
(BoeckRNA_ELT2_chip_Heatmap(real & changing & embryo_specific, "Log(RNA Timecourse), No threshold on variance")
```

(RNA Timecourse), No threshold on variance

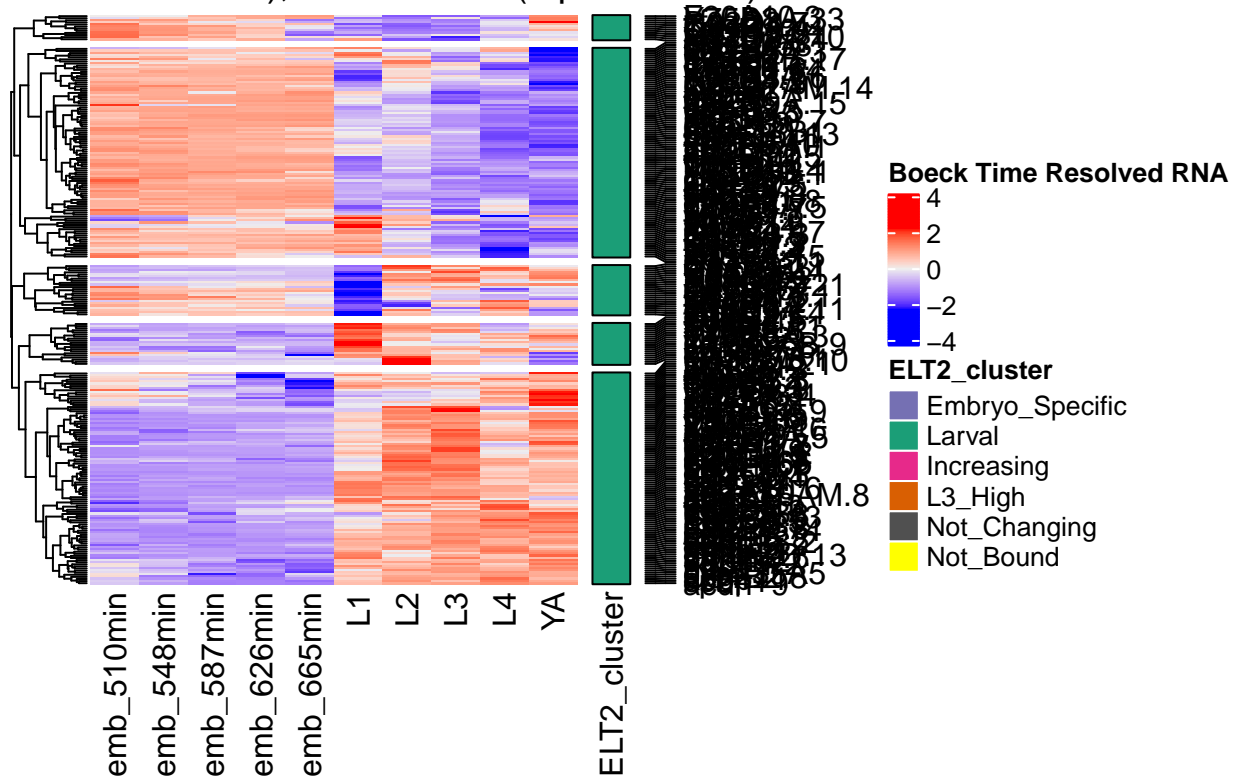


```
pdf(file="embryo.pdf",height=10.5,width=10.5)
print(emb)
dev.off()
```

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

```
gene_names=rownames(chip_rna_matrix_scaled)[real & changing & larval]
(BoeckRNA_ELT2_chip_Heatmap(real & changing & larval, "Log(RNA Timecourse), variance > 1 (top 20% data)
```

RNA Timecourse), variance > 1 (top 20% data)

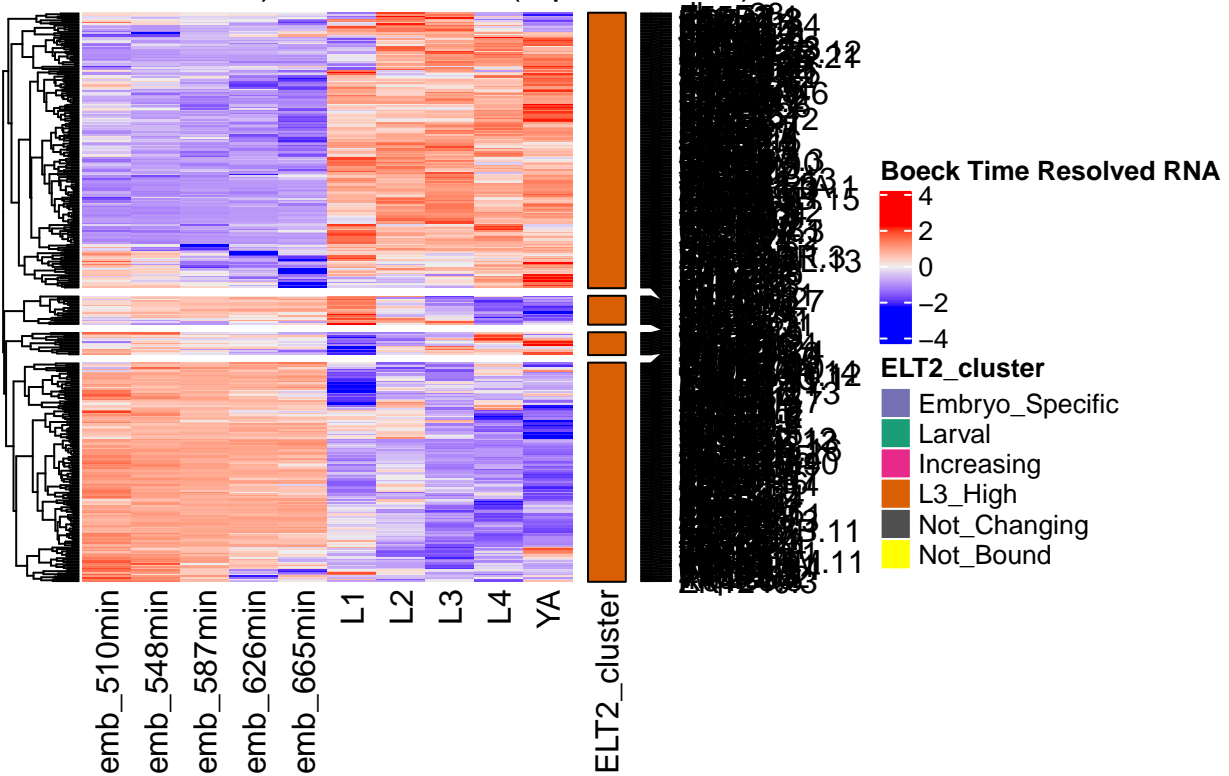


```
pdf(file="larval.pdf",height=14,width=14)
print(larv)
dev.off()
```

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

```
gene_names=rownames(chip_rna_matrix_scaled)[real & changing & l3_high]
(BoeckRNA_ELT2_chip_Heatmap(real & changing & l3_high, "Log(RNA Timecourse), variance > 1 (top 20% data)
```


RNA Timecourse), variance > 1 (top 20% data)

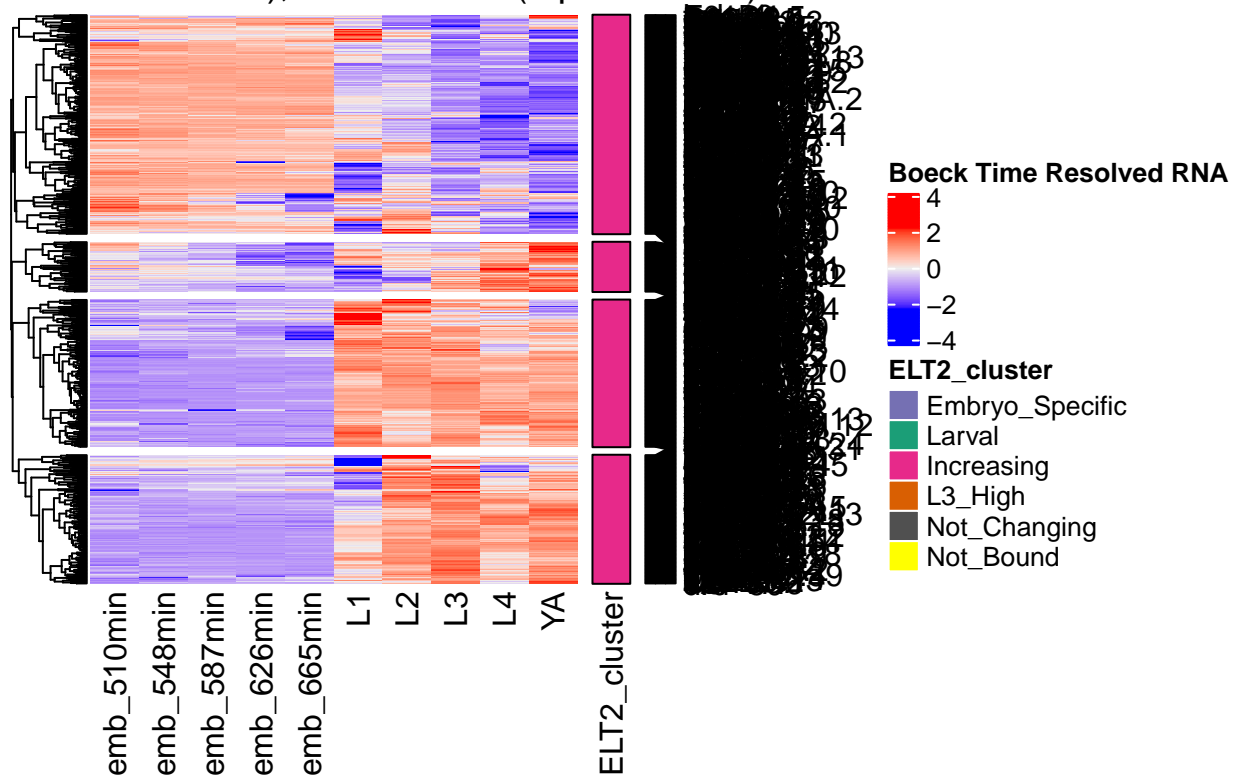


```
pdf(file="L3.pdf",height=14,width=14)
print(L3plot)
dev.off()
```

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

```
gene_names=rownames(chip_rna_matrix_scaled)[real & changing & increasing]
(BoeckRNA_EL2_chip_Heatmap(real & changing & increasing, "Log(RNA Timecourse), variance > 1 (top 20% d
```

RNA Timecourse), variance > 1 (top 20% data)



```
pdf(file="increasing.pdf",height=21,width=21)
print(inc)
dev.off()
```

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