## RWC23\_ELT2\_ChIP\_Boeck\_Time\_Resolved\_RNA

Note: Ensure BioConductor is version 3.10 or above

```
Install libraries
```

```
# fill this in
Note: you must load biomaRt before loading tidyverse
Load libraries
library(biomaRt)
library(tidyverse)
## Warning: package 'tidyverse' was built under R version 4.0.2
## -- Attaching packages ------ tidyverse 1.3.0 --
## v ggplot2 3.3.2
                   v purrr
                              0.3.4
## v tibble 3.0.2 v dplyr 1.0.0
## v tidyr 1.1.1 v stringr 1.4.0
## v readr
          1.4.0 v forcats 0.5.0
## Warning: package 'tidyr' was built under R version 4.0.2
## Warning: package 'readr' was built under R version 4.0.2
## Warning: package 'forcats' was built under R version 4.0.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                 masks stats::lag()
## x dplyr::select() masks biomaRt::select()
library(ComplexHeatmap)
## Warning: package 'ComplexHeatmap' was built under R version 4.0.2
## Loading required package: grid
## ComplexHeatmap version 2.4.3
## 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.
##
## This message can be suppressed by:
    suppressPackageStartupMessages(library(ComplexHeatmap))
## =============
```

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 <-</pre>
 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(</pre>
 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"))
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")
##
## -- Column specification -----
## cols(
    WBGeneID = col_character()
##
## )
```

### Import wTF3.0 worm transcription factor database

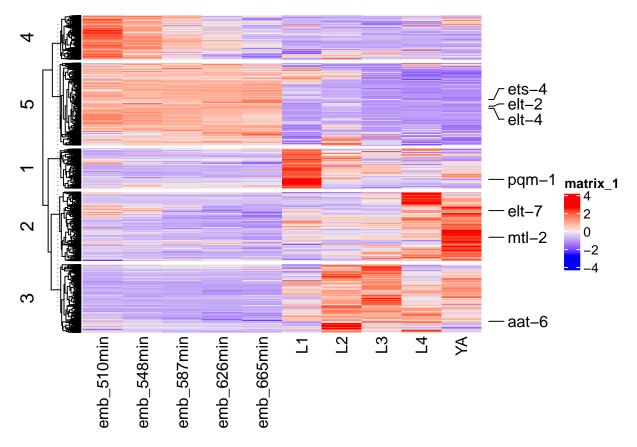
```
wTF3.0 <-
  read.csv(
    "../01_ChIPseq_RNAseq_Integration/01_input/TF3-0_namesonly.txt",
    sep = "\t",
    header = TRUE
) %>% select(WBGeneID)
```

#### 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)
##
                     wbps_gene_id wikigene_name emb_510min emb_548min emb_587min
     wormbase_gseq
## 1
         T13A10.10 WBGene00000005
                                                     0.1841
                                                                0.1632
                                                                            0.1776
                                           aat-4
## 2
           T11F9.4 WBGene00000007
                                           aat-6
                                                     0.1513
                                                                0.1482
                                                                            0.1586
## 3
           ZK455.1 WBGene00000040
                                                     2.3243
                                                                1.9498
                                                                            1.8170
                                           aco-1
## 4
           T25C8.2 WBGene00000067
                                           act-5
                                                    15.2874
                                                               16.6729
                                                                           16.8900
## 5
           F57F5.4 WBGene00000073
                                           add-2
                                                     0.7871
                                                                0.7277
                                                                            0.6445
          D2030.10 WBGene00000084
                                                                0.1878
## 6
                                           aex-1
                                                     0.1429
                                                                            0.1805
     emb 626min emb 665min
##
                                               L2
                                   L1
                                                         L3
                                                                    1.4
                                                                               YΑ
## 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
                                       0.0941584 0.122310 0.0752852
## 6
         0.1832
                    0.1716 0.1049800
                                                                        0.155656
time_resolved_rna_intestine_matrix <-</pre>
  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
##
## WBGene0000005
                      0.1841
                                 0.1632
                                             0.1776
                                                        0.1677
                                                                   0.1630
## WBGene0000007
                      0.1513
                                 0.1482
                                             0.1586
                                                        0.1554
                                                                   0.1584
## WBGene00000040
                      2.3243
                                 1.9498
                                             1.8170
                                                        1.8299
                                                                   1.9978
## WBGene0000067
                     15.2874
                                16.6729
                                            16.8900
                                                       16.6729
                                                                   18.0843
## WBGene00000073
                      0.7871
                                 0.7277
                                             0.6445
                                                        0.4962
                                                                   0.3919
## WBGene00000084
                      0.1429
                                 0.1878
                                             0.1805
                                                        0.1832
                                                                   0.1716
                                                L3
                                                           L4
                          L1
                                      L2
                                                                     YΑ
## WBGene0000005
                   0.0436931
                              0.2184170
                                         0.265660
                                                    0.3224440
                                                               0.408817
## WBGene0000007
                   0.1681510
                              0.2751570
                                         0.349014
                                                    0.3264440
                                                               0.271406
## WBGene00000040
                   5.0249900
                              5.9824800 8.917410
                                                    2.3600200
## WBGene00000067 29.1548000 49.1039000 71.569300 29.3725000 34.417200
## WBGene00000073
                  0.5606450
                                         0.335628
                              0.3947570
                                                    0.1979400
                                                               0.387552
## WBGene00000084
                  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) <-</pre>
  colnames(time_resolved_rna_intestine_matrix)
```

Store index of relavent 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)
                     wbps_gene_id wikigene_name emb_510min emb_548min emb_587min
##
     wormbase_gseq
           T11F9.4 WBGene00000007
## 1
                                                     0.1513
                                                                           0.1586
                                          aat-6
                                                                0.1482
## 2
           C33D3.1 WBGene00001250
                                                     1.0771
                                                                0.7690
                                                                           0.7890
                                          elt-2
## 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
                                                                           0.9506
                                          pqm-1
                                                     1.1130
                                                                0.9787
           C18G1.2 WBGene00015981
## 6
                                          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
                                                                           YΑ
                    0.1584 0.1681510 0.275157 0.3490140 0.3264440 0.2714060
## 1
         0.1554
## 2
         0.8991
                    0.8958 0.2715040 0.531793 0.5176460 0.3497100 0.3836130
## 3
                    0.5908 0.0731321 0.093782 0.0656069 0.0929745 0.0279745
         0.5519
## 4
                    0.1162 2.6928700 3.669790 6.0893500 6.8993700 11.9476000
         0.1147
## 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 <-</pre>
 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()
```

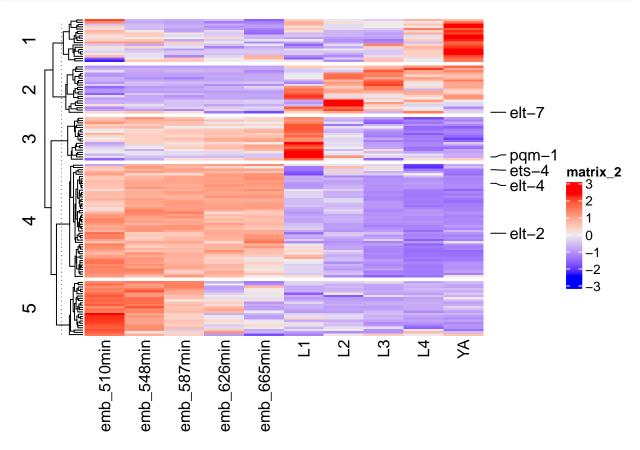
Filter heatmap for only transcription factors. This is very ugly, fix later.

```
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
                 C18G1.2 WBGene00015981
## 2 elt-7
                C39B10.6 WBGene00001252
## 3 elt-4
                 F40F8.7 WBGene00004096
## 4 pqm-1
## 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"))
```



## Import ELT-2 ChIP-seq binding data

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

```
## )
## i Use `spec()` for the 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
                                                                         27.4
## 2 2.11 1.94 4.05 4.46 4.95 5.94
                                             0
                                                                         12.4
                                                    1
## 3 1.22 1.53 2.61
                       2.85
                             2.45
                                   2.86
                                             0
                                                    0
                                                                        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
                                                                         10
## 6 1.89 2.10 3.43 2.85 3.42 3.53
                                             0
                                                    0
                                                                        124.
                                                           1
## # ... 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$cluster.description <-</pre>
  factor(
    chip_rna_df$cluster.description,
    levels = c(
      "Embryo_Specific",
      "Larval",
      "Increasing",
      "L3_High",
      "Not_Changing",
      "Not_Bound"
    )
  )
```

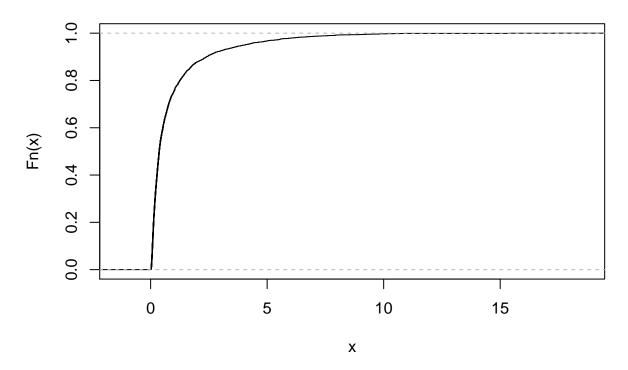
## Subset heatmap based on ELT-2 binding pattern

```
#### 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 <-</pre>
  chip_rna_df %>% select(emb_510min:YA) %>% as.matrix()
  #chip_rna_df %>% select(emb_548min,emb_626min,L1,L2,L3,L4) %>% as.matrix()
#### Handle O'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</pre>
# 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
```

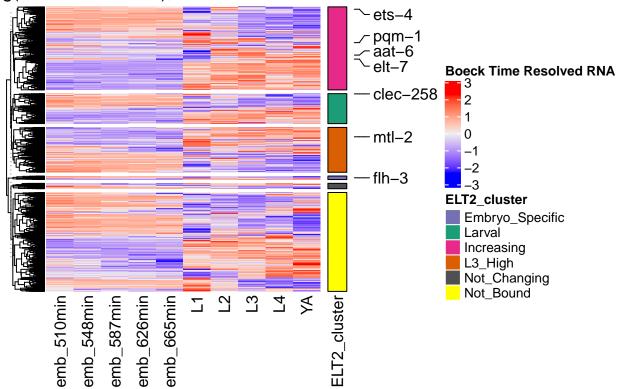
## ecdf(rowvariances)



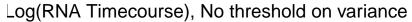
```
# therefore, we will choose to exclude the lowest 5% of the rows by their variance
changing = rowvariances > 1 # quantile(rowvariances, .1) # or .05
chip_rna_matrix = chip_rna_matrix_log
rownames(chip_rna_matrix) <- chip_rna_df$wikigene_name</pre>
chip_rna_matrix_scaled <- row_scale(chip_rna_matrix) # calls base::scale() via RWC23_Functions.R
for (name in gene_names) {
  index <- which(rownames(chip_rna_matrix_scaled) == name)</pre>
  for (i in 1:length(index)) {
    print(c(name, index[i]))
  }
}
## [1] "elt-2" "2013"
## [1] "elt-7" "1085"
## [1] "elt-4" "2014"
## [1] "pqm-1" "414"
## [1] "mtl-2" "1320"
## [1] "ets-4" "1523"
## [1] "aat-6" "1239"
BoeckRNA_ELT2_chip_Heatmap <- function(subsetrows, label) {</pre>
  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 )
```

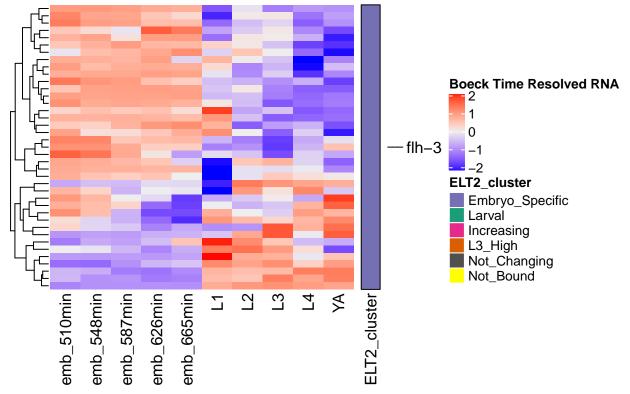
```
BoeckRNA_ELT2_chip <- Heatmap(</pre>
  chip_rna_matrix_scaled[subsetrows,],
  name = "Boeck Time Resolved RNA",
 row_split = chip_rna_df$cluster.description[subsetrows],
  column_title = label,
 row_title = NULL,
 cluster_columns = FALSE
 rowAnnotation(
   ELT2_cluster = chip_rna_df$cluster.description[subsetrows],
   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
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"
#changing = rowvariances > 0.1355294 # .05 thresh from chipseq
### ALL ###
changing = rowvariances > 0
BoeckRNA_ELT2_chip_Heatmap(real & changing, "Log(RNA Timecourse), No threshold on variance")
```

# g(RNA Timecourse), No threshold on variance



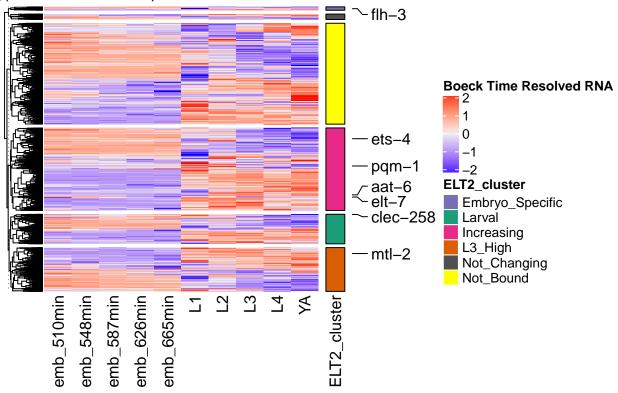
BoeckRNA\_ELT2\_chip\_Heatmap(real & changing & embryo\_specific, "Log(RNA Timecourse), No threshold on var



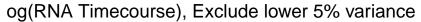


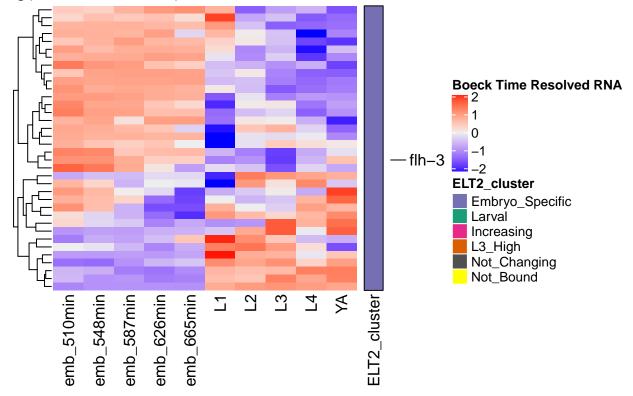
```
### Exclude the least changing, lower 5% of variance ###
changing = rowvariances > quantile(rowvariances, .05)
BoeckRNA_ELT2_chip_Heatmap(real & changing, "Log(RNA Timecourse), Exclude lower 5% variance")
```

# g(RNA Timecourse), Exclude lower 5% variance

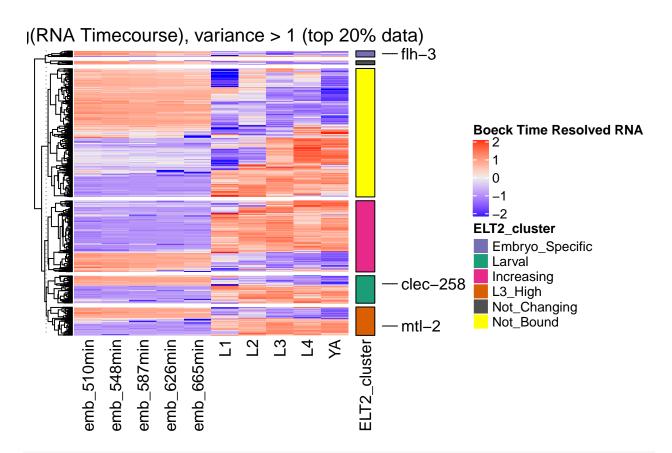


BoeckRNA\_ELT2\_chip\_Heatmap(real & changing & embryo\_specific, "Log(RNA Timecourse), Exclude lower 5% va



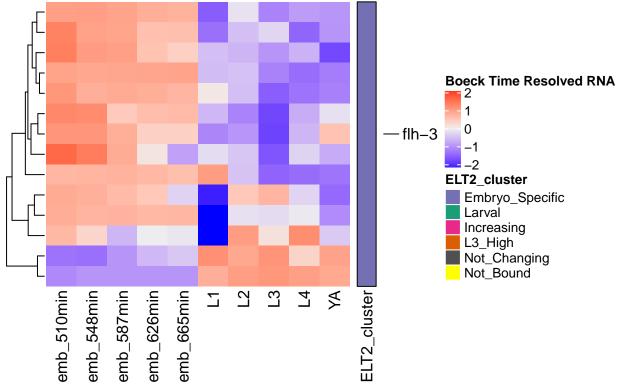


```
### Include only rows with a variance greater than 1 ###
changing = rowvariances > 1
BoeckRNA_ELT2_chip_Heatmap(real & changing, "Log(RNA Timecourse), variance > 1 (top 20% data)")
```

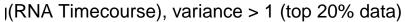


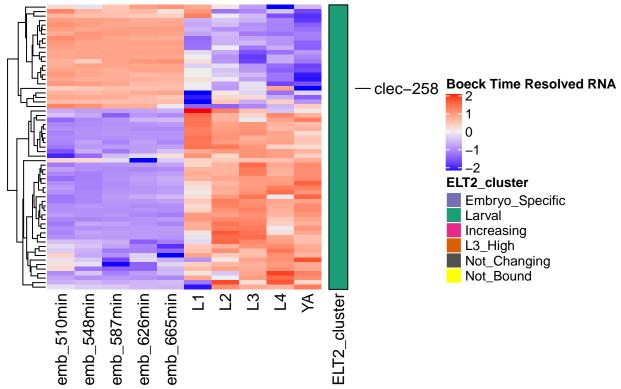
BoeckRNA\_ELT2\_chip\_Heatmap(real & changing & embryo\_specific, "Log(RNA Timecourse), variance > 1 (top 2





BoeckRNA\_ELT2\_chip\_Heatmap(real & changing & larval, "Log(RNA Timecourse), variance > 1 (top 20% data)"





```
# pdf(file = "./03_plots/201008_Boeck_RNA_ELT2ChIP_Patterns_Subset.pdf",
# width = 7, height = 7)
# BoeckRNA_ELT2_chip
# dev.off()
```