irCLIPv2 dataset from 4 RBPs and 3 RDAPs in HEK293T and HepG2 from three RNP subzones

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This is the pipeline used to analyze the irCLIPv2 datasets for 4 RBPs and 3 RDAPs from three different gel sections (RNP subzone #1-3). The experiment was performed in HEK293T. Mapping, counting, and Differential enrichment analysis between RNP subzones and no-UV samples was performed using a custom snakemake pipeline. Please see this Github link for more information.

1. Load the data

After DEWseq analysis against noUV using a custom snakemake pipeline, we first characterized the significant regions.

```
#Load the libraries
library (RIdeogram)
library (RColorBrewer)
library (scales)
library (ggplot2)
library (tidyverse)
library (plyranges)
library (Repitools)
library (GenomicRanges)
library (grid)
library (gridExtra)
library (ComplexHeatmap)
library (ChIPpeakAnno)
library (eulerr)
library (ggridges)
library (AnnotationDbi)
library (gintools)
library (EnrichedHeatmap)
library (rtracklayer)
library (caTools)
library (memes)
library (universalmotif)
library (magrittr)
library (BSgenome. Hsapiens. UCSC. hg38)
library(paletteer)
library (gUtils)
library (circlize)
library (multcomp)
```

```
\label{eq:hnrnpc} \mbox{HNRNPC} \longleftarrow \mbox{read.delim} (\mbox{"$\sim$/Documents/Postdoc/PD\_Projects/3\_irCLIP\_RNP/Seq/1\_Subzones/1\_Regions} \mbox{ $DEW-PD\_Projects/3\_irCLIP\_RNP/Seq/1\_Subzones/1\_Regions} \mbox{ $DEW-PD\_Projects/3\_irCLIP\_RNP/Seq/1\_Subzones/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/1\_Regions/
         Seq/deseq2_norm/2_Subzone_sum/3_pvalue_0.05_fc1/HNRNPC_DEWSeq_sign_regions_sum_categorization
          .txt", header = TRUE)
HNRNPC$rbp <- "HNRNPC"
colnames(HNRNPC)[5] <- "region_strand"
HNRNPC <- subset (HNRNPC, chromosome != "chrM")
HNRNPM <- read.delim("~/Documents/Postdoc/PD Projects/3 irCLIP-RNP/Seg/1 Subzones/1 Regions DEW-
        Seq/deseq2\_norm/2\_Subzone\_sum/3\_pvalue\_0.05\_fc1/\underline{HNRNPM\_DEWSeq\_sign\_regions\_sum\_categorization}
         .txt", header = TRUE)
HNRNPM$rbp <- "HNRNPM"
colnames (HNRNPM) [5] <- "region strand"
HNRNPM <- subset (HNRNPM, chromosome != "chrM")
HNRNPU <- read.delim("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/1_Regions_DEW-
        Seq/deseq2_norm/2_Subzone_sum/3_pvalue_0.05_fc1/HNRNPU_DEWSeq_sign_regions_sum_categorization
         .txt", header = TRUE)
HNRNPU$rbp <- "HNRNPU"
colnames (HNRNPU) [5] <- "region_strand"
HNRNPU <- subset (HNRNPU, chromosome != "chrM")
ELAVL1 <- read.delim("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/1_Regions_DEW-
        Seq/deseq2 norm/2 Subzone sum/3 pvalue 0.05 fc1/ELAVL1 DEWSeq sign regions sum categorization
          txt", header = TRUE)
ELAVL1$rbp <- "ELAVL1"
colnames(ELAVL1)[5] <- "region_strand"
ELAVL1 <- subset (ELAVL1, chromosome != "chrM")
Seq/deseq2 norm/2 Subzone sum/3 pvalue 0.05 fc1/HNRNPFH DEWSeq sign regions sum
         categorization.txt", header = TRUE)
HNRNPFH$rbp <- "HNRNPFH"
colnames(HNRNPFH)[5] <- "region_strand"
HNRNPFH <- subset (HNRNPFH, chromosome != "chrM")
KHSRP <- read.delim("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/1_Regions_DEW-
        Seq/deseq2_norm/2_Subzone_sum/3_pvalue_0.05_fc1/KHSRP_DEWSeq_sign_regions_sum_categorization.
        txt", header = TRUE)
KHSRP$rbp <- "KHSRP"
colnames (KHSRP) [5] <- "region_strand"
KHSRP <- subset (KHSRP, chromosome != "chrM")
THOC4 <-- read.delim("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/1_Regions_DEW-
        Seq/deseq2\_norm/2\_Subzone\_sum/3\_pvalue\_0.05\_fc1/THOC4\_DEWSeq\_sign\_regions\_sum\_categorization.
        txt", header = TRUE)
THOC4$rbp <- "THOC4"
colnames (THOC4) [5] <- "region strand"
THOC4 <- subset (THOC4, chromosome != "chrM")
```

2. Region characterization

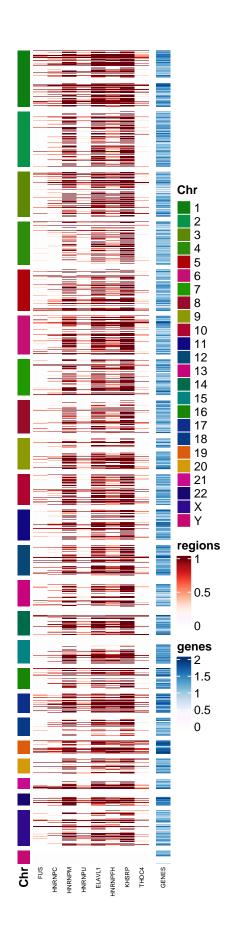
Heatmap of binding density throughout the genome

The first visualization is the distribution of the significant regions across the genome.

```
#Get gene density and human karyotype from the package RIdeogram
data(gene_density, package="RIdeogram")
gene_density2 <- gene_density[,-4]
data(human_karyotype, package="RIdeogram")
human_karyotype$Chr <- paste0("chr", human_karyotype$Chr, sep = "")
gd_function <- function (br, gd_all, rbp) {
```

```
rbp.gd <- br %% dplyr::select(chromosome,region_begin,region_end)
    rbp.gd <- unique(rbp.gd)
    colnames(rbp.gd) <- c("Chr", "Start", "End")
rbp.gd$Chr <- gsub("chr", "", rbp.gd$Chr)</pre>
    rbp.gd2 <- bedtoolsr::bt.intersect(gd_all, rbp.gd, C = T)
    colnames(rbp.gd2) <- c("Chr", "Start", "End", rbp)
    return (rbp.gd2)
FUS.gd <- gd_function(FUS, gene_density2, "FUS")
HNRNPC.gd <- gd_function(HNRNPC, gene_density2, "HNRNPC")
HNRNPM.gd <- gd_function(HNRNPM, gene_density2, "HNRNPM")
HNRNPU.gd <- gd_function(HNRNPU, gene_density2, "HNRNPU")
ELAVL1.gd <- gd_function(ELAVL1, gene_density2, "ELAVL1")
HNRNPFH.gd <- gd_function(HNRNPFH, gene_density2, "HNRNPFH")
KHSRP.gd <- gd_function(KHSRP, gene_density2, "KHSRP")
THOC4.gd <- gd_function(THOC4, gene_density2, "THOC4")
#Combine the region densities
all.rbp \leftarrow cbind(FUS.gd[4]\ ,\ HNRNPC.gd[4]\ ,\ HNRNPM.gd[4]\ ,\ HNRNPU.gd[4]\ ,\ ELAVL1.gd[4]\ ,\ HNRNPFH.gd[4]\ ,\ HN
         [4], KHSRP.gd[4], THOC4.gd[4])
rownames(all.rbp) <- paste0("row_", seq(nrow(all.rbp)))
all.rbp <- log10(all.rbp)
all.rbp[all.rbp = Inf] < 0
#Create annotation for the Heatmap
annotation <- FUS.gd[1]
annotation $Chr <- factor (annotation $Chr, levels = c(1:22,"X","Y"))
rownames(annotation) <- row.names(all.rbp)</pre>
my. breaks <- c(seq(0, 1, by=0.0125))
my.colors <- c(colorRampPalette(colors = c("#fdfdfd", "#fff3fd", brewer.pal(9, "Reds")))(length(
        my.breaks)))
#Generate Heatmap
p <- pheatmap(
   mat
                                       = all.rbp,
    color = my.colors,
    breaks = my. breaks,
    name = "regions",
    annotation row = annotation,
    show_colnames
                                     = TRUE.
    show_rownames
                                      = FALSE,
    drop levels
                                       = TRUE.
    fontsize
                                       = 5,
                                       = FALSE,
    cluster\_rows
                                      = FALSE,
    cluster_cols
    scale
                                       = "none"
    gaps\_row = head(as.numeric(cumsum(table(annotation\$Chr))), -1)
#Add the density of the genes to the generated heatmap
gene_anno <- GFFex(input = "~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/2_
        Characterization/gencode.v39.annotation.gff3.gz", karyotype = "/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/2_Characterization/human_karyotype.txt",
        feature = "gene", window = 1000000)
colnames(gene_anno) <- c("Chr", "Start", "End", "GENES")
gene_anno$GENES <- log10(gene_anno$GENES)
gene_annoGENES[is.infinite(gene_anno<math>GENES)] < 0
gene_anno$Chr <- gsub("chr", "", gene_anno$Chr)
gene_anno$Chr <- factor(gene_anno$Chr, c(1:22,"X","Y"), ordered=TRUE)
gene_anno <- gene_anno [do.call(order, gene_anno [, c("Chr", "Start", "End", "GENES")]), ]
genes <- gene_anno[4]
rownames(genes) <- paste0("row_", seq(nrow(genes)))
#Genes
annotation2 <- gene_anno[1]
rownames(annotation2) <- row.names(genes)</pre>
```

```
my.breaks)))
p2 \leftarrow pheatmap(
 mat
                 = genes,
 name = "genes",
 color = my. color;
breaks = my. breaks,
solnames = TRUE,
 show_rownames
                = FALSE,
                = TRUE,
 drop_levels
 fontsize
                = 5,
 {\tt cluster\_rows}
                 = FALSE,
                 = FALSE,
 cluster_cols
 scale
                = "none"
 {\tt gaps\_row = head(as.numeric(cumsum(table(annotation\$Chr))), -1)}
p + p2
```



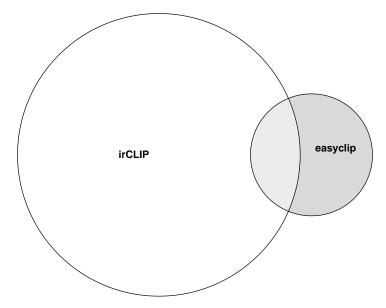
```
# Save the heatmap
pdf("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/2_Characterization/3_pvalue_0.05
    __fc1/HM_Gene_density.pdf", height = 20, width = 5)
p + p2
dev.off()
```

Comparison to easyCLIP

Next, we compared the regions of HNRNPC to the regions identified by Porter et al. using the easyCLIP protocol (PMID: 33692367). These regions were determined using the MACS2 tool.

```
#Easyclip data from Porter et al
\label{eq:hnrnpc.ec} \begin{split} \text{HNRNPC.EC} \leftarrow \text{read.delim} (\text{"/Users/lducoli/Documents/Postdoc/PD\_Projects/3\_irCLIP\_RNP/Seq/1\_Subzones}) \end{split}
                /2_Characterization/easyclip/HNRNPC_genomics_10perc_peaks.narrowPeak.txt", header = TRUE)
overlap_function <- function (br, EC) {
       rbp <- br %% dplyr::select(chromosome,region_begin,region_end)
        rbp$region_begin <- round(((rbp$region_begin+rbp$region_end)/2)-250,0)
        rbp$region end <- rbp$region begin+500
        rbp <- GRanges(rbp[,grep("chromosome", colnames(rbp))], IRanges(start=rbp[,grep("region_begin",
                    colnames(rbp))], end=rbp[,grep("region_end", colnames(rbp))]))
        rbp <- unique(rbp)
       EC <- EC %% dplyr::select(chrom, start, end)
       EC\$ start \leftarrow round (((EC\$ start +EC\$ end)/2)-250,0)
       EC\$end \leftarrow EC\$start+500
       end=EC[, grep("end", colnames(EC))]))
       EC <- unique (EC)
        gr.list <- list(rbp,EC)
        return (gr. list)
HNRNPC.cp <- overlap_function(HNRNPC, HNRNPC.EC)
HNRNPC_res <- makeVennDiagram(Peaks-HNRNPC.cp, NameOfPeaks=c("HNRNPC", "easyclip"), plot = FALSE)
#HNRNPC
HNRNPC_high_venn <- eulerr::euler(c(irCLIP = length(HNRNPC$regionID)-HNRNPC_res$vennCounts
                   [,3][4], easyclip = HNRNPC_res$vennCounts[,3][2], "irCLIP&easyclip" = HNRNPC_res$vennCounts
                   [,3][4]), shape = "ellipse")
HNRNPC_plot <- plot(HNRNPC_high_venn, main = paste("HNRNPC:", length(HNRNPC$regionID)—HNRNPC_res
                \label{eq:counts} $$\operatorname{vennCounts}[3][4], \ \operatorname{HNRNPC}_{\operatorname{res}} \operatorname{vennCounts}[3][4], \ \operatorname{HNRNPC}_{\operatorname{res}} \operatorname{vennCounts}[3][2], \ \ \operatorname{"P-value} = ", \ \ \operatorname{hnRNPC}_{\operatorname{res}} \operatorname{vennCounts}[3][2], \ \ \operatorname{"P-value} = ", \ \ \operatorname{hnRNPC}_{\operatorname{res}} \operatorname{hnR
                signif (HNRNPC_res$p.value[[3]], digits=2)))
HNRNPC_plot
```

HNRNPC: 5841 397 765 P-value = 4.7e-196



Piechart of the binding distribution

Finally, the proportion of binding regions of the RBPs was visualized.

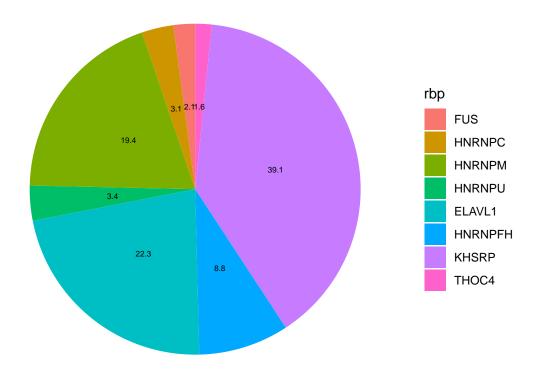
```
#Store regions as Granges
FUS.gr <- GRanges(FUS[,grep("chromosome", colnames(FUS))],
                        IRanges(start=FUS[,grep("region_begin", colnames(FUS))], end=FUS[,grep("region_
     end", colnames(FUS))]),
                        names = FUS[,grep("regionID", colnames(FUS))], strand = FUS[,grep("region_
     strand", colnames(FUS))],
rbp = "FUS")
HNRNPC.gr <- GRanges(HNRNPC[, grep("chromosome", colnames(HNRNPC))],
                        IRanges(start=HNRNPC[,grep("region_begin", colnames(HNRNPC))], end=HNRNPC[,grep
     ("region_end", colnames(HNRNPC))]),
                        names = HNRNPC[,grep("regionID", colnames(HNRNPC))], strand = HNRNPC[,grep("
     \underline{\tt region\_strand"}\,,\,\,\, \underline{\tt colnames(HNRNPC))\,]\,,
                        rbp = "HNRNPC")
\label{eq:hnrnpm} \text{HNRNPM}.\ \text{gr} < -\ \text{GRanges}(\text{HNRNPM}[\ , \text{grep}(\ \text{"chromosome"}\ ,\ \ \text{colnames}(\text{HNRNPM})\ )\ ]\ ,
                        IRanges(start=INRNPM], grep("region_begin", colnames(HNRNPM))], end=INRNPM[,grep
     ("region_end", colnames(HNRNPM))]),
names = HNRNPM[,grep("regionID", colnames(HNRNPM))], strand = HNRNPM[,grep("
     region_strand", colnames(HNRNPM))],
                        rbp = "HNRNPM")
\label{eq:hnrnpu} \textit{HNRNPU}.\ \textit{gr} \leftarrow \textit{GRanges}\left(\textit{HNRNPU}\right[,\textit{grep}\left("\textit{chromosome}",\ \textit{colnames}\left(\textit{HNRNPU}\right)\right)],
                        IRanges(start=HNRNPU[,grep("region_begin", colnames(HNRNPU))], end=HNRNPU[,grep
     ("region_end", colnames(HNRNPU))]),
                        names = HNRNPU[, grep("regionID", colnames(HNRNPU))], strand = HNRNPU[, grep("
      \begin{array}{c} {\rm region\_strand}\,",\ {\rm colnames}\,({\rm HNRNPU})\,)\,]\,, \\ {\rm rbp}\,=\,"{\rm HNRNPU}") \end{array}
```

```
ELAVL1.gr <- GRanges(ELAVL1[, grep("chromosome", colnames(ELAVL1))],
                                                  IRanges(start=ELAVL1[,grep("region_begin", colnames(ELAVL1))], end=ELAVL1[,grep
           ("region_end", colnames(ELAVL1))]),
                                                 names = ELAVL1[,grep("regionID", colnames(ELAVL1))], strand = ELAVL1[,grep("
           region_strand", colnames(ELAVL1))],
                                                 rbp = "ELAVL1"
HNRNPFH.gr <- GRanges (HNRNPFH], grep ("chromosome", colnames (HNRNPFH))],
                                                 IRanges(start = HNRNPFH[, grep("region\_begin", colnames(HNRNPFH))], end = HNRNPFH[, grep("region\_begin", colnames(HNRNPFH)]), end 
           grep("region_end", colnames(HNRNPFH))]),
                                                 names = HNRNPFH[, grep("regionID", colnames(HNRNPFH))], strand = HNRNPFH[, grep("
           region_strand", colnames(HNRNPFH))],
                                                 rbp = "HNRNPFH")
 KHSRP.\,gr\,\leftarrow\,GRanges(KHSRP[\,,grep(\,"chromosome"\,,\,\,colnames(KHSRP)\,)\,]\,,
                                                  IRanges(start=KHSRP[,grep("region_begin", colnames(KHSRP))], end=KHSRP[,grep("
                                              colnames (KHSRP))]),
           region_end",
                                                 names = KHSRP[,grep("regionID", colnames(KHSRP))], strand = KHSRP[,grep("region
           _strand", colnames(KHSRP))],
                                                 rbp = "KHSRP
THOC4.\,gr\, <\!-\, GRanges(THOC4[\,,grep(\,"chromosome"\,,\,\, colnames(THOC4)\,)\,]\,,
                                                 IRanges (start=THOC4[, grep("region begin", colnames(THOC4))], end=THOC4[, grep("
           region_end", colnames(THOC4))]),
                                                 names = THOC4[\ ,grep(\ "regionID"\ ,\ colnames(THOC4))\ ]\ ,\ strand = THOC4[\ ,grep(\ "regionID"\ ,\ colnames]\ ]
           <u>_strand</u>", colnames(THOC4))],
                                                 rbp = "THOC4")
HNRNPC. gr
```

```
## GRanges object with 6238 ranges and 2 metadata columns:
###
             segnames
                                     ranges strand
                                                                          names
##
                <Rle>
                                  <IRanges> <Rle>
                                                                   <character>
                 chr1 100016040-100016110
                                                       chr1_100016040_10001..
                                                  +
##
         [1]
         2
                 chr1 100016057-100016127
                                                       chr1_100016057_10001..
1111
                                                  +
                 chr1 100018855-100018905
         [3]
                                                       chr1 100018855 10001...
##
                                                  +
                 chr1 100018861-100018911
##
         4
                                                  +
                                                       chr1_100018861_10001..
                           1010359\!-\!1010409
###
                 chr1
                                                  +
                                                       chr1_1010359_1010409...
###
      [6234]
                 chrX
                         74586715\!-\!74586765
                                                       chrX_74586715_745867..
##
                         74588275 - 74588345
                                                       {\rm chr}X\_74588275\_745883\dots
##
      6235
                 chrX
      6236
                                                       chrX_74588435_745885..
##
                 chrX
                         74588435\!-\!74588505
      [6237]
                                                  +
                                                       chrX_77903736_779037...
##
                 chrX
                         77903736-77903786
     [6238]
                 chrX
                         80324393 - 80324483
                                                      chrX 80324393 803244...
###
##
                      rbp
1111
             <character>
         [1]
                  HNRNPC
##
         [2]
                  HNRNPC
##
###
         [3]
                  HNRNPC
##
         4
                  HNRNPC
         [5]
                  HNRNPC
##
##
      [6234]
                  HNRNPC
##
      6235
                  HNRNPC
##
      [6236]
                  HNRNPC
##
##
      [6237]
                  HNRNPC
##
     [6238]
                  HNRNPC
##
     seqinfo: 23 sequences from an unspecified genome; no seqlengths
```

```
#Generate a reduced data.frame of the regions — if you want to run it uncomment it (take aroung 40min...)
# u = unique(c(FUS.gr, HNRNPC.gr, HNRNPM.gr, HNRNPU.gr, ELAVL1.gr, HNRNPFH.gr, KHSRP.gr, THOC4.gr
```

```
# geneRange <- sortSeqlevels(u)
# geneRange <- sort (geneRange)
# geneReduce <- GenomicRanges::reduce(geneRange, drop.empty.ranges = TRUE, with.revmap = TRUE,
    \min. gapwidth = 1
# mcols(geneReduce)$rbp <- "Undefined"
# pb <- txtProgressBar(min = 0, max = length(geneReduce), style = 3)
# for (i in seq_len(length(geneReduce))) {
      mergeInd <- unlist (mcols (geneReduce) [i, 1])
      mcols(geneReduce)[i\,,\ "rbp"] \leftarrow paste(mcols(geneRange)[mergeInd\,,\ "rbp"]\,,\ collapse = ",\ ")
      setTxtProgressBar(pb = pb, value = i)
# }
# regionRes <- as.data.frame(geneReduce)
# regionRes <- regionRes[,-6]
# regionRes$regions 		— paste("Region", 1:length(regionRes$seqnames), sep = "_")
# write.table(regionRes, "~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/2_
    Characterization/3_pvalue_0.05_fc1/Confident_region.txt", quote = F, sep = "\t", row.names =
regionRes <- read.delim("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/2_
    Characterization/3_pvalue_0.05_fc1/Confident_region.txt", header = TRUE)
#Generate Piechart
regionRes2 <- regionRes \%% mutate(rbp = str split(rbp, ", ")) \%% unnest(rbp)
regionRes3 <- regionRes2 %% dplyr::count(rbp)
regionRes3$rbp <- factor(regionRes3$rbp)
regionRes3$rbp <- factor(regionRes3$rbp, levels = c("FUS", "HNRNPC", "HNRNPM", "HNRNPU", "ELAVL1"
    , "HNRNPFH", "KHSRP", "THOC4"))
regionRes3 <- regionRes3 %%
  arrange(desc(rbp)) %%
  mutate(prop = round((n / sum(regionRes3$n))*100,1)) \%\%
  mutate(ypos = cumsum(prop) - 0.5*prop)
\begin{split} & ggplot(regionRes3\,,~aes(x="",~y=prop\,,~fill=rbp)) + \\ & geom\_bar(width\,=\,1,~stat\,=\,"identity")\,+ \end{split}
    coord_polar("y", start=0) +
  theme_minimal()+
  theme(
  axis.title.x = element_blank(),
  axis.title.y = element_blank(),
  panel.border = element_blank(),
  panel.grid=element blank().
  axis.ticks = element_blank(),
  plot.title=element_text(size=14, face="bold")
  theme(axis.text.x=element_blank()) +
  geom_text(aes(y = ypos, label = prop), color = "black", size=2)
```



```
pdf("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/2_Characterization/3_pvalue_0.05
    _fc1/Pie_chart.pdf", width = 5, height = 5)
ggplot(regionRes3, aes(x="", y=prop, fill=rbp))+
geom_bar(width = 1, stat = "identity") +
    coord_polar("y", start=0) +
    theme_minimal()+
    theme(
    axis.title.x = element_blank(),
    panel.border = element_blank(),
    panel.grid=element_blank(),
    panel.grid=element_blank(),
    plot.title=element_text(size=14, face="bold")
    )+
    theme(axis.text.x=element_blank()) +
    geom_text(aes(y = ypos, label = prop), color = "black", size=2)
dev.off()
```

3. Profile plot of RBPs around the highest nucleotide in the lowest section of the gel

Here, we looked whether irCLIPv2 could generate a 3'shift that can be reconnected to the presence of another RBP.

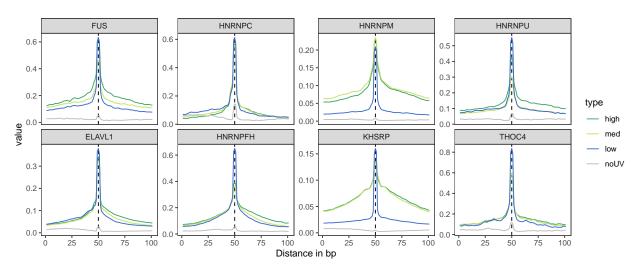
```
#Function for the generation of the profile plots
shift3p <- function(data, rbp) {
    cat("Prepare granges\n")
```

```
p <- subset(data, region_strand == "+" & chromosome != "chrM")
 \begin{split} \text{gr.p} \leftarrow & \text{GRanges}(p[\,,\text{grep}(\text{"chr\_S1"},\,\,\text{colnames}(p))\,]\,,\\ & \text{IRanges}(\text{start} = & p[\,,\text{grep}(\text{"begin\_S1"},\,\,\text{colnames}(p))\,]\,, \end{split}
                                                 end=p[,grep("end_S1", colnames(p))]),
                                 names = p[,grep("max\_S1\_id", colnames(p))],
                                 strand = p[,grep("strand_S1", colnames(p))],
category = p[,grep("category", colnames(p))])
m <- subset(data, region_strand == "_" & chromosome != "chrM")
 gr.m <- GRanges(m[, grep("chr_S1", colnames(m))],
                                 IRanges(start=m[,grep("begin_S1", colnames(m))],
                                                  end\!\!=\!\!m[\;,grep\left(\,"end\_S1\,"\,,\;colnames\left(m\right)\,\right)\,]\,)\;,
                                 names = m[, grep("max_S1_id", colnames(m))], strand = m[, grep("strand_S1", colnames(m))],
                                  category = m[, grep("category", colnames(m))])
 \mathrm{start}\,(\,\mathrm{gr}\,.\,\mathrm{p})\,<\!\!-\,(\,\mathrm{start}\,(\,\mathrm{gr}\,.\,\mathrm{p})\!+\!\!\mathrm{end}\,(\,\mathrm{gr}\,.\,\mathrm{p})\,)/2
 end(gr.p) \leftarrow start(gr.p)
 start(gr.m) <- (start(gr.m)+end(gr.m))/2
 end(gr.m) <- start(gr.m)
 cat("Load bigwig files \n")
 setwd("/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/1_Bigwig_sum")
setwd("/Users/lducoli/Documents/Postdoc/PD\_Projects/3\_irCLIP-RNP/Seq/1\_Subzones/1\_bw.S1.r1.p \leftarrow import.bw(paste(rbp, "293T\_UVC\_S1\_R1.p.scaleddewseq.bw", sep = "_")) bw.S1.r1.m \leftarrow import.bw(paste(rbp, "293T\_UVC\_S1\_R1.m.scaleddewseq.bw", sep = "_")) bw.S1.r2.p \leftarrow import.bw(paste(rbp, "293T\_UVC\_S1\_R2.p.scaleddewseq.bw", sep = "_")) bw.S1.r2.m \leftarrow import.bw(paste(rbp, "293T\_UVC\_S1\_R2.m.scaleddewseq.bw", sep = "_")) bw.S2.r1.p \leftarrow import.bw(paste(rbp, "293T\_UVC\_S2\_R1.p.scaleddewseq.bw", sep = "_")) bw.S2.r1.m \leftarrow import.bw(paste(rbp, "293T\_UVC\_S2\_R1.m.scaleddewseq.bw", sep = "_")) bw.S2.r2.p \leftarrow import.bw(paste(rbp, "293T\_UVC\_S2\_R2.p.scaleddewseq.bw", sep = "_")) bw.S2.r2.m \leftarrow import.bw(paste(rbp, "293T\_UVC\_S2\_R2.m.scaleddewseq.bw", sep = "_")) bw.S3.r1.p \leftarrow import.bw(paste(rbp, "293T\_UVC\_S2\_R2.m.scaleddewseq.bw", sep = "_")) bw.S3.r1.m \leftarrow import.bw(paste(rbp, "293T\_UVC\_S3\_R1.p.scaleddewseq.bw", sep = "_")) bw.S3.r2.p \leftarrow import.bw(paste(rbp, "293T\_UVC\_S3\_R2.p.scaleddewseq.bw", sep = "_")) bw.S3.r2.m \leftarrow import.bw(paste(rbp, "293T\_UVC\_S3\_R2.m.scaleddewseq.bw", sep = 
bw.S3.r2.m <- import.bw(paste(rbp, "293T_UVC_S3_R2.m.scaleddewseq.bw", sep = "
noUV.bw.r1.p <- import.bw(paste(rbp, "293I_UVC_S5_RZ.iii.scateddewseq.bw", sep = "_"))
noUV.bw.r1.p <- import.bw(paste(rbp, "293I_noUV_R1.p.scaleddewseq.bw", sep = "_"))
noUV.bw.r1.m <- import.bw(paste(rbp, "293I_noUV_R1.m.scaleddewseq.bw", sep = "_"))
noUV.bw.r2.p <- import.bw(paste(rbp, "293I_noUV_R2.p.scaleddewseq.bw", sep = "_"))
noUV.bw.r2.m <- import.bw(paste(rbp, "293I_noUV_R2.m.scaleddewseq.bw", sep = "_"))
 cat("Create profile matrices\n")
 S1.mat1.p = normalizeToMatrix(bw.S1.r1.p, gr.p, value_column = "score", extend = 200, mean_mode
       = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
 S1.mat1.m = normalizeToMatrix(bw.S1.r1.m, gr.m, value_column = "score", extend = 200, mean_mode
      = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
 S1.mat2.p = normalizeToMatrix(bw.S1.r2.p, gr.p, value_column = "score", extend = 200, mean_mode
       = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
 S1.mat2.m = normalizeToMatrix(bw.S1.r2.m, gr.m, value column = "score", extend = 200, mean mode
       = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
 S2.matl.p = normalizeToMatrix(bw.S2.rl.p, gr.p, value_column = "score", extend = 200, mean_mode
       = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
 S2.mat1.m = normalizeToMatrix(bw.S2.r1.m, gr.m, value_column = "score", extend = 200, mean_mode
       = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
 S2.mat2.p = normalizeToMatrix(bw.S2.r2.p, gr.p, value_column = "score", extend = 200, mean_mode
       = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
 S2.mat2.m = normalizeToMatrix(bw.S2.r2.m, gr.m, value_column = "score", extend = 200, mean_mode
       = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
 S3.mat1.p = normalizeToMatrix(bw.S3.r1.p, gr.p, value_column = "score", extend = 200, mean_mode
       = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
 S3.mat1.m = normalizeToMatrix(bw.S3.r1.m, gr.m, value_column = "score", extend = 200, mean_mode
       = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
 S3.mat2.p = normalizeToMatrix(bw.S3.r2.p, gr.p, value_column = "score", extend = 200, mean_mode
 = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
S3.mat2.m = normalizeToMatrix(bw.S3.r2.m, gr.m, value_column = "score", extend = 200, mean_mode
       = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
noUV.mat1.p = normalizeToMatrix(noUV.bw.r1.p, gr.p, value_column = "score", extend = 200, mean_
```

```
mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
noUV.mat1.m = normalizeToMatrix(noUV.bw.r1.m, gr.m, value_column = "score", extend = 200, mean_
    mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
no UV.\,mat 2.\,p = normalize ToMatrix (no UV.\,bw.\,r2.\,p, \ gr.\,p, \ value\_column = "score", \ extend = 200, \ mean\_column = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 1000 = 10000 = 10000 = 10000 = 10000 = 10000 = 10000 = 10000 = 10000 = 10000 = 10000 = 10000 = 100000
   mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
noUV.mat2.m = normalizeToMatrix(noUV.bw.r2.m, gr.m, value_column = "score", extend = 200, mean_
    mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
cat("Merge replicates\n")
S1.mat.p = (S1.mat1.p[]+S1.mat2.p[])/2
rownames(S1.mat.p) <- gr.p$names
S2.mat.p = (S2.mat1.p[]+S2.mat2.p[])/2
rownames(S2.mat.p) <- gr.p$names
S3. mat.p = (S3. mat1.p[]+S3. mat2.p[])/2
rownames(S3.mat.p) \leftarrow gr.p$names
noUV.mat.p = (noUV.mat1.p[]+noUV.mat2.p[])/2
rownames(noUV.mat.p) <- gr.p$names
S1. mat.m = (S1. mat1.m[] + S1. mat2.m[])/2
rownames (S1.mat.m) <- gr.m$names
S2.mat.m = (S2.mat1.m[]+S2.mat2.m[])/2
rownames (S2.mat.m) <- gr.m$names
S3.mat.m = (S3.mat1.m[]+S3.mat2.m[])/2
rownames (S3.mat.m) <- gr.m$names
noUV. mat.m = (noUV. mat1.m[] + noUV. mat2.m[])/2
rownames (noUV.mat.m) <- gr.m$names
cat("Center on S1 max\n")
s.sub.p <- S1.mat.p
s.sub.m <- S1.mat.m
get\_centered\_region \leftarrow function(x, y)  {
   mat = list()
    for(i in 1:length(rownames(x))){
        center \leftarrow which \max(x[i, 175:225])
        shift <-y[i, (175+center-50):(175+center+50)]
        mat[[i]] <- shift
   mat <- matrix(unlist(mat), ncol = length(rownames(x)))
   mat <- t (mat)
   rownames (mat) <- rownames (x)
    return (mat)
S1.p <- get_centered_region(s.sub.p, S1.mat.p)
S2.p <- get_centered_region(s.sub.p, S2.mat.p)
S3.p <- get_centered_region(s.sub.p, S3.mat.p)
noUV.p <- get_centered_region(s.sub.p, noUV.mat.p)</pre>
S1.m <- get_centered_region(s.sub.m, S1.mat.m)
S2.m <- get_centered_region(s.sub.m, S2.mat.m)
S3.m <- get_centered_region(s.sub.m, S3.mat.m)
noUV.m <- get_centered_region(s.sub.m, noUV.mat.m)
cat("Create table for profile plot\n")
S1.mat \leftarrow rbind(S1.p, S1.m)
S2.mat <- rbind(S2.p, S2.m)
S3.mat <- rbind(S3.p, S3.m)
noUV.mat <- rbind(noUV.p, noUV.m)
k = 3
S1.mat2 = t(runmean(t(S1.mat), k))
```

```
S2.mat2 = t(runmean(t(S2.mat), k))
  S3.mat2 = t(runmean(t(S3.mat), k))
  noUV.mat2 = t(runmean(t(noUV.mat), k))
  S1.gg <- data.frame("value" = colMeans(S1.mat2, na.rm=TRUE))
  S1.gg$type <- "low
  S1.gg\$xlab <- c(1:dim(S1.mat2)[2])
  S2.gg <- data.frame("value" = colMeans(S2.mat2, na.rm=TRUE))
  S2.gg$type <- "med'
  S2.gg\$xlab \leftarrow c(1:dim(S2.mat2)[2])
  S3.gg <- data.frame("value" = colMeans(S3.mat2, na.rm=TRUE))
  S3.gg$type <- "high"
  S3.gg\$xlab <- c(1:dim(S3.mat2)[2])
  noUV.gg <- data.frame("value" = colMeans(noUV.mat2, na.rm=TRUE))
  noUV.gg$type <- "noUV
  noUV.gg\$xlab \leftarrow c(1:dim(noUV.mat2)[2])
  all <- rbind(S1.gg,S2.gg,S3.gg,noUV.gg)
  all$type <- factor(all$type)
  all $type <- factor(all $type, levels = c("high", "med", "low", "noUV"))
  return(all)
#If you want to run our code make sure to uncomment the two paragraphs below! It takes a lot of
# FUS.plot <- shift3p(FUS, "FUS")
# HNRNPC. plot <- shift3p (HNRNPC, "HNRNPC")
# HNRNPM. plot <- shift3p (HNRNPM, "HNRNPM")
# HNRNPU.plot <- shift3p(HNRNPU, "HNRNPU")
# ELAVL1.plot <- shift3p(ELAVL1, "ELAVL1")
# HNRNPFH. plot <- shift3p (HNRNPFH, "HNRNPFH")
# KHSRP. plot <- shift3p(KHSRP, "KHSRP")
# THOC4. plot <- shift3p (THOC4, "THOC4")
# saveRDS(FUS.plot, "/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/4_
    Peak_shift/3_pvalue_0.05_fc1/FUS_3pshift.rds"
# saveRDS(HNRNPC.plot, "/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/
    4\_Peak\_shift/3\_pvalue\_0.05\_fc1/HNRNPC\_3pshift.rds")
# saveRDS(HNRNPM.plot, "/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/
    4_Peak_shift/3_pvalue_0.05_fc1/HNRNPM_3pshift.rds")
# saveRDS(HNRNPU.plot, "/Users/Iducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/
    4_Peak_shift/3_pvalue_0.05_fc1/HNRNPU_3pshift.rds")
# saveRDS(ELAVL1.plot, "/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/
4_Peak_shift/3_pvalue_0.05_fc1/ELAVL1_3pshift.rds")
# saveRDS(HNRNPFH.plot, "/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones
    /4_Peak_shift/3_pvalue_0.05_fc1/HNRNPFH_3pshift.rds"
# saveRDS(KHSRP.plot, "/Users/lducoli/Documents/Postdoc/PD Projects/3 irCLIP-RNP/Seq/1 Subzones/4
    _Peak_shift/3_pvalue_0.05_fc1/KHSRP_3pshift.rds")
# saveRDS(THOC4.plot, "/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/4
    _Peak_shift/3_pvalue_0.05_fc1/THOC4_3pshift.rds")
FUS. plot <- readRDS("/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/4_
    Peak_shift/3_pvalue_0.05_fc1/FUS_3pshift.rds")
HNRNPC. plot <- readRDS("/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/
    4_Peak_shift/3_pvalue_0.05_fc1/HNRNPC_3pshift.rds")
HNRNPM.plot <- readRDS("/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/
    4_Peak_shift/3_pvalue_0.05_fc1/HNRNPM_3pshift.rds")
HNRNPU.plot <- readRDS("/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/
    4_Peak_shift/3_pvalue_0.05_fc1/HNRNPU_3pshift.rds")
ELAVLI.plot <- readRDS("/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/
    4_Peak_shift/3_pvalue_0.05_fc1/ELAVL1_3pshift.rds")
HNRNPFH. plot <- readRDS("/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones
    /4_Peak_shift/3_pvalue_0.05_fc1/HNRNPFH_3pshift.rds")
KHSRP.plot <- readRDS("/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/4
     Peak_shift/3_pvalue_0.05_fc1/KHSRP_3pshift.rds")
THOC4.plot <- readRDS("/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/4
    _Peak_shift/3_pvalue_0.05_fc1/THOC4_3pshift.rds")
```

```
FUS. plot$rbp <- "FUS"
HNRNPC. plot $rbp <- "HNRNPC"
HNRNPM. plot$rbp <- "HNRNPM"
HNRNPU. plot$rbp <- "HNRNPU"
ELAVL1. plot$rbp <- "ELAVL1"
HNRNPFH. plot$rbp <- "HNRNPFH"
KHSRP. plot$rbp <- "KHSRP"
THOC4. plot$rbp <- "THOC4"
all <-- rbind (FUS. plot , HNRNPC. plot , HNRNPM. plot , HNRNPU. plot , HNRNPFH. plot , KHSRP. plot , ELAVL1.
    plot, THOC4. plot)
all$rbp <- factor(all$rbp, levels = c("FUS", "HNRNPC", "HNRNPM", "HNRNPU", "ELAVL1", "HNRNPFH", "
    KHSRP", "THOC4"))
cols <- c("high" = "#3fa966", "med" = "#c8e06a", "low" = "#3d6ccd", "noUV" = "grey")
all.plot <- all \%\%
  ggplot( aes(x=xlab, y=value, group = type, color = type)) +
  geom_vline(xintercept = c(50), linetype="dashed")+
  # geom_area(aes(fill = type), alpha = 0.05, position = 'identity') +
  geom_line() +
  scale_color_manual(values = cols)+
  # xlim(c(145, 245))+
  xlab("Distance in bp") +
  theme bw() +
  theme( panel.grid.major = element_blank(),
         panel.grid.minor = element_blank(),
         panel.background = element_blank(),
         axis.line = element_blank(),
         plot.title = element text(hjust = 0.5)) +
  facet_wrap(vars(rbp), ncol = 4, nrow = 2, scales = "free_y")
all.plot
```



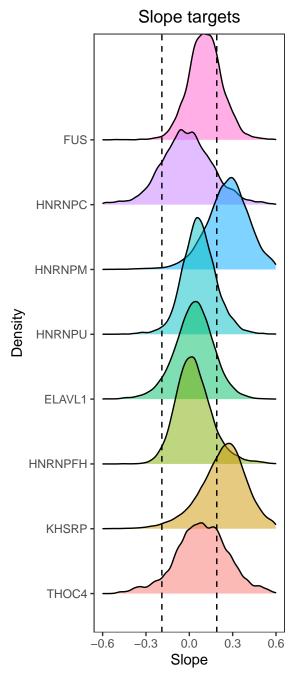
```
 \begin{array}{lll} pdf("\sim/Documents/Postdoc/PD\_Projects/3\_irCLIP\_RNP/Seq/1\_Subzones/4\_Peak\_shift/3\_pvalue\_0.05\_fc1/Profileplot\_regions.pdf", width = 10, height = 4) \\ all.plot \\ dev.off() \end{array}
```

4. Slope categorization

Slope calculation

The regions were categorized according to their slope calculated from the total normalized RT stop counts. Please see the snakemake at this Github link for more information.

```
datam <- rbind (FUS.HNRNPC.HNRNPM.HNRNPU.ELAVL1.HNRNPFH.KHSRP.THOC4)
datam$rbp <- factor(datam$rbp, levels = c("THOC4", "KHSRP", "HNRNPFH", "ELAVL1", "HNRNPU", "HNRNPM", "
    HNRNPC", "FUS"))
datam.l \leftarrow subset(datam, lmslope \leftarrow -sd(datam$lmslope, na.rm = TRUE))
datam.m <- subset (datam, lmslope > -sd (datam$lmslope, na.rm = TRUE) & lmslope < sd (datam$lmslope,
     na.rm = TRUE))
datam.m$category_all <- "med"
{\tt datam.h} \leftarrow {\tt subset} \left( {\tt datam} \,, \,\, {\tt lmslope} \, > \, {\tt sd} \left( {\tt datam\$lmslope} \,, \,\, {\tt na.rm} \, = \, {\tt TRUE} \right) \right)
datam.h$category_all <- "high"
datam <- rbind(datam.l, datam.m, datam.h)
write.table(datam, "~/Documents/Postdoc/PD Projects/3 irCLIP-RNP/Seq/1 Subzones/3 Categorization/
    3_pvalue_0.05_fc1/All_RBP_categorization.txt", quote = F, row.names = F, sep = "\t")
ggplot(data=datam, aes(x = lmslope, fill = rbp, y = rbp))+
  geom_density_ridges(alpha = 0.5)+
  geom_vline(aes(xintercept=0+sd(lmslope, na.rm = TRUE)), color="black", linetype="dashed", size
    =0.5)+
  geom_vline(aes(xintercept=0-sd(lmslope, na.rm = TRUE)), color="black", linetype="dashed", size
  xlab("Slope") + ylab("Density") + xlim(-0.6, 0.6) +
  ggtitle("Slope targets") +
  theme_bw() +
  theme(legend.position = "none", panel.grid.major = element_blank(),
        panel.grid.minor = element_blank(),
        panel.background = element_blank(),
        axis.line = element_blank(),
        plot.title = element text(hjust = 0.5))
```



```
panel.background = element_blank(),
    axis.line = element_blank(),
    plot.title = element_text(hjust = 0.5))
dev.off()
```

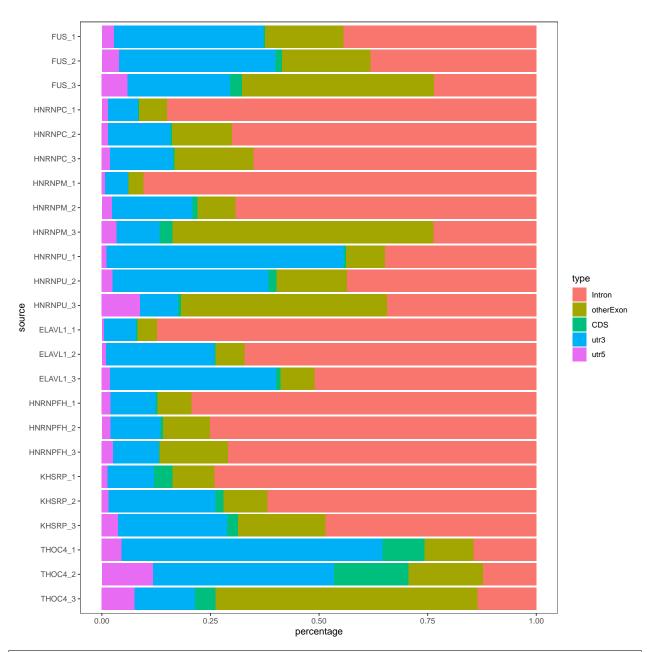
Genomic annotation of categorized regions

```
#Load annotation
txdb \leftarrow loadDb("\sim/Documents/Postdoc/PD\_Projects/3\_irCLIP\_RNP/Seq/1\_Subzones/0\_Annotation/gencode.
    v39. annotation")
#Create a granges for all RBPs
datam.gr <- GRanges(datam[,grep("chromosome", colnames(datam))],
                   IRanges(start=datam[,grep("region_begin", colnames(datam))], end=datam[,grep("
    region_end", colnames(datam))]),
                   names = datam[,grep("regionID", colnames(datam))], strand = datam[,grep("region
    _strand", colnames(datam))],
                   rbp = datam[,grep("rbp", colnames(datam))],
                   category = datam[,grep("category_all", colnames(datam))])
datam.grl <- GRangesList("FUS_3" = datam.gr %% filter(rbp == "FUS" & category == "low"),
                    "FUS_2" = datam.gr %% filter(rbp == "FUS" & category == "med"),
                    "FUS_1" = datam.gr %% filter(rbp == "FUS" & category == "high"),
                    "HNRNPC_3" = datam.gr %% filter(rbp = "HNRNPC" & category = "low"),
                    "HNRNPC_1" = datam.gr %% filter(rbp = "HNRNPC" & category = "med"),
"HNRNPC_1" = datam.gr %% filter(rbp = "HNRNPC" & category = "high"),
                    "HNRNPM_3" = datam.gr %% filter(rbp = "HNRNPM" & category = "low"),
                    "HNRNPM_2" = datam.gr %% filter(rbp == "HNRNPM" & category == "med"),
                    "HNRNPM_1" = datam.gr %% filter(rbp == "HNRNPM" & category == "high"),
"HNRNPU_3" = datam.gr %% filter(rbp == "HNRNPU" & category == "low"),
                    "HNRNPU_2" = datam.gr %% filter(rbp = "HNRNPU" & category = "med"),
                    "HNRNPU 1" = datam.gr %% filter (rbp = "HNRNPU" & category = "high"),
                    "ELAVL1_3" = datam.gr %% filter(rbp == "ELAVL1" & category == "low"),
                    "ELAVL1_2" = datam.gr %% filter(rbp == "ELAVL1" & category == "med"),
"ELAVL1_1" = datam.gr %% filter(rbp == "ELAVL1" & category == "high"),
                    "HNRNPFH_3" = datam.gr %% filter(rbp = "HNRNPFH" & category = "low"),
                    "HNRNPFH_2" = datam.gr %% filter(rbp = "HNRNPFH" & category = "med"),
                    "HNRNPFH_1" = datam.gr %% filter(rbp = "HNRNPFH" & category = "high"),
                    "KHSRP_3" = datam.gr %% filter(rbp == "KHSRP" & category == "low"),
                    "KHSRP_2" = datam.gr %% filter(rbp == "KHSRP" & category == "med"),
                    "KHSRP_1" = datam.gr %% filter(rbp = "KHSRP" & category = "high"),
                    "THOC4_3" = datam.gr %% filter(rbp = "THOC4" & category = "low"),
                    "THOC4_2" = datam.gr %% filter(rbp = "THOC4" & category = "med"),
                    "THOC4_1" = datam.gr %% filter(rbp = "THOC4" & category = "high"))
#Generate the plot for the annotation
dist <- genomicElementDistribution(datam.grl, TxDb = txdb, promoterRegion=c(upstream=2000,
    downstream=0), geneDownstream=c(upstream=0, downstream=2000), ignore.strand = FALSE, plot =
    FALSE)
data <- dist*plot$data %% filter(category = "Exon level") %% mutate(across('type', str_replace
      'undefined', 'Intron'))
data$type <- factor(data$type, levels = c("Intron", "otherExon", "CDS", "utr3", "utr5"))
ggplot(data, aes_string(x = "source", y = "percentage", fill = "type")) + geom_bar(stat = "
    identity") + coord flip() +
            theme_bw() + theme(panel.grid.major = element_blank(), panel.grid.minor = element_
    blank(), panel.background = element_blank(),
                                 axis.line = element_blank(), plot.title = element_text(hjust =
    0.5)) +
            scale_x_discrete(limits = c("THOC4_3", "THOC4_2", "THOC4_1", "KHSRP_3", "KHSRP_2","
    KHSRP_1",
                                           "HNRNPFH_3", "HNRNPFH_2", "HNRNPFH_1", "ELAVL1_3", "ELAVL1
    _2","ELAVL1_1",
```

```
"HNRNPU_3", "HNRNPU_2", "HNRNPU_1", "HNRNPM_3", "HNRNPM_2"

, "HNRNPM_1", "HNRNPC_3", "HNRNPC_2", "HNRNPC_1", "FUS_3", "FUS_2", "FUS_1"

))
```



```
"HNRNPFH_3", "HNRNPFH_2", "HNRNPFH_1", "ELAVL1_3", "ELAVL1_3", "ELAVL1_3", "ELAVL1_3", "HNRNPM_1", "HNRNPM_1", "HNRNPM_1", "HNRNPM_2", "HNRNPC_3", "HNRNPC_2", "HNRNPC_1", "FUS_3", "FUS_2", "FUS_1" ))
dev.off()
```

5. Motif analysis

To understand which motifs are enriched 3' downstream from the highest peak, we performed enrichment analysis using AME. A tool from the MEME suite.

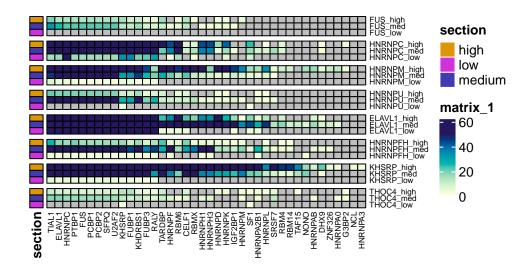
```
#Load the binding regions and genome
all <- read.delim("/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/3_
    Categorization/3_pvalue_0.05_fc1/All_RBP_categorization.txt", header = TRUE)
all$rbp_category <- paste(all$rbp, all$category_all, sep = "_")
human.genome <- BSgenome.Hsapiens.UCSC.hg38
expandRange = function(x, upstream, downstream) {
  strand_is_minus = strand(x) == "-
  on_plus = which(!strand_is_minus)
  on minus = which(strand is minus)
  start(x)[on\_plus] = start(x)[on\_plus] - upstream
  start(x)[on\_minus] = start(x)[on\_minus] - downstream
  end(x)[on\_plus] = end(x)[on\_plus] + downstream
  end(x)[on\_minus] = end(x)[on\_minus] + upstream
  return(x)
#Subset
h \leftarrow all \,\,\%\% \,\, filter\,(\, category\_all \,=\!\!\!\!=\!\!\!\!"high")
m <- all %% filter(category_all == "med")
l <- all %% filter (category_all == "low")
#Generate granges
h.gr <- GRanges(h[,grep("chr_S3", colnames(h))],
                           IRanges(start=h[,grep("begin S1", colnames(h))], end=h[,grep("end S1",
    colnames(h))]),
                           names = h[,grep("regionID", colnames(h))], strand = h[,grep("strand_S1
    ", colnames(h))],
                        section = h[,grep("category_all", colnames(h))], rbp_category = h[,grep("
    rbp_category", colnames(h))])
\label{eq:hgreen} \mbox{h.gr.exp} \leftarrow \mbox{h.gr.} \\ \mbox{% expandRange(upstream = 0, downstream = 225)}
m.gr <- GRanges(m[,grep("chr_S2", colnames(m))],
                        IRanges(start=m[,grep("begin_S1", colnames(m))], end=m[,grep("end_S1",
    colnames(m))]),
                        names = m[, grep("regionID", colnames(m))], strand = m[, grep("strand_S1",
    colnames (m))],
                        section = m[,grep("category_all", colnames(m))], rbp_category = m[,grep("
    rbp category", colnames(m))])
m.gr.exp <- m.gr %% expandRange(upstream = 0, downstream = 125)
1.gr <- GRanges(1[,grep("chr_S1", colnames(1))],
                        IRanges(start=l[,grep("begin_S1", colnames(1))], end=l[,grep("end_S1",
    colnames(1))]),
                        names = l[,grep("regionID", colnames(1))], strand = l[,grep("strand_S1",
    colnames(1))],
                        section = l[,grep("category_all", colnames(l))], rbp_category = l[,grep("
    rbp_category", colnames(1))])
1.gr.exp <- 1.gr \%% expandRange(upstream = 0, downstream = 50)
```

```
#Get sequence for motif analysis
gr <- bind_ranges(h.gr.exp, m.gr.exp, l.gr.exp)
# gr.seq.all <- unique(gr) %% get_sequence(human.genome)
gr.seq <- gr %% split(mcols(.) $rbp_category) %% get_sequence(human.genome)
gr.seq
```

```
## BStringSetList of length 24
## [["ELAVL1_high"]] chr1:100002555-100002830=TTTATTTCATCCTTIGTATTTTGTTTAGTTTCT...
### [["ELAVL1_low"]] chr1:100094646-100094746=CTACACGTCCCIGTCACTCCACCACCTAATTTTTG...
### [["ELAVL1_med"]] chr1:100000855-100001030=CATCTTTAATCCATTTTCGGGTTTTTTTTTTTAATG...
### [["FUS_high"]] chr1:108696396-108696671=TTTTATTTCGTGGTTTTAATGGTCTCAAGAACTTC...
### [["FUS_low"]] chr1:173865319-173865419=GATTTTAAAATTAAAGCAGATGCGAATCCICTGAGAA...
### [["FUS_med"]] chr1:103553673-103553848=TATCTGGTTTACGTATTTATAATGTGTTTAAAATTTCG...
### [["HNRNPC_high"]] chr1:100016040-100016315=TTAAAAGCACGTACGATACTTCTATTTTTTTTTT...
### [["HNRNPC_low"]] chr1:11024083-11024183=ATAGCAGTTAATTCCCCTTTTTTGACCCTTTTTGAGAT...
### [["HNRNPC_med"]] chr1:100018855-100019030=TTTTTATATTGAACATTCACGGTTTTAATTCTTTT...
### [["HNRNPFH_high"]] chr1:10045754-10046029=TTGACACCCCGGAGAAACGGTCAAACACTAGATGGT...
### <14 more elements>
```

```
#Load and prepare database
meme db <- read meme("~/Documents/Postdoc/PD Projects/3 irCLIP-RNP/Seq/1 Subzones/0 Annotation/
    motif/all.meme")
meme_db_df <- meme_db %% to_df()
meme_db_df$altname <- sapply(strsplit(meme_db_df$name, "_"), function(x) x[1])
gene <- unique (meme_db_df$altname)
motif list <- list()
for (i in 1:length(gene)) {
meme <- meme_db_df %% dplyr::filter(altname == gene[i]) %% remove_duplicate_motifs()
motif_list[[i]] \leftarrow meme
meme db df dedup <- motif list %% dplyr::bind rows(.id = "altname")
meme_db_df_dedup$altname <- sapply(strsplit(meme_db_df_dedup$name, "_"), function(x) x[1])
meme_db_new <- meme_db_df_dedup %% to_list()
#Run AME-if you want to run the code uncomment the next two lines (it takes around 4h on a M1
    Mac)
# ame_all <- gr.seq %% runAme(database = meme_db_new, silent = FALSE) %% dplyr::bind_rows(.id =
     "rbp category")
#Save AME results
# saveRDS(ame all, "~/Documents/Postdoc/PD Projects/3 irCLIP-RNP/Seq/1 Subzones/5 motif analysis/
    3_pvalue_0.05_fc1/AME_all_shuffleCtrl_bothsides.rds")
ame_all <- readRDS("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/5_motif_analysis/
    3_pvalue_0.05_fc1/AME_all_shuffleCtrl_bothsides.rds")
#Create heatmap of AME results
levels <- c("THOC4_low", "THOC4_med", "THOC4_high", "KHSRP_low", "KHSRP_med", "KHSRP_high", "HNRNPFH_low", "HNRNPFH_med", "HNRNPFH_high",
            "ELAVL1_low", "ELAVL1_med", "ELAVL1_high", "HNRNPU_low", "HNRNPU_med", "HNRNPU_high", "HNRNPM_low", "HNRNPM_med", "HNRNPM_high"
            "HNRNPC_low", "HNRNPC_med", "HNRNPC_high", "FUS_low", "FUS_med", "FUS_high")
ame_all$rbp_category <- factor(ame_all$rbp_category, levels = levels)
heatmap_mat <- ame_all %% dplyr::group_by(rbp_category, motif_alt_id, consensus) %% mutate(adj
    .pvalue = -log10(adj.pvalue)) %% dplyr::filter(adj.pvalue = max(adj.pvalue), motif_alt_id!
    = "NA")
```

```
heatmap_mat$motif_gene <- paste(heatmap_mat$motif_alt_id, heatmap_mat$consensus, sep = "_")
heatmap_mat$rbp <- sapply(strsplit(as.character(heatmap_mat$rbp_category), "_"), function(x) x
     [1])
heatmap_mat2 <- heatmap_mat %% dplyr::group_by(rbp, motif_gene) %% dplyr::summarise_at(vars(
    adj.pvalue), list(adj.pvalue = mean))
heatmap_mat2$protID <- sapply(strsplit(as.character(heatmap_mat2$motif_gene), "_"), function(x) x
     [1]
heatmap_mat3 <- heatmap_mat2 %% dplyr::group_by(rbp, protID) %% dplyr::filter(adj.pvalue == max
    (adj.pvalue))
heatmap_mat <- heatmap_mat %% dplyr::group_by(rbp_category, motif_alt_id) %% dplyr::filter(
    motif_gene %in% heatmap_mat3$motif_gene) %% dplyr::filter(adj.pvalue == max(adj.pvalue))
heatmap_mat <- heatmap_mat %% dplyr::select(rbp_category, motif_alt_id, adj.pvalue) %% unique()
     %% pivot_wider(names_from = motif_alt_id, values_from = adj.pvalue)
heatmap hm \leftarrow as.matrix(heatmap_mat[-1])
rownames(heatmap_hm) <- heatmap_mat$rbp_category
heatmap_hm <- heatmap_hm[match(levels(heatmap_mat$rbp_category), rownames(heatmap_hm)),]
FUS.TMT <- read.delim("/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/IMT_analysis/
     3_DEP/FUS_TMT_twofactor_results_sub.txt", header = TRUE)
FUS.TMT <- subset (FUS.TMT, order_sign != "FALSE_FALSE_FALSE_FALSE")
HNRNPC.TMT <- read.delim("/Users/lducoli/Documents/Postdoc/PD Projects/3 irCLIP-RNP/MS/IMT
     analysis/3\_DEP/hnC\_TMT\_two factor\_results\_sub.txt", \ header = TRUE)
HNRNPC.TMT <- subset (HNRNPC.TMT, order sign != "FALSE FALSE FALSE FALSE")
HNRNPM.TMT <- read.delim("/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/IMT_
    analysis/3_DEP/hnM_TMT_twofactor_results_sub.txt", header = TRUE)
HNRNPM.TMT <- subset (HNRNPM.TMT, order_sign != "FALSE_FALSE_FALSE_FALSE")
 \overline{\text{HNRNPU.TMT}} \leftarrow \text{read.delim}("/\text{Users/lducoli/Documents/Postdoc/PD\_Projects/3\_irCLIP\_RNP/MS/IMT\_range}) 
analysis/3_DEP/hnU_TMT_twofactor_results_sub.txt", header = TRUE)
HNRNPU.TMT <- subset (HNRNPU.TMT, order_sign != "FALSE_FALSE_FALSE")
genes <- unique(c(FUS.TMI$gene, HNRNPC.TMI$gene, HNRNPM.TMI$gene, HNRNPU.TMI$gene))
heatmap_hm <- heatmap_hm[, colnames(heatmap_hm) %in% genes]
heatmap hm <- heatmap hm[length(rownames(heatmap hm)):1,order(colSums(-heatmap hm, na.rm=TRUE))]
rownames (heatmap_hm) [3] <- "FUS_low"
annotation_col <- data frame(section = rep(c("high", "medium", "low"), length(rownames(heatmap_hm
    ))/3))
rownames(annotation_col) <- rownames(heatmap_hm)
# Heatmap color range
my. breaks \langle -c(\text{seg}(0, 50, \text{by}=0.01)) \#, \text{seg}(60, 150, \text{by}=1))
my.colors <- c("white", rev(paletteer_c("grDevices::YlGnBu", length(my.breaks)-1)))
#Heatmap
{\tt heatmap\_ame} \mathrel{<\!\!\!-} {\tt pheatmap} (
                     = heatmap hm,
  border_color = "black",
  annotation row = annotation col,
  cellwidth = 6,
  cellheight = 6,
  fontsize number=5.5,
  color = my. colors,
  breaks = my.breaks,
                    = TRUE,
  show colnames
  show_rownames
                     = TRUE,
  drop_levels
                     = TRUE,
  fontsize
                     = 5.5,
  cluster_rows
                     = FALSE,
  cluster cols
                     = FALSE,
  gaps_row = c(3,6,9,12,15,18,21,24),
  na_col = "lightgrey'
heatmap_ame
```



```
pdf("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/5_motif_analysis/3_pvalue_0.05_fc1/Heatmap_AME_bothsides.pdf", width = 10, height = 4) heatmap_ame dev.off()
```

#6. Density plot between RBP-RDAP nearest peaks We reasoned that the logFC calculated by DEWSeq in the regions should be greater in the higher regions for the RDAPs that we know are coming in the higher subzones.

```
all <- read.delim("/Users/lducoli/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/3_Categorization/3_pvalue_0.05_fc1/All_RBP_categorization.txt", header = TRUE)
all$rbp_category <- paste(all$rbp, all$category_all, sep = "_")
HNRNPC.dds <- readRDS("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/1_Regions_DEW-Seq/deseq2_norm/1_Subzone_separate/3_pvalue_0.05_fc1/HNRNPC_DEWseq_res.rds")

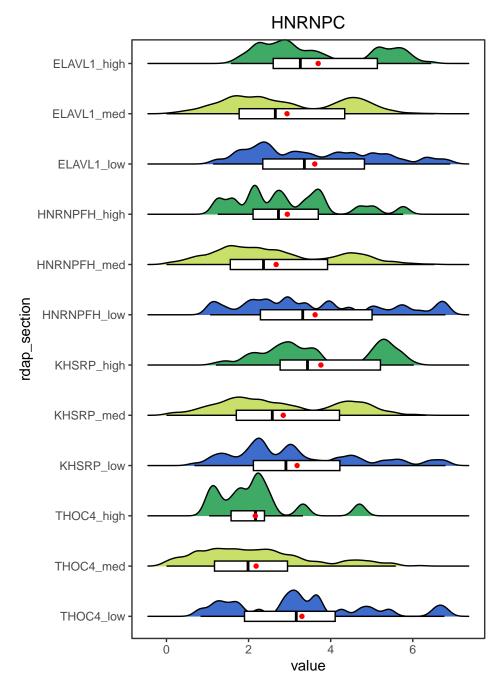
#Subset
h <- all %% filter(category_all = "high")
m <- all %% filter(category_all = "med")
l <- all %% filter(category_all = "low")

#Generate granges
h.gr <- GRanges(h[,grep("chr_S3", colnames(h))],
```

```
IRanges(start=h[,grep("begin_S3", colnames(h))], end=h[,grep("end_S3",
        colnames(h))]),
                                               names = h[,grep("regionID", colnames(h))], strand = h[,grep("strand_S3")
        ", colnames(h))],
                                          section = h[,grep("category\_all", colnames(h))], \ rbp\_category = h[,grep("category", grep("category")], \ rbp\_category = h[,grep("category")], \ rbp\_category = h[,grep("ca
        rbp_category", colnames(h))],
                                           \max_{fc} = h[, grep("\max_{S3} fc", colnames(h))])
h.gr \leftarrow gr.mid(h.gr)
m. gr <- GRanges(m[, grep("chr_S2", colnames(m))],
                                          IRanges(start=m[,grep("begin_S2", colnames(m))], end=m[,grep("end_S2",
        colnames(m))]),
                                          names = m[,grep("regionID", colnames(m))], strand = m[,grep("strand_S2",
       colnames(m))],
                                          section = m[,grep("category_all", colnames(m))], rbp_category = m[,grep("
        rbp category", colnames(m))],
                                           \max fc = m[, grep("max S2 fc", colnames(m))])
m. gr \leftarrow gr. mid(m. gr)
l.gr <- GRanges(l[,grep("chr_S1", colnames(l))],
                                          IRanges(start=l[,grep("begin_S1", colnames(l))], end=l[,grep("end_S1",
        colnames(1))]),
                                          names = l[,grep("regionID", colnames(1))], strand = l[,grep("strand_S1",
        colnames(1))],
                                          section = 1 [, grep("category_all", colnames(1))], rbp_category = 1 [, grep("
        rbp_category", colnames(l))],
                                           \max_{fc} = l [, grep("\max_{S1_{fc}}", colnames(1))])
l.gr \leftarrow gr.mid(l.gr)
logfc_table <- function(rbp, rdap, section1, section2, section3, gr1, gr2, gr3) {</pre>
distance_genomic <- function(rbp, rdap, section, gr) {
RBP_RDAP.dist <- gr.dist(gr[gr$rbp_category == paste(rbp, section, sep = "_")], gr[gr$rbp_
        category == paste(rdap, section, sep = "_")], ignore.strand=FALSE)
rownames(RBP_RDAP.dist) <- gr[gr$rbp_category == paste(rbp, section, sep = "_")]$names
colnames (RBP_RDAP.dist) <- gr[gr$rbp_category == paste(rdap, section, sep = "_")]$names
RBP_RDAP. dist <- as.data.frame(RBP_RDAP.dist)
RBP RDAP. dist$RBP bdgsites <- rownames(RBP RDAP. dist)
rownames (RBP_RDAP. dist) <- NULL
RBP_RDAP.dist <- RBP_RDAP.dist[c(length(gr[gr$rbp_category == paste(rdap, section, sep = "_")]$
        names) +1,1:length(gr[gr$rbp_category == paste(rdap, section, sep = "_")]$names))]
dist <- list()
for(i in 1:length(RBP_RDAP.dist$RBP_bdgsites)){
    gene \leftarrow sapply(strsplit(sapply(strsplit(RBP_RDAP.dist$RBP_bdgsites[i], "_"), function(x) x[5]),
         ":"), function(x) x[1])
    data\_columns \leftarrow c(1, grep(gene, colnames(RBP\_RDAP.dist)))
    if ( length (data_columns) == 1 ) {
        dist[[i]] <- NA
    } else {
    a <- RBP_RDAP. dist[, data_columns]
   b <- data.frame(dist=t(a[i,]), RBP_bdgsites = RBP_RDAP.dist$RBP_bdgsites[i])
    b < -b[-1,]
   b$RDAP_bdgsites <- rownames(b)
   rownames(b) <- NULL
    colnames(b)[1] <- "dist"
   b \leftarrow b[,c(3,1:2)]
    dist[[i]] <- b
    }
RDAP <- do.call(rbind, dist)
RDAP <- subset (RDAP, dist != "NA")
RDAP$RBP bdgsites <- factor(RDAP$RBP bdgsites)
RDAP$dist <- as.numeric(RDAP$dist)
RDAP$logdist <- log10(RDAP$dist)
RDAP$kbdist <- RDAP$dist/1000
```

```
RDAP$logdist[is.infinite(RDAP$logdist)] <- 0
RDAP$section <- section
RDAP <- RDAP %% arrange(dist, RDAP_bdgsites) %% group_by(RDAP_bdgsites) %% dplyr::slice(1)
RDAP <- RDAP %% arrange(dist, RBP bdgsites) %% group by(RBP bdgsites) %% dplyr::slice(1)
return(RDAP)
RDAP_dist_high <- distance_genomic(rbp, rdap, section1, gr1)
RDAP_dist_med <- distance_genomic(rbp, rdap, section2, gr2)
RDAP_dist_low <- distance_genomic(rbp, rdap, section3, gr3)
RDAP_dist_high <- RDAP_dist_high %% filter(dist < 250)
RDAP dist med <- RDAP dist med %% filter (dist < 150)
RDAP_dist_low <- RDAP_dist_low %% filter (dist < 75)
RDAP high <- gr1[gr1$rbp category == paste(rbp, section1, sep = "")]
RDAP_high <- RDAP_high [RDAP_high $names %in% RDAP_dist_high $RBP_bdgsites]
RDAP med <- gr2[gr2$rbp category == paste(rbp, section2, sep = "")]
RDAP_med <- RDAP_med [RDAP_med$names %in% RDAP_dist_med$RBP_bdgsites]
RDAP_low <- gr3[gr3$rbp_category == paste(rbp, section3, sep = "_")]
RDAP_low <- RDAP_low RDAP_low RDAP_low names %in% RDAP_dist_low RBP_bdgsites
RBP\_RDAP < - \ data.frame (names = c (RDAP\_high\$names, RDAP\_med\$names, RDAP\_low\$names), \ value = c (RDAP\_high\$names, RDAP\_med\$names), \ value = c (RDAP\_med\$names, RDAP\_med\$names), \ value = c
          \label{lowsmax_fc} $$ _{\rm high\max_fc} \ , \ RDAP\_low\max_fc} \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(section1\ , \ length(RDAP\_high\$)) \ , \ section = c(rep(
          max fc)), rep(section2, length(RDAP med$max fc)), rep(section3, length(RDAP low$max fc))),
          rdap = rep(rdap, length(c(RDAP\_high\$max\_fc, RDAP\_med\$max\_fc, RDAP\_low\$max\_fc))))
return (RBP_RDAP)
#If you want to run the code you need to uncomment the two paragraph below (it takes around 10min
# RBP_RDAP1 <- logfc_table("HNRNPC", "KHSRP", "high", "med", "low", h.gr, m.gr, l.gr)
# RBP_RDAP2 <- logfc_table("HNRNPC", "ELAVL1", "high", "med", "low", h.gr, m.gr, l.gr)
# RBP_RDAP3 <- logfc_table("HNRNPC", "HNRNPFH", "high", "med", "low", h.gr, m.gr, l.gr)
# RBP_RDAP4 <- logfc_table("HNRNPC", "THOC4", "high", "med", "low", h.gr, m.gr, l.gr)
# saveRDS(RBP_RDAP1, "~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/6_LogFC_plot/3_
          pvalue_0.05_fc1/HNRNPC_KHSRP_logFCtable.rds")
# saveRDS(RBP_RDAP2, "~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/6_LogFC_plot/3_
          pvalue_0.05_fc1/HNRNPC_ELAVL1_logFCtable.rds")
# saveRDS(RBP_RDAP3, "~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/6_LogFC_plot/3_
          pvalue_0.05_fc1/HNRNPC_HNRNPFH_logFCtable.rds")
# saveRDS(RBP_RDAP4, "~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/6_LogFC_plot/3_
          pvalue 0.05 fc1/HNRNPC THOC4 logFCtable.rds")
RBP_RDAP1 <- readRDS("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/6_LogFC_plot/3_
          pvalue 0.05 fc1/HNRNPC KHSRP logFCtable.rds")
RBP_RDAP2 <- readRDS("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/6_LogFC_plot/3_
          pvalue 0.05 fc1/HNRNPC ELAVL1 logFCtable.rds")
RBP_RDAP3 <-- readRDS("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/6_LogFC_plot/3_
          pvalue_0.05_fc1/HNRNPC_HNRNPFH_logFCtable.rds")
RBP_RDAP4 <- readRDS("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/6_LogFC_plot/3_
          pvalue_0.05_fc1/HNRNPC_THOC4_logFCtable.rds")
RBP RDAP <- rbind (RBP RDAP1, RBP RDAP2, RBP RDAP3, RBP RDAP4)
RBP_RDAP$rdap_section <- paste(RBP_RDAP$rdap, RBP_RDAP$section, sep = "_")
write.table(RBP_RDAP, "~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/1_Subzones/6_LogFC_plot/3
         _pvalue_0.05_fc1/HNRNPC_LogFC_nearest.txt", quote = F, sep = "\t", row.names = F)
RBP_RDAP$rdap_section <- factor(RBP_RDAP$rdap_section, levels= c("THOC4_low", "THOC4_med", "THOC4_
          high", "KHSRP_low", "KHSRP_med", "KHSRP_high",
                                                                                                                                                          "HNRNPFH low", "HNRNPFH med", "
         HNRNPFH_high",
                                                                                                                                                          "ELAVL1_low", "ELAVL1_med","
         ELAVL1 high"))
```

```
logFCplot \leftarrow ggplot(RBP\_RDAP, \ aes(x = value \ , \ y = rdap\_section \, , \ group = rdap\_section \, , \ fill = rdap \, , \
             section)) +
       stat_density_ridges(scale = .85,geom = "density_ridges_gradient",
                                                                         alpha = 0.7, bandwidth = 0.15) +
       geom_boxplot(aes(x = value, y = rdap_section), color = "black", inherit.aes = FALSE, width =
              0.2, outlier.shape = NA) +
       stat_summary(fun=mean, geom="point", shape=20, size=2, color="red", fill="red") +
       coord_cartesian(clip = "off") +
       scale_fill_manual(values = my_pal, guide = "none") +
       scale_y_discrete(expand = c(0.04, 0)) +
       ggtitle("HNRNPC") + theme_bw() +
       theme( panel.grid.major = element_blank(),
                              panel.grid.minor = element_blank(),
                              panel.background = element_blank(),
                              axis.line = element_blank(),
                              plot.title = element_text(hjust = 0.5))
logFCplot
```



```
RBP_RDAP$section <- factor(RBP_RDAP$rdap_section)
HNRNPC.res_aov <- aov(value ~ rdap_section, data = RBP_RDAP)
post_test <- glht(HNRNPC.res_aov, linfct = mcp(rdap_section = "Tukey"), alternative = "greater")
summary(post_test)
```

```
##
## Simultaneous Tests for General Linear Hypotheses
##
## Multiple Comparisons of Means: Tukey Contrasts
##
##
##
##
##
##
##
##
##
##
Fit: aov(formula = value ~ rdap_section , data = RBP_RDAP)
```

```
## Linear Hypotheses:
                                      Estimate Std. Error t value Pr(>t)
###
## THOC4 med - THOC4 low <= 0
                                     -1.115943
                                                  0.313655
                                                             -3.558 1.0000
## THOC4_high - THOC4_low <= 0
                                                  0.497519
                                     -1.134790
                                                             -2.281 1.0000
## KHSRP_low - THOC4_low <= 0
                                     -0.118089
                                                  0.346453
                                                             -0.341 1.0000
## KHSRP_med - THOC4_low <= 0
                                                             -1.484 1.0000
                                     -0.454068
                                                  0.305993
## KHSRP high - THOC4 low <= 0
                                      0.462354
                                                  0.339789
                                                              1.361 0.8827
## HNRNPFH_low - THOC4_low <= 0
                                      0.321144
                                                  0.379210
                                                              0.847 \ 0.9923
## HNRNPFH_med - THOC4_low <= 0
                                     -0.628176
                                                  0.306382
                                                             -2.050 1.0000
                                     -0.355435
## HNRNPFH_high - THOC4_low <= 0
                                                  0.439397
                                                             -0.809 \ 1.0000
## ELAVL1_low - THOC4_low <= 0
                                                              0.929 \ 0.9864
                                      0.312809
                                                  0.336831
\#\# ELAVL1\_med - THOC4\_low <= 0
                                     -0.363014
                                                  0.304602
                                                             -1.192 1.0000
## ELAVL1_high - THOC4_low <= 0
                                      0.395955
                                                  0.320828
                                                              1.234 0.9290
## THOC4_high - THOC4_med <= 0
                                                  0.402528
                                                             -0.047 1.0000
                                     -0.018847
## KHSRP_low - THOC4_med <= 0
                                      0.997854
                                                  0.185832
                                                              5.370
                                                                     < 0.01 ***
## KHSRP_med - THOC4_med <= 0
                                      0.661875
                                                  0.090197
                                                              7.338
                                                                     < 0.01 ***
## KHSRP high - THOC4 med <= 0
                                      1.578297
                                                  0.173091
                                                              9.118
                                                                     < 0.01 ***
## HNRNPFH_low - THOC4_med <= 0
                                      1.437087
                                                  0.241462
                                                              5.952
                                                                     < 0.01 ***
## HNRNPFH_med - THOC4_med <= 0
                                      0.487767
                                                  0.091506
                                                              5.330
                                                                     < 0.01 ***
## HNRNPFH_high - THOC4_med <= 0
                                      0.760508
                                                  0.327983
                                                              2.319\ 0.2706
## ELAVL1_low - THOC4_med \le 0
                                      1.428752
                                                  0.167209
                                                              8.545
                                                                     < 0.01 ***
## ELAVL1 med - THOC4 med <= 0
                                      0.752929
                                                  0.085360
                                                              8.821
                                                                     < 0.01 ***
## ELAVL1_high - THOC4_med \le 0
                                      1.511898
                                                  0.132038
                                                             11.450
                                                                     < 0.01 ***
## KHSRP_low - THOC4_high <= 0
                                      1.016702
                                                  0.428578
                                                              2.372 0.2418
## KHSRP med - THOC4 high <= 0
                                      0.680722
                                                  0.396587
                                                              1.716 \ 0.6762
## KHSRP_high - THOC4_high <= 0
                                      1.597145
                                                  0.423209
                                                              3.774
                                                                    < 0.01 **
## HNRNPFH_low - THOC4_high <= 0
                                      1.455934
                                                  0.455466
                                                              3.197 0.0271 *
## HNRNPFH_med - THOC4_high <= 0
                                      0.506614
                                                  0.396886
                                                              1.276 0.9149
## HNRNPFH high - THOC4 high <= 0
                                      0.779356
                                                  0.506674
                                                              1.538 \ 0.7922
## ELAVL1_low - THOC4_high <= 0
                                      1.447599
                                                  0.420838
                                                              3.440 0.0125 *
## ELAVL1 med - THOC4 high <= 0
                                      0.771776
                                                  0.395515
                                                              1.951 0.5074
## ELAVL1 high - THOC4 high <= 0
                                      1.530745
                                                  0.408142
                                                              3.751 < 0.01
## KHSRP_med - KHSRP_low <= 0
                                     -0.335980
                                                  0.172587
                                                             -1.947 1.0000
                                                  0.227181
## KHSRP_high - KHSRP_low <= 0
                                      0.580443
                                                              2.555 \ 0.1603
## HNRNPFH_low - KHSRP_low <= 0
                                                              1.553 \ 0.7825
                                      0.439232
                                                  0.282762
## HNRNPFH med - KHSRP low <= 0
                                     -0.510087
                                                  0.173274
                                                             -2.944 \ 1.0000
## HNRNPFH_high - KHSRP_low <= 0
                                     -0.237346
                                                  0.359477
                                                             -0.660 1.0000
\#\# ELAVL1\_low - KHSRP\_low <= 0
                                      0.430898
                                                  0.222732
                                                              1.935 0.5198
## ELAVL1\_med - KHSRP\_low <= 0
                                     -0.244926
                                                  0.170109
                                                             -1.440 1.0000
## ELAVL1_high - KHSRP_low <= 0
                                      0.514044
                                                  0.197698
                                                              2.600 0.1441
## KHSRP_high - KHSRP_med <= 0
                                      0.916422
                                                  0.158786
                                                              5.771 < 0.01 ***
## HNRNPFH_low - KHSRP_med <= 0
                                      0.775212
                                                  0.231422
                                                              3.350 0.0162
## HNRNPFH_med - KHSRP_med <= 0
                                     -0.174108
                                                  0.060215
                                                             -2.891 1.0000
## HNRNPFH_high - KHSRP_med <= 0
                                      0.098634
                                                  0.320664
                                                              0.308 1.0000
## ELAVL1_low - KHSRP_med <= 0
                                      0.766877
                                                  0.152353
                                                              5.034 < 0.01
## ELAVL1 med - KHSRP med <= 0
                                      0.091054
                                                  0.050387
                                                              1.807 0.6111
\#\# ELAVL1\_high - KHSRP\_med <= 0
                                      0.850023
                                                  0.112635
                                                              7.547 < 0.01 ***
## HNRNPFH_low - KHSRP_high <= 0
                                     -0.141210
                                                  0.274557
                                                             -0.514 \ 1.0000
## HNRNPFH_med - KHSRP_high <= 0
                                     -1.090530
                                                  0.159533
                                                             -6.836 1.0000
## HNRNPFH high - KHSRP high <= 0
                                                             -2.316 1.0000
                                     -0.817789
                                                  0.353059
\#\# ELAVL1\_low - KHSRP\_high <= 0
                                                  0.212218
                                                             -0.705 1.0000
                                     -0.149545
\#\# ELAVL1\_med - KHSRP\_high <= 0
                                     -0.825368
                                                  0.156089
                                                             -5.288 \ 1.0000
## ELAVL1 high - KHSRP high <= 0
                                     -0.066399
                                                  0.185773
                                                             -0.357 1.0000
                                                             -4.093 1.0000
## HNRNPFH_med - HNRNPFH_low <= 0
                                     -0.949320
                                                  0.231936
## HNRNPFH_high - HNRNPFH_low <= 0
                                                  0.391144
                                                             -1.730 1.0000
                                    -0.676578
## ELAVL1_low - HNRNPFH_low <= 0
                                     -0.008335
                                                  0.270887
                                                             -0.031 1.0000
## ELAVL1_med - HNRNPFH_low <= 0
                                     -0.684158
                                                  0.229581
                                                             -2.980 1.0000
## ELAVL1_high - HNRNPFH_low <= 0
                                      0.074811
                                                  0.250709
                                                              0.298 \ 1.0000
                                                              0.850 \ 0.9922
## HNRNPFH_high - HNRNPFH_med <= 0
                                      0.272741
                                                  0.321035
## ELAVL1 low - HNRNPFH med <= 0
                                      0.940985
                                                  0.153131
                                                              6.145
                                                                    < 0.01 ***
## ELAVL1 med - HNRNPFH med <= 0
                                      0.265162
                                                  0.052695
                                                              5.032
                                                                     < 0.01 ***
## ELAVL1_high - HNRNPFH_med <= 0
                                                                    <0.01 ***
                                      1.024131
                                                  0.113686
                                                              9.008
## ELAVL1_low - HNRNPFH_high <= 0
                                      0.668244
                                                  0.350213
                                                              1.908 0.5384
## ELAVL1_med - HNRNPFH_high <= 0
                                     -0.007580
                                                  0.319338
                                                             -0.024 1.0000
## ELAVL1_high - HNRNPFH_high <= 0
                                      0.751390
                                                  0.334849
                                                              2.244 \ 0.3110
\#\# ELAVL1\_med - ELAVL1\_low <= 0
                                     -0.675823
                                                  0.149540
                                                             -4.519 1.0000
## ELAVL1_high - ELAVL1_low \le 0
                                      0.083146
                                                  0.180305
                                                              0.461 0.9998
```

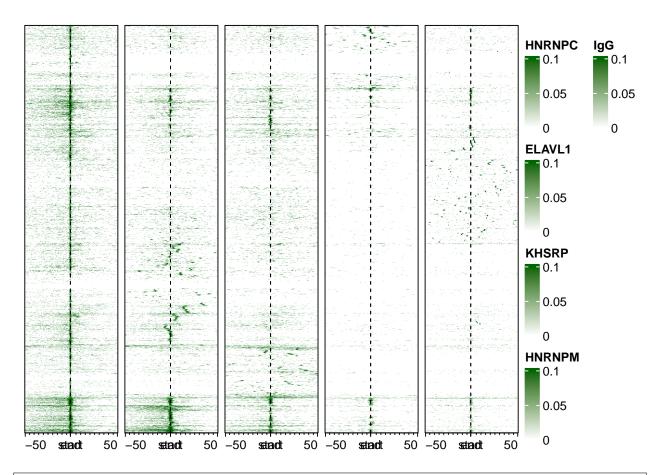
```
## ELAVL1_high - ELAVL1_med <= 0  0.758969  0.108800  6.976  <0.01 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## (Adjusted p values reported --- single-step method)
```

#7. Re-CLIP integration Here, we integrated the results from the Re-CLIP experiment where we sequentially pull down HNRNPC and several RDAPs followed by RNA-seq. The experiment was performed in HEK293T. These data went through the same custom snakemake pipeline found here: link.

```
#Load dds from DEWSeq
HNRNPC. bdg <- read.delim("~/Documents/Postdoc/PD Projects/3 irCLIP-RNP/Seq/1 Subzones/1 Regions
    DEW-Seq/deseq2_norm/2_Subzone_sum/3_pvalue_0.05_fc1/HNRNPC_DEWSeq_sign_regions_sum_
     categorization.txt", header = TRUE)
HNRNPC. bdg$rbp <- "HNRNPC"
colnames (HNRNPC.bdg) [5] <- "region_strand"
HNRNPC. bdg <- subset (HNRNPC. bdg, chromosome != "chrM")
data <- HNRNPC.bdg
rbp <- "HNRNPC"
#Prepare granges
p <- subset(data, region_strand == "+" & chromosome != "chrM")
gr.p <- GRanges(p[,grep("chr_S1", colnames(p))],
                     IRanges(start=p[,grep("begin_S1", colnames(p))],
                               end=p[,grep("end_S1", colnames(p))]),
                     names = p[,grep("max_S1_id", colnames(p))],
strand = p[, grep("strand_S1", colnames(p))],
category = p[, grep("category", colnames(p))])
m <- subset(data, region_strand == "-" & chromosome != "chrM")
\operatorname{gr.m} < - \operatorname{GRanges}(m[\,,\operatorname{grep}(\,\text{"chr\_S1"}\,,\ \operatorname{colnames}(m)\,)\,]\,,
                     IRanges(start=m[,grep("begin_S1", colnames(m))],
                     end=m[,grep("end_S1", colnames(m))]),
names = m[,grep("max_S1_id", colnames(m))],
strand = m[,grep("strand_S1", colnames(m))],
category = m[,grep("category", colnames(m))])
start(gr.p) \leftarrow (start(gr.p)+end(gr.p))/2
end(gr.p) \leftarrow start(gr.p)
start(gr.m) \leftarrow (start(gr.m)+end(gr.m))/2
end(gr.m) <- start(gr.m)
#Load bigwig files
setwd("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/Seq/2_293T_reclip/1_Bigwig_sum")
bw.ELAVL1.r1.p <- import.bw("HuR_293T-hnC_110-350_R1.p.scaleddewseq.bw")
bw.ELAVL1.r1.m <- import.bw("HuR 293T-hnC 110-350 R1.m.scaleddewseg.bw")
bw.ELAVL1.r2.p <- import.bw("HuR_293T-hnC_110-350_R2.p.scaleddewseq.bw")
bw.ELAVL1.r2.m <- import.bw("HuR_293T-hnC_110-350_R2.m.scaleddewseq.bw")
bw.IgG.r1.p <- import.bw("IgG_A431-hnC_EGF0_R1.p.scaleddewseq.bw")
bw.IgG.r1.m <- import.bw("IgG A431-hnC EGF0 R1.m.scaleddewseg.bw")
bw.IgG.r2.p <- import.bw("IgG_A431-hnC_EGF0_R2.p.scaleddewseq.bw")
bw. IgG. r2.m <- import.bw("IgG_A431-hnC_EGF0_R2.m. scaleddewseq.bw")
bw.KHSRP.rl.p <- import.bw("KHSRP_293T-hnC_140-350_Rl.p.scaleddewseq.bw")
bw.KHSRP.r1.m <- import.bw("KHSRP_293T-hnC_140-350_R1.m.scaleddewseq.bw")
bw. KHSRP. \ r2. \ p <- import.bw ("KHSRP\_293T-hnC\_140-350\_R2. \ p. scaleddewseq.bw")
bw.KHSRP.r2.m <- import.bw("KHSRP_293T-hnC_140-350_R2.m.scaleddewseq.bw")
```

```
\label{lower} bw. HNRNPC.r1.p <- import.bw("hnRNPC_293T-hnC_all_R1.p.scaleddewseq.bw") \\ bw. HNRNPC.r1.m <- import.bw("hnRNPC_293T-hnC_all_R1.m.scaleddewseq.bw") \\
bw.HNRNPC.r2.p <- import.bw("hnRNPC_293T-hnC_all_R1.p.scaleddewseq.bw")
bw.HNRNPC.r2.m <- import.bw("hnRNPC 293T-hnC all R1.m.scaleddewseg.bw")
bw.HNRNPM.rl.p <- import.bw("hnRNPM_293T-hnC_140-350_Rl.p.scaleddewseg.bw")
bw.HNRNPM.r1.m <- import.bw("hnRNPM_293T-hnC_140-350_R1.m. scaleddewseq.bw")
bw.HNRNPM.r2.p <- import.bw("hnRNPM 293T-hnC 140-350 R2.p.scaleddewseg.bw")
bw.HNRNPM.r2.m <- import.bw("hnRNPM_293T-hnC_140-350_R2.m.scaleddewseq.bw")
#Create profile matrices#
ELAVL1.mat1.p = normalizeToMatrix(bw.ELAVL1.r1.p, gr.p, value_column = "score", extend = 200,
    mean_mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
ELAVL1.mat1.m = normalizeToMatrix(bw.ELAVL1.r1.m, gr.m, value_column = "score", extend = 200,
    mean_mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
ELAVL1.mat2.p = normalizeToMatrix(bw.ELAVL1.r2.p, gr.p, value_column = "score", extend = 200,
    mean_mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
ELAVL1.mat2.m = normalizeToMatrix(bw.ELAVL1.r2.m, gr.m, value column = "score", extend = 200,
    mean_mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
IgG.matl.p = normalizeToMatrix(bw.IgG.rl.p, gr.p, value_column = "score", extend = 200, mean_mode
     = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
IgG.mat1.m = normalizeToMatrix(bw.IgG.r1.m, gr.m, value column = "score", extend = 200, mean mode
     = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
IgG.mat2.p = normalizeToMatrix(bw.IgG.r2.p, gr.p, value_column = "score", extend = 200, mean_mode
= "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
IgG.mat2.m = normalizeToMatrix(bw.IgG.r2.m, gr.m, value_column = "score", extend = 200, mean_mode
     = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
KHSRP.matl.p = normalizeToMatrix(bw.KHSRP.rl.p, gr.p, value column = "score", extend = 200, mean
    mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
KHSRP.mat1.m = normalizeToMatrix(bw.KHSRP.r1.m, gr.m, value_column = "score", extend = 200, mean_
    mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
KHSRP.mat2.p = normalizeToMatrix(bw.KHSRP.r2.p, gr.p, value_column = "score", extend = 200, mean_
    mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
KHSRP.mat2.m = normalizeToMatrix(bw.KHSRP.r2.m, gr.m, value_column = "score", extend = 200, mean_
    mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
HNRNPC.matl.p = normalizeToMatrix(bw.HNRNPC.rl.p, gr.p, value_column = "score", extend = 200,
    mean_mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
HNRNPC.mat1.m = normalizeToMatrix(bw.HNRNPC.r1.m, gr.m, value_column = "score", extend = 200,
    mean_mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
\label{eq:hnnpc} \begin{split} & \texttt{HNRNPC.mat2.p} = \texttt{normalizeToMatrix(bw.HNRNPC.r2.p}, \ \ \texttt{gr.p}, \ \ \texttt{value\_column} = \texttt{"score"}, \ \ \texttt{extend} = 200, \end{split}
    mean_mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
HNRNPC.mat2.m = normalizeToMatrix(bw.HNRNPC.r2.m, gr.m, value_column = "score", extend = 200,
    mean_mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
HNRNPM.matl.p = normalizeToMatrix(bw.HNRNPM.rl.p, gr.p, value_column = "score", extend = 200,
    mean\_mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
HNRNPM.mat1.m = normalizeToMatrix(bw.HNRNPM.r1.m, gr.m, value_column = "score", extend = 200,
    mean_mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
HNRNPM.mat2.p = normalizeToMatrix(bw.HNRNPM.r2.p, gr.p, value_column = "score", extend = 200,
    mean_mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
HNRNPM. mat2.m = normalizeToMatrix(bw.HNRNPM.r2.m, gr.m, value column = "score", extend = 200,
    mean_mode = "w0", w = 1, smooth = FALSE, background = 0, limit = NA)
ELAVL1.mat_list.p = list(ELAVL1.mat1.p, ELAVL1.mat2.p)
ELAVL1.mat.p = getSignalsFromList(ELAVL1.mat_list.p)
rownames(ELAVL1.mat.p) <- gr.p$names
IgG.mat\_list.p = list(IgG.mat1.p, IgG.mat2.p)
IgG.mat.p = getSignalsFromList(IgG.mat_list.p)
rownames(IgG.mat.p) <- gr.p$names
KHSRP.mat_list.p = list(KHSRP.mat1.p, KHSRP.mat2.p)
KHSRP.mat.p = getSignalsFromList(KHSRP.mat_list.p)
rownames(KHSRP.mat.p) <- gr.p$names
```

```
HNRNPC.mat_list.p = list(HNRNPC.mat1.p, HNRNPC.mat2.p)
HNRNPC.mat.p = getSignalsFromList(HNRNPC.mat_list.p)
rownames(\widehat{HNRNPC.mat.p}) <- gr.p\$names
HNRNPM.mat_list.p = list(HNRNPM.mat1.p, HNRNPM.mat2.p)
HNRNPM.mat.p = getSignalsFromList(HNRNPM.mat_list.p)
rownames (HNRNPM.mat.p) <- \ gr.p\$names
ELAVL1.mat list.m = list(ELAVL1.mat1.m, ELAVL1.mat2.m)
ELAVL1.mat.m = getSignalsFromList(ELAVL1.mat_list.m)
rownames (ELAVL1. mat.m) <- gr.m$names
IgG.mat\_list.m = list(IgG.mat1.m, IgG.mat2.m)
IgG.mat.m = getSignalsFromList(IgG.mat_list.m)
rownames (IgG.mat.m) <- gr.m$names
KHSRP.mat list.m = list(KHSRP.mat1.m, KHSRP.mat2.m)
KHSRP.mat.m = getSignalsFromList(KHSRP.mat_list.m)
rownames (KHSRP.mat.m) <- gr.m$names
HNRNPC.mat_list.m = list(HNRNPC.mat1.m, HNRNPC.mat2.m)
HNRNPC.mat.m = getSignalsFromList(HNRNPC.mat list.m)
rownames (HNRNPC.mat.m) <- gr.m\$names
HNRNPM. mat list.m = list (HNRNPM. mat1.m, HNRNPM. mat2.m)
HNRNPM.mat.m = getSignalsFromList(HNRNPM.mat_list.m)
rownames(HNRNPM.mat.m) \leftarrow gr.m$names
HNRNPC.mat <- rbind (HNRNPC.mat.p, HNRNPC.mat.m)
rownames(HNRNPC.mat) <- c(gr.p$names, gr.m$names)
KHSRP.mat <- rbind(KHSRP.mat.p, KHSRP.mat.m)
rownames\left(K\!H\!S\!R\!P.mat\right) < - \ c\left(\,gr.p\$names\,,\ gr.m\$names\right)
ELAVL1.mat <- rbind(ELAVL1.mat.p, ELAVL1.mat.m)
rownames (ELAVL1.mat) <- c(gr.p$names, gr.m$names)
HNRNPM.mat <- rbind (HNRNPM.mat.p, HNRNPM.mat.m)
rownames(HNRNPM.mat) < c(gr.p$names, gr.m$names)
IgG.mat <- rbind(IgG.mat.p, IgG.mat.m)</pre>
rownames(IgG.mat) <- c(gr.p$names, gr.m$names)
#Center on max nucleotide
s.sub.p \leftarrow HNRNPC.mat
get\_centered\_region <- \ function(x, \ y) \ \{
  mat = list()
  for(i in 1:length(rownames(x))){
    center \leftarrow which \max(x[i, 175:225])
    shift <-y[i, (175+center-50):(175+center+50)]
    mat[[i]] <- shift
  mat <- matrix(unlist(mat), ncol = length(rownames(x)))
  mat <- t (mat)
  rownames(mat) <- rownames(x)
  attr(mat, "upstream_index") = 1:49
attr(mat, "target_index") = 50
attr(mat, "downstream_index") = 51:101
attr(mat, "target_is_single_point") = TRUE
attr(mat, "extend") = c(50, 50)
  class(mat) = c("normalizedMatrix", "matrix")
  return (mat)
HNRNPC <- get_centered_region(s.sub.p, HNRNPC.mat)</pre>
KHSRP <- get_centered_region(s.sub.p, KHSRP.mat)
ELAVL1 <- get_centered_region(s.sub.p, ELAVL1.mat)
HNRNPM <- get_centered_region(s.sub.p, HNRNPM.mat)
IgG <- get_centered_region(s.sub.p, IgG.mat)
#Perform clustering
```



All the visualizations were saved as pdf and modified in illustrator.

sessionInfo()

```
## R version 4.2.1 (2022-06-23)
## Platform: x86_64-apple-darwin17.0 (64-bit)
## Running under: macOS Big Sur ... 10.16
## Matrix products: default
## BLAS:
           /Library/Frameworks/R. framework/Versions/4.2/Resources/lib/libRblas.0.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/4.2/Resources/lib/libRlapack.dylib
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/en_US.UTF-8
###
## attached base packages:
   [1] grid
                  stats4
                             stats
                                        graphics grDevices utils
                                                                        datasets
   [8] methods
## other attached packages:
   [1] GenomicFeatures_1.48.4
                                             {\tt multcomp\_1.4-25}
    [3]
        TH.\,\mathrm{data}\_1.1\!-\!2
                                             MASS_7.3 - 60.0.1
##
        survival 3.5-8
                                             mvtnorm 1.2-4
##
        circlize\_0.4.16
                                             gUtils\_0.2.0
##
##
    [9] data.table 1.15.2
                                             paletteer_1.6.0
   [11] BSgenome. Hsapiens. UCSC. hg38_1.4.4 BSgenome_1.64.0
##
   [13] Biostrings_2.66.0
                                             XVector_0.38.0
        magrittr_2.0.3
                                             universal motif\_1.14.1
   [17] memes_1.4.1
                                             caTools_1.18.2
   [19] rtracklayer 1.56.1
                                             EnrichedHeatmap 1.26.0
        gintools_0.1.3
   [21]
                                             AnnotationDbi_1.60.2
       Biobase_2.58.0
eulerr_7.0.1
                                             ggridges_0.5.6
    23
##
##
   [25]
                                             ChIPpeakAnno_3.30.1
        ComplexHeatmap\_2.14.0
   [27]
##
                                             gridExtra_2.3
   [29] Repitools_1.42.0
                                             plyranges_1.16.0
                                             GenomeInfoDb\_1.34.9
   [31] GenomicRanges_1.50.2
    [33]
        IRanges_2.32.0
                                             S4Vectors_0.36.2
   [35]
        BiocGenerics_0.44.0
                                             lubridate_1.9.3
   [37]
                                             \mathtt{stringr\_1.5.1}
##
        forcats\_1.0.0
   [39] dplyr_1.1.4
                                             purrr_1.0.2
   [41]
        readr_2.1.5
                                             tidyr_1.3.1
##
    [43]
        tibble_3.2.1
                                             tidyverse_2.0.0
        ggplot2_3.5.0
                                             scales_1.3.0
##
   [45]
##
   [47] RColorBrewer_1.1-3
                                             RIdeogram 0.2.2
##
   loaded via a namespace (and not attached):
###
##
     [1]
         utf8_1.2.4
                                        tidyselect_1.2.1
         RSQLite_2.3.5
                                        BiocParallel_1.32.6
##
      [3]
         munsell_0.5.0
##
      5
                                       codetools 0.2-19
##
         preprocessCore_1.60.2
                                        withr\_3.0.0
                                        filelock\_1.0.3
###
     [9]
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         highr_0.10
                                        knitr_1.45
##
    [11]
                                        {\tt labeling\_0.4.3}
##
    [13]
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##
    [15]
         MatrixGenerics 1.10.0
                                        GenomeInfoDbData 1.2.9
##
    [17]
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     19
                                        vctrs_0.6.5
         bit64_4.0.5
##
     21
         generics_0.1.3
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###
                                       timechange_0.3.0
##
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         regioneR_1.28.0
    [27]
                                       R6_2.5.1
##
         doParallel_1.0.17
                                       \mathtt{clue}\_0.3\!-\!65
##
     29]
##
    [31]
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                                        rsvg_2.6.0
```

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grImport2 0.3-1
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                                        cachem_1.0.8
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                                         Ringo_1.60.0
##
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##
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                                         gtable_0.3.4
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                                         affy_1.76.0
##
##
     43
         sandwich_3.1-0
                                        ensembldb_2.20.2
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                                        genefilter_1.78.0
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         GlobalOptions_0.1.2
##
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     [49]
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                                        prismatic_1.1.1
###
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         BiocManager\_1.30.22
                                        yaml_2.3.8
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         RBGL_1.72.0
                                         tools\_4.2.1
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##
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##
                                        {\tt zlibbioc\_1.44.0}
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##
         RCurl\_1.98-1.14
##
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                                        zoo_1.8-12
###
##
         SummarizedExperiment_1.28.0 cluster_2.1.6
         magick_2.8.2
###
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         truncnorm_1.0-9
                                         ProtGenerics_1.30.0
##
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     73
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##
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###
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##
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##
     81
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                                        biomaRt\_2.52.0
         KernSmooth\_2.23-22
                                        crayon_1.5.2
##
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##
    [85]
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###
    [87]
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     [89]
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                                         Matrix 1.6-5
                                         \overline{\text{bedtoolsr}}_2.30.0-4
    [93]
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###
         vsn_3.66.0
##
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                                        parallel 4.2.1
    [97]
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##
    [99]
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                                        InteractionSet\_1.24.0
###
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                                        annotate_1.76.0
         multtest_2.52.0
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                                        digest\_0.6.35
###
         graph 1.74.0
                                        polylabelr 0.2.0
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                                        edgeR\_3.40.2
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                                        \texttt{curl}\_5.2.1
                                         gtools_3.9.5
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         rjson_0.2.21
                                        lifecycle_1.0.4
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         lattice_0.22-5
                                        KEGGREST_1.38.0
##
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##
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##
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                                         rematch2_2.1.2
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###
   [129]
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                                        DESeq2_1.38.3
###
##
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                                        gsmoothr_0.1.7
```