

irCLIP-RNP dataset of species mixing experiment to determine contribution of mouse/human unique peptides

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This is the pipeline used to analyze the irCLIP-RNP datasets of species-mixing experiment. We have subjected to MS two gel sections ranging from 30 to 70kDa (named “free RNA ligation zone”) and from 70 to 350kDa (named “whole RNP zone”). irCLIP-RNP for HNRNPC was performed after mixing UVC human cells (HEK293T) and noUV mouse cells (3T3s). irCLIP-RNP for HNRNPA2B1 was performed with noUV HEK293T and UVC 3T3 cells.

1. Prepare the dataset

```
#Load the libraries
library(trackViewer)
library(dplyr)
library(httr)
library(org.Hs.eg.db)
library(org.Mm.eg.db)
library(paletteer)
library(ggplot2)
library(data.table)
library(ggExtra)
library(DEP2)
library(DESeq2)
library(ggpubr)
```

In the first step, we prepared the dataset to create a SummarizedExperiment object starting from the peptide.txt output file from MaxQuant.

```
# Load the peptides
data <- read.csv("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/irCLIP-RNP_control/1_Species_
mixing/0_Data/peptides.txt",
  sep = "\t")

# Remove RPL proteins
data <- data[!grep("RPL", data$Gene.names), ]

# Remove not detected peptides
data$rowsum <- rowSums(data[, grep("Intensity.", colnames(data),
  value = TRUE)])
data <- subset(data, rowsum != 0)
```

2. Create a SummarizedExperiment

We used the following design to create a SummarizedExperiment.

```
#Load design matrix
design <- read.delim("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/irCLIP-RNP_control/1_
Species_mixing/0_Data/Design_matrix.txt")
design
```

```
##          label          condition replicate
## 1 LFQ.intensity.BZ15A   HNRNPC_low         1
## 2 LFQ.intensity.BZ15B   HNRNPC_high        1
## 3 LFQ.intensity.BZ16A   HNRNPC_low         2
## 4 LFQ.intensity.BZ16B   HNRNPC_high        2
## 5 LFQ.intensity.BZ17A HNRNPA2B1_low        1
## 6 LFQ.intensity.BZ17B HNRNPA2B1_high        1
## 7 LFQ.intensity.BZ18A HNRNPA2B1_low        2
## 8 LFQ.intensity.BZ18B HNRNPA2B1_high        2
```

```
#Create a summarized experiment, filter, normalize, and impute
data$ID <- c(1:nrow(data))
data$name <- paste(data$Gene.names, data$ID, sep = "_")
ecols <- grep("Intensity.", colnames(data), value = TRUE)
pe <- make_pe(data, columns = ecols, expdesign = design)

#Filter, normalize, and impute
pe = filter_pe(pe, filter_formula = ~ Reverse != '+' & Potential.contaminant != "+" & Unique..
  Proteins. == "yes")
pe <- normalize_pe(pe, method = "vsr", i = "peptideRaw")
set.seed(13)
pe <- impute_pe(pe, fun = "GSimp", name = "peptideImp", i = "peptideNorm")
```

```
## Iteration 1 start...end!
## Iteration 2 start...end!
## Iteration 3 start...end!
## Iteration 4 start...end!
## Iteration 5 start...end!
## Iteration 6 start...end!
## Iteration 7 start...end!
## Iteration 8 start...end!
## Iteration 9 start...end!
## Iteration 10 start...end!
```

3. ROC analysis

We first performed a ROC analysis to compare UVC-human/noUV-mouse and noUV-human/UVC-mouse samples from the two gel sections.

```
#Prepare data.frame for ROC analysis
unique_pep <- assay(pe[[3]])
unique_pep <- as.data.frame(unique_pep)
rownames(unique_pep) <- pe[[3]]@elementMetadata$name

#Average between samples
```

```

unique_pep$HNRNPC_low <- rowMeans(unique_pep[c(1,3)])
unique_pep$HNRNPC_high <- rowMeans(unique_pep[c(2,4)])
unique_pep$HNRNPA2B1_low <- rowMeans(unique_pep[c(5,7)])
unique_pep$HNRNPA2B1_high <- rowMeans(unique_pep[c(6,8)])

#Subset mouse unique peptides
mouse_unique_pep <- unique_pep[grepl("[a-z]", rownames(unique_pep)), ]
mouse_unique_pep$Gene.names <- rownames(mouse_unique_pep)

mouse_HNRNPClow <- mouse_unique_pep[,c(13,9)]
colnames(mouse_HNRNPClow) <- c("Gene.names", "Int")

mouse_HNRNPChigh <- mouse_unique_pep[,c(13,10)]
colnames(mouse_HNRNPChigh) <- c("Gene.names", "Int")

mouse_HNRNPA2B1low <- mouse_unique_pep[,c(13,11)]
colnames(mouse_HNRNPA2B1low) <- c("Gene.names", "Int")

mouse_HNRNPA2B1high <- mouse_unique_pep[,c(13,12)]
colnames(mouse_HNRNPA2B1high) <- c("Gene.names", "Int")

#Subset human unique peptides
'%!in%' <- function(x,y){ '%in%'(x,y)}
human_unique_pep <- subset(unique_pep, rownames(unique_pep) %!in% rownames(mouse_unique_pep))
human_unique_pep$Gene.names <- rownames(human_unique_pep)
human_HNRNPClow <- human_unique_pep[,c(13,9)]
colnames(human_HNRNPClow) <- c("Gene.names", "Int")
human_HNRNPChigh <- human_unique_pep[,c(13,10)]
colnames(human_HNRNPChigh) <- c("Gene.names", "Int")
human_HNRNPA2B1low <- human_unique_pep[,c(13,11)]
colnames(human_HNRNPA2B1low) <- c("Gene.names", "Int")
human_HNRNPA2B1high <- human_unique_pep[,c(13,12)]
colnames(human_HNRNPA2B1high) <- c("Gene.names", "Int")

#Perform cumulative fraction analysis
i <- seq(0, 30, by = 0.01)

mouse_ECDF_HNRNPChigh <- data.frame(seq = i, ecdf = 1-(ecdf(mouse_HNRNPChigh$Int)(i)))
colnames(mouse_ECDF_HNRNPChigh) <- c("Seq_m", "ROC_m")
human_ECDF_HNRNPChigh <- data.frame(seq = i, ecdf = 1-(ecdf(human_HNRNPChigh$Int)(i)))
colnames(human_ECDF_HNRNPChigh) <- c("Seq_h", "ROC_h")
HNRNPChigh <- cbind(human_ECDF_HNRNPChigh, mouse_ECDF_HNRNPChigh)
HNRNPChigh$type <- "HNRNPC_HumanUVC_MousenoUV_high"

mouse_ECDF_HNRNPClow <- data.frame(seq = i, ecdf = 1-(ecdf(mouse_HNRNPClow$Int)(i)))
colnames(mouse_ECDF_HNRNPClow) <- c("Seq_m", "ROC_m")
human_ECDF_HNRNPClow <- data.frame(seq = i, ecdf = 1-(ecdf(human_HNRNPClow$Int)(i)))
colnames(human_ECDF_HNRNPClow) <- c("Seq_h", "ROC_h")
HNRNPClow <- cbind(human_ECDF_HNRNPClow, mouse_ECDF_HNRNPClow)
HNRNPClow$type <- "HNRNPC_HumanUVC_MousenoUV_low"

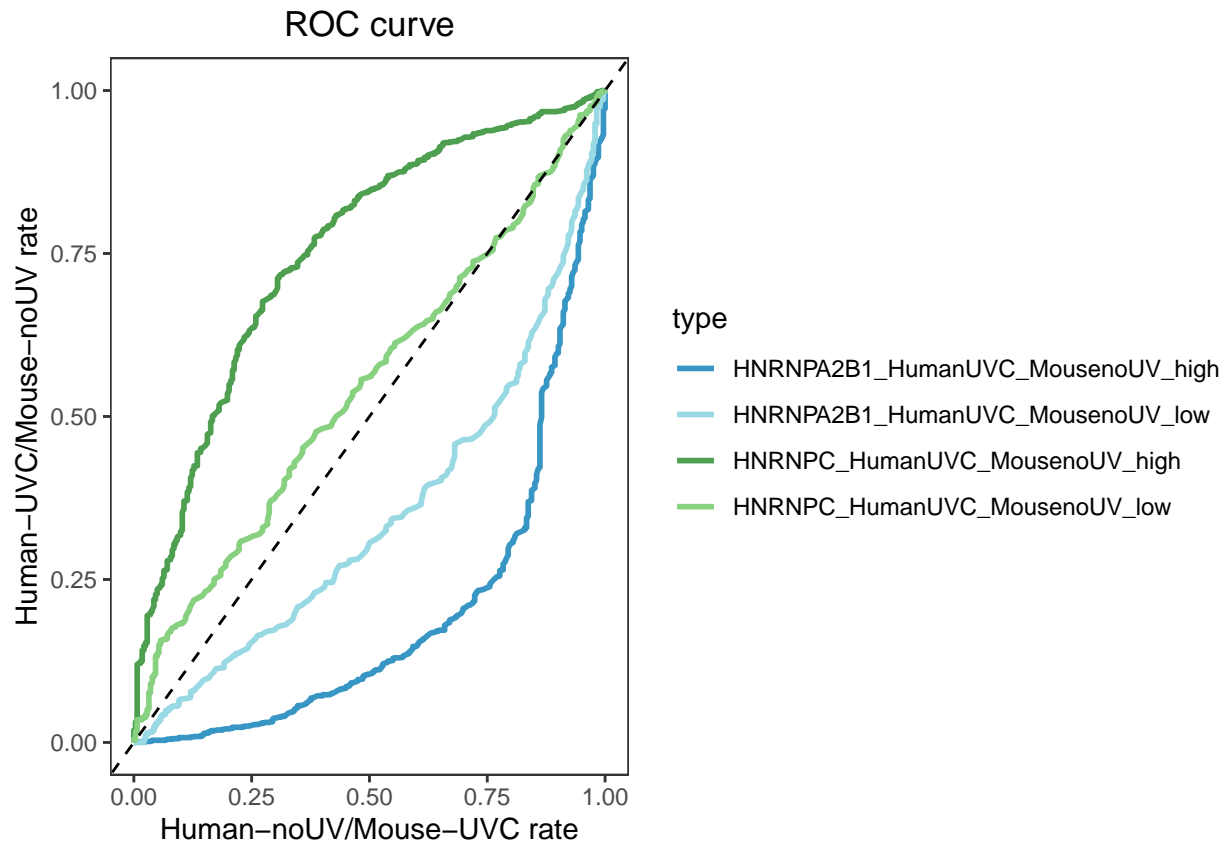
mouse_ECDF_HNRNPA2B1high <- data.frame(seq = i, ecdf = 1-(ecdf(mouse_HNRNPA2B1high$Int)(i)))
colnames(mouse_ECDF_HNRNPA2B1high) <- c("Seq_m", "ROC_m")
human_ECDF_HNRNPA2B1high <- data.frame(seq = i, ecdf = 1-(ecdf(human_HNRNPA2B1high$Int)(i)))
colnames(human_ECDF_HNRNPA2B1high) <- c("Seq_h", "ROC_h")
HNRNPA2B1high <- cbind(human_ECDF_HNRNPA2B1high, mouse_ECDF_HNRNPA2B1high)
HNRNPA2B1high$type <- "HNRNPA2B1_HumanUVC_MousenoUV_high"

mouse_ECDF_HNRNPA2B1low <- data.frame(seq = i, ecdf = 1-(ecdf(mouse_HNRNPA2B1low$Int)(i)))
colnames(mouse_ECDF_HNRNPA2B1low) <- c("Seq_m", "ROC_m")
human_ECDF_HNRNPA2B1low <- data.frame(seq = i, ecdf = 1-(ecdf(human_HNRNPA2B1low$Int)(i)))
colnames(human_ECDF_HNRNPA2B1low) <- c("Seq_h", "ROC_h")
HNRNPA2B1low <- cbind(human_ECDF_HNRNPA2B1low, mouse_ECDF_HNRNPA2B1low)
HNRNPA2B1low$type <- "HNRNPA2B1_HumanUVC_MousenoUV_low"

#Create a dataframe for plotting
roc <- rbind(HNRNPChigh, HNRNPClow, HNRNPA2B1high, HNRNPA2B1low)

```

```
ggplot(roc %>% arrange(ROC_h, ROC_m), aes(x = ROC_m, y = ROC_h, color = type)) + geom_line(
  linewidth = 1) +
  geom_abline(intercept = 0, slope = 1, linetype = "dashed") +
  scale_color_manual(values=c("#3896C4", "#98D9E4", "#4E9F50", "#87D180")) +
  theme_bw() +
  theme(panel.grid.major = element_blank(), #legend.position = "none",
        panel.grid.minor = element_blank(),
        panel.background = element_blank(),
        axis.line = element_blank(),
        plot.title = element_text(hjust = 0.5)) +
  ggtitle("ROC curve") +
  ylab("Human-UVC/Mouse-noUV rate") +
  xlab("Human-noUV/Mouse-UVC rate")
```



```
#Save ROC curve as pdf
pdf("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/irCLIP-RNP_control/1_Species_mixing/2_ROC/
  ROC_MouseHuman.pdf", width = 5, height = 5)
ggplot(roc %>% arrange(ROC_h, ROC_m), aes(x = ROC_m, y = ROC_h, color = type)) + geom_line(
  linewidth = 1) +
  geom_abline(intercept = 0, slope = 1, linetype = "dashed") +
  scale_color_manual(values=c("#3896C4", "#98D9E4", "#4E9F50", "#87D180")) +
  theme_bw() +
  theme(panel.grid.major = element_blank(), legend.position = "none",
        panel.grid.minor = element_blank(),
        panel.background = element_blank(),
        axis.line = element_blank(),
        plot.title = element_text(hjust = 0.5)) +
  ggtitle("ROC curve") +
  ylab("Human-UVC/Mouse-noUV rate") +
  xlab("Human-noUV/Mouse-UVC rate")
dev.off()
```

4. Distribution of mouse/Human unique peptide accross RBPs

At this point, we displayed the ratio of imputed intensity of mouse/human unique peptides across some RBPs in the Human-UVC/Mouse-noUV and Human-noUV/Mouse-UVC samples. We used only the intensities coming from the gel section referred here as “high” (a.k.a. whole RNP zone) ranging from 70-350kDa.

```
#Function to draw the lolliplot
get_lolliplot <- function(gene, species, taxid, orgDB){
  APIurl <- "https://www.ebi.ac.uk/proteins/api/"
  eid <- mget(gene, get(sub(".", "db", "SYMBOL2EG", orgDB))) [[1]]
  chr <- mget(eid, get(sub(".", "db", "CHR", orgDB))) [[1]]
  accession <- unlist(lapply(eid, function(.ele){mget(.ele, get(sub(".", "db", "UNIPROT", orgDB)))}))
  featureURL <- paste0(APIurl, "features?offset=0&size=1&reviewed=true", "&types=DNA_BIND%2CMOTIF%2C",
    "CDOMAIN", "&taxid=", taxid,
    "&accession=", paste(accession, collapse = "%2C"))
  response <- GET(featureURL)
  content <- httr::content(response)
  content <- content[[1]]
  acc <- content$accession
  sequence <- content$sequence
  gr <- GRanges(chr, IRanges(1, nchar(sequence)))
  domains <- do.call(rbind, content$features)
  domains <- GRanges(chr, IRanges(as.numeric(domains[, "begin"]), as.numeric(domains[, "end"]), names = domains[, "description"]))
  domains$fill <- 1+seq_along(domains)
  domains$height <- 0.04
  fasta = readLines(paste("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/irCLIP-RNP_control/1_
    Species_mixing/3_Intensity_plot/", gene, "_", species, "_sequence.fasta", sep = ""))
  ids = grepl(">", fasta)
  rbp.f = data.frame(id = sub(">", "", fasta[ids]), s = tapply(fasta[!ids], cumsum(ids)[!ids],
    function(x) {paste(x, collapse = "")}))
  rbp <- unique_pep[rownames(unique_pep) %like% gene, ]
  rbp$name <- rownames(rbp)
  rbp <- merge(rbp, data[c(37,84)], by = "name")
  SNP <- sort(rbp$Start.position)

  sample.gr <- GRanges(as.character(seqnames(domains)@values), IRanges(SNP, width=1, names=paste0(
    SNP)))
  total_rbp_int <- (2^((rbp$HNRNPC_high_1+rbp$HNRNPC_high_2)/2)+2^((rbp$HNRNPA2B1_high_1+rbp$
    HNRNPA2B1_high_2)/2))

  sample.gr$value1 <- 2^((rbp$HNRNPC_high_1+rbp$HNRNPC_high_2)/2)/total_rbp_int
  sample.gr$value2 <- 2^((rbp$HNRNPA2B1_high_1+rbp$HNRNPA2B1_high_2)/2)/total_rbp_int
  sample.gr$color <- rep(list(c("#87CEFA", "#98CE31")), length(SNP))
  sample.gr$border <- "gray30"

  features <- GRanges(as.character(seqnames(domains)@values), IRanges(c(1), width=nchar(rbp.f$s),
    names=paste0("block", 1)))
  features$height <- c(0.03)

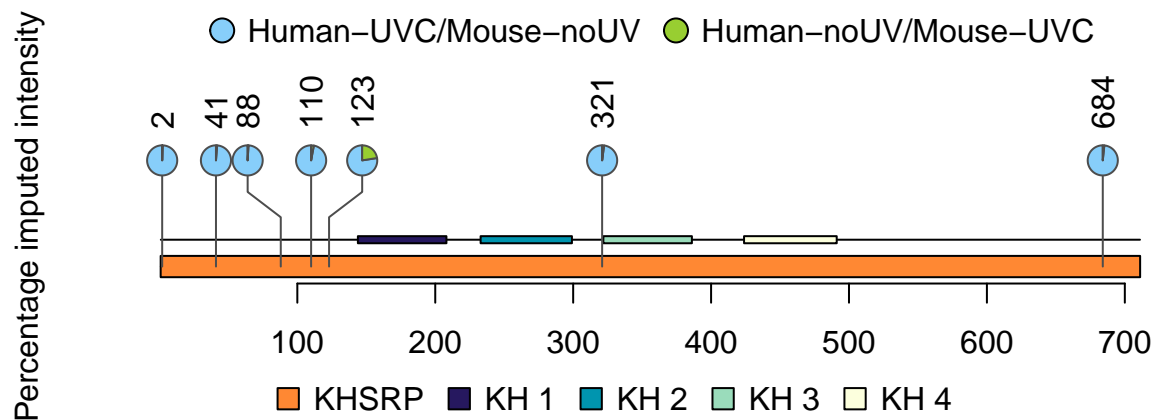
  legend <- list(labels=c("Human-UVC/Mouse-noUV", "Human-noUV/Mouse-UVC"), fill=c("#87CEFA", "#98
    CE31"))

  features.mul <- c(features, domains)
  features.mul$height[c(2:(length(domains)+1))] <- c(rep(0.01, length(domains)))
  features.mul$fill <- c("#FF8833", c(paletteer::c("grDevices::YlGnBu", (length(features.mul)-1))))
  features.mul$featureLayerID <- c("tx_1", paste("tx", rep(2, each=length(domains)), sep="_"))
  names(features.mul) <- c(gene, names(domains))
  lolliplot <- lolliplot(sample.gr, features.mul, legend = legend, ylab="Percentage imputed
    intensity", type="pie")
```

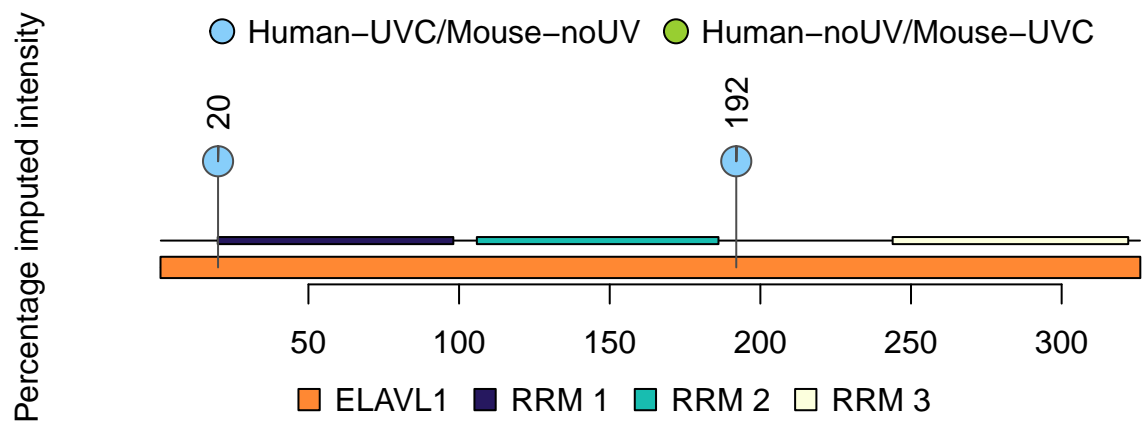
```
return(lolliplot)
}
```

Human RBPs

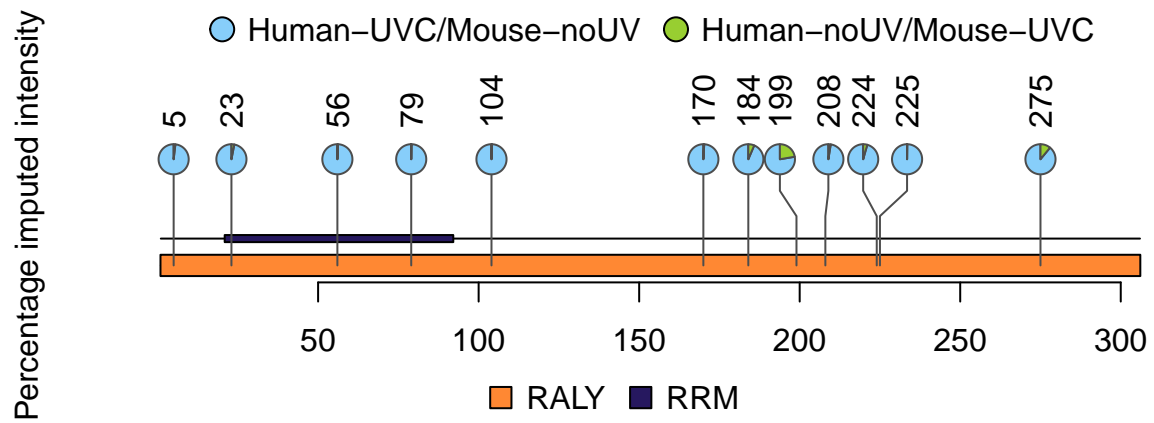
```
#Get lolliplot for human RBPs
KHSRP.h.lolliplot <- get_lolliplot("KHSRP", "Human", "9606", "org.Hs.eg.db")
```



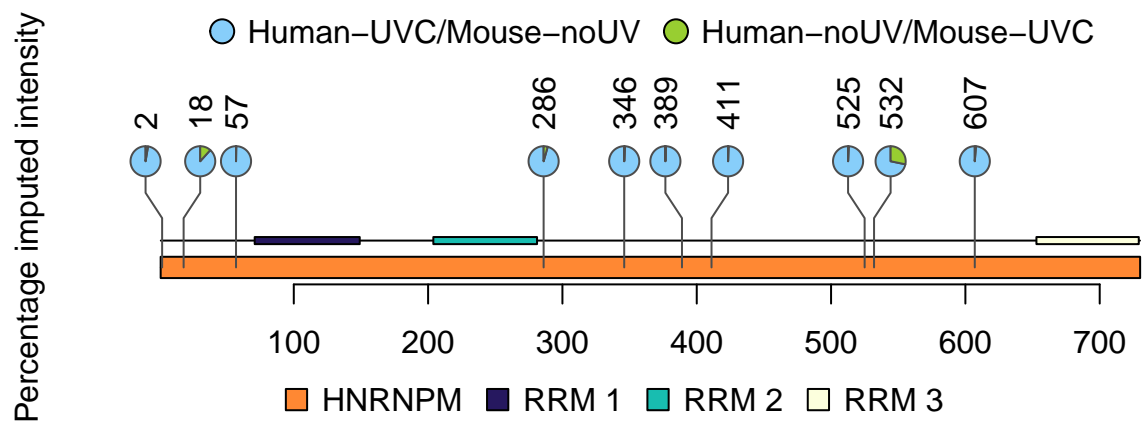
```
ELAVL1.h.lolliplot <- get_lolliplot("ELAVL1", "Human", "9606", "org.Hs.eg.db")
```



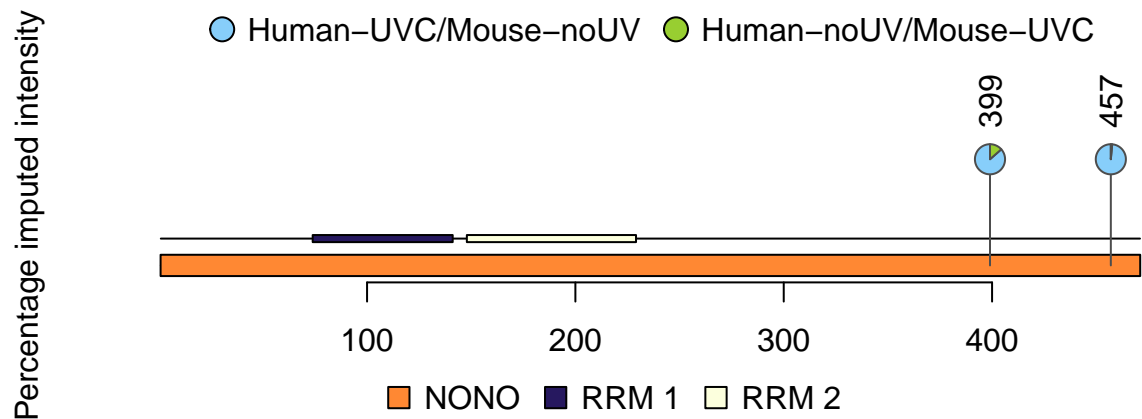
```
RALY.h.lollipopplot <- get_lollipopplot("RALY", "Human", "9606", "org.Hs.eg.db")
```



```
HNRNPM.h.lollipop <- get_lollipop("HNRNPM", "Human", "9606", "org.Hs.eg.db")
```

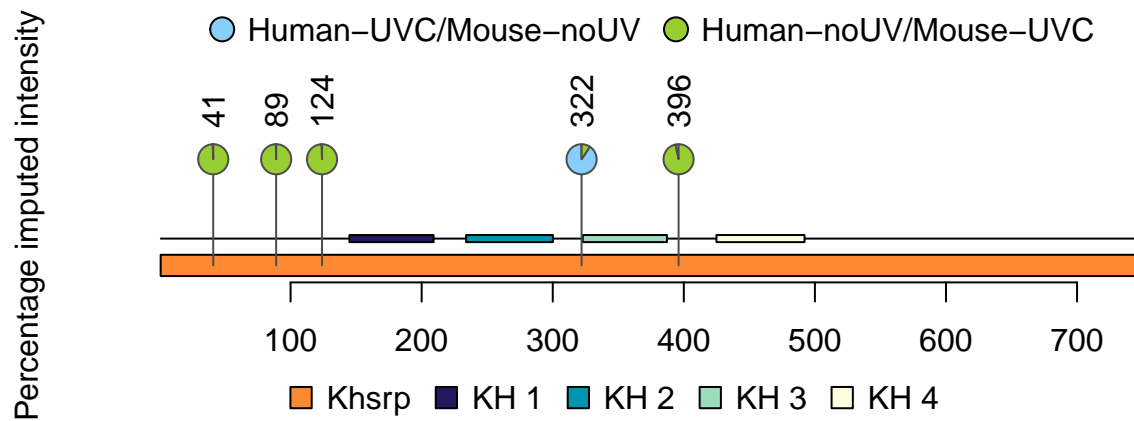
```
NONO.h.lollipopplot <- get_lollipopplot("NONO", "Human", "9606", "org.Hs.eg.db")
```



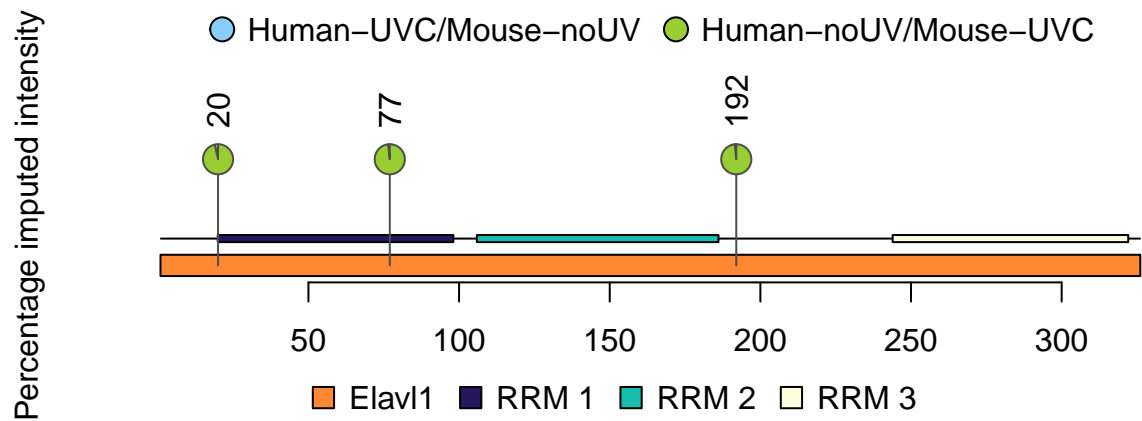
```
#Save human lolliplot as pdf
pdf("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/irCLIP-RNP_control/1_Species_mixing/3_
Intensity_plot/Human_Lolliplot.pdf")
KHSRP.h.lolliplot <- get_lolliplot("KHSRP", "Human", "9606", "org.Hs.eg.db")
ELAVL1.h.lolliplot <- get_lolliplot("ELAVL1", "Human", "9606", "org.Hs.eg.db")
RALY.h.lolliplot <- get_lolliplot("RALY", "Human", "9606", "org.Hs.eg.db")
HNRNPM.h.lolliplot <- get_lolliplot("HNRNPM", "Human", "9606", "org.Hs.eg.db")
NONO.h.lolliplot <- get_lolliplot("NONO", "Human", "9606", "org.Hs.eg.db")
dev.off()
```

Mouse RBPs

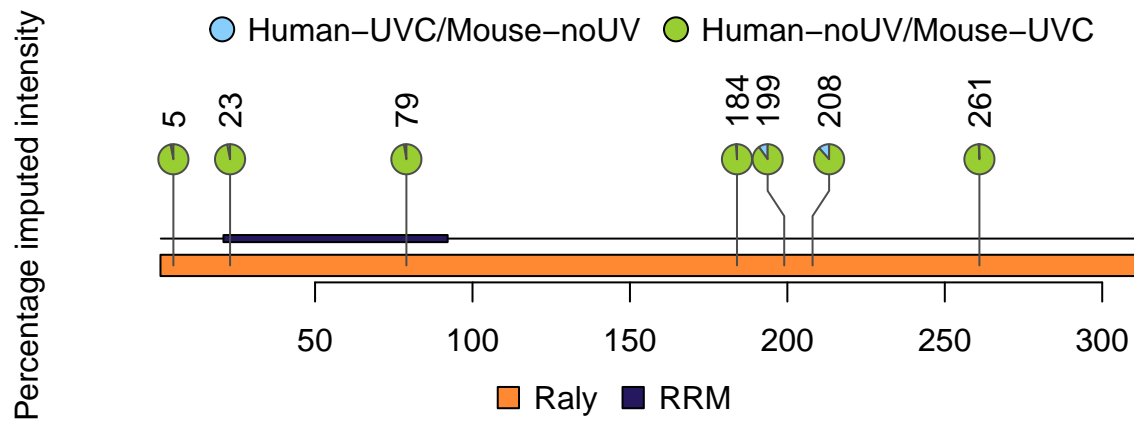
```
#Get lolliplot for mouse RBPs
Khsrp.m.lolliplot <- get_lolliplot("Khsrp", "Mouse", "10090", "org.Mm.eg.db")
```



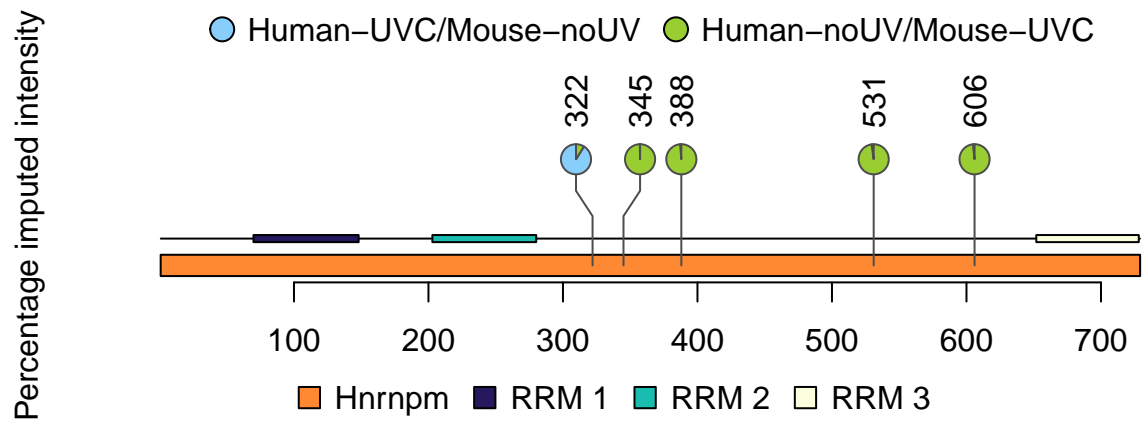
```
Elavl1.m.lollipopplot <- get_lollipopplot("Elavl1", "Mouse", "10090", "org.Mm.eg.db")
```



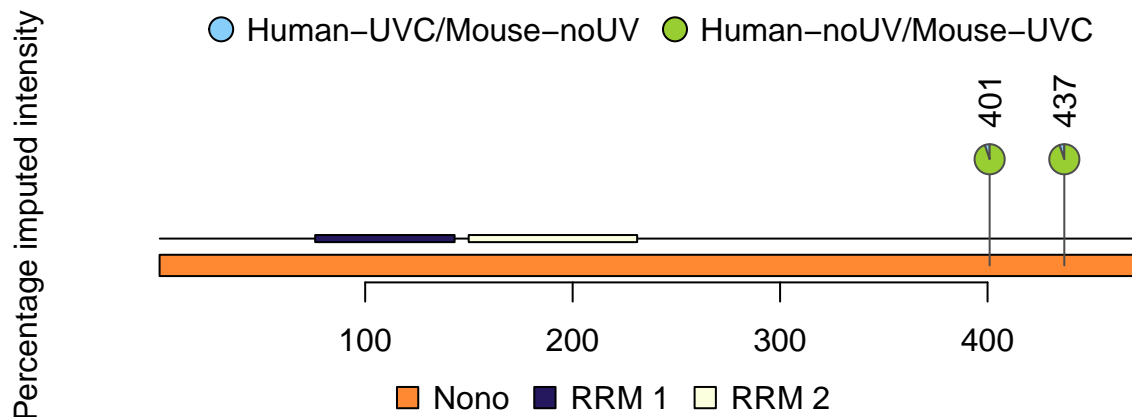
```
Raly.m.lollipopplot <- get_lollipopplot("Raly", "Mouse", "10090", "org.Mm.eg.db")
```



```
Hnrnpm.m.lollipopplot <- get_lollipopplot("Hnrnpm", "Mouse", "10090", "org.Mn.eg.db")
```



```
Nono.m.lollipop <- get_lollipop("Nono", "Mouse", "10090", "org.Mm.eg.db")
```



```
#Save mouse lolliplot as pdf
pdf("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/irCLIP-RNP_control/1_Species_mixing/3_
Intensity_plot/Mouse_Lolliplot.pdf")
Khsrp.m.lolliplot <- get_lolliplot("Khsrp", "Mouse", "10090", "org.Mm.eg.db")
Elavl1.m.lolliplot <- get_lolliplot("Elavl1", "Mouse", "10090", "org.Mm.eg.db")
Raly.m.lolliplot <- get_lolliplot("Raly", "Mouse", "10090", "org.Mm.eg.db")
Hnrnp.m.lolliplot <- get_lolliplot("Hnrnp", "Mouse", "10090", "org.Mm.eg.db")
Nono.m.lolliplot <- get_lolliplot("Nono", "Mouse", "10090", "org.Mm.eg.db")
dev.off()
```

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
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
##
## attached base packages:
## [1] grid stats4 stats graphics grDevices utils datasets
## [8] methods base
```

```

##
## other attached packages:
## [1] ggpubr_0.6.0 DESeq2_1.38.3
## [3] DEP2_0.4.8.24 R6_2.5.1
## [5] limma_3.54.2 MSnbase_2.24.2
## [7] ProtGenerics_1.30.0 mzR_2.32.0
## [9] MsCoreUtils_1.10.0 SummarizedExperiment_1.28.0
## [11] MatrixGenerics_1.10.0 matrixStats_1.2.0
## [13] ggExtra_0.10.1 data.table_1.15.2
## [15] ggplot2_3.5.0 paletteer_1.6.0
## [17] org.Mm.eg.db_3.15.0 org.Hs.eg.db_3.15.0
## [19] AnnotationDbi_1.60.2 Biobase_2.58.0
## [21] httr_1.4.7 dplyr_1.1.4
## [23] trackViewer_1.32.1 Rcpp_1.0.12
## [25] GenomicRanges_1.50.2 GenomeInfoDb_1.34.9
## [27] IRanges_2.32.0 S4Vectors_0.36.2
## [29] BiocGenerics_0.44.0
##
## loaded via a namespace (and not attached):
## [1] rappdirs_0.3.3 rtracklayer_1.56.1
## [3] tidyr_1.3.1 missForest_1.5
## [5] bit64_4.0.5 knitr_1.45
## [7] DelayedArray_0.24.0 rpart_4.1.23
## [9] KEGGREST_1.38.0 RCurl_1.98-1.14
## [11] AnnotationFilter_1.22.0 doParallel_1.0.17
## [13] generics_0.1.3 GenomicFeatures_1.48.4
## [15] preprocessCore_1.60.2 RSQlite_2.3.5
## [17] proxy_0.4-27 bit_4.0.5
## [19] xml2_1.3.6 httpuv_1.6.14
## [21] assertthat_0.2.1 TCseq_1.22.6
## [23] xfun_0.42 hms_1.1.3
## [25] evaluate_0.23 promises_1.2.1
## [27] fansi_1.0.6 restfulr_0.0.15
## [29] progress_1.2.3 dbplyr_2.4.0
## [31] Rgraphviz_2.40.0 igraph_2.0.3
## [33] DBI_1.2.2 geneplotter_1.76.0
## [35] htmlwidgets_1.6.4 purrr_1.0.2
## [37] ellipsis_0.3.2 RSpectra_0.16-1
## [39] QFeatures_1.8.0 backports_1.4.1
## [41] prismatic_1.1.1 annotate_1.76.0
## [43] biomaRt_2.52.0 deldir_2.0-4
## [45] vctrs_0.6.5 imputeLCMD_2.1
## [47] ensemble_2.20.2 abind_1.4-5
## [49] cachem_1.0.8 withr_3.0.0
## [51] Gviz_1.40.1 itertools_0.1-3
## [53] BSgenome_1.64.0 checkmate_2.3.1
## [55] GenomicAlignments_1.34.1 fdrtool_1.2.17
## [57] prettyunits_1.2.0 MultiAssayExperiment_1.24.0
## [59] cluster_2.1.6 BiocBaseUtils_1.0.0
## [61] grImport_0.9-7 lazyeval_0.2.2
## [63] crayon_1.5.2 labeling_0.4.3
## [65] glmnet_4.1-8 edgeR_3.40.2
## [67] pkgconfig_2.0.3 nnet_7.3-19
## [69] rlang_1.1.3 lifecycle_1.0.4
## [71] miniUI_0.1.1.1 sandwich_3.1-0
## [73] downloader_0.4 filelock_1.0.3
## [75] affyio_1.68.0 BiocFileCache_2.4.0
## [77] dichromat_2.0-0.1 randomForest_4.7-1.1
## [79] graph_1.74.0 rngtools_1.5.2
## [81] Matrix_1.6-5 carData_3.0-5
## [83] zoo_1.8-12 Rhdf5lib_1.20.0
## [85] base64enc_0.1-3 GlobalOptions_0.1.2
## [87] png_0.1-8 rjson_0.2.21
## [89] bitops_1.0-7 rhdf5filters_1.10.1
## [91] Biostrings_2.66.0 blob_1.2.4
## [93] doRNG_1.8.6 shape_1.4.6.1
## [95] stringr_1.5.1 jpeg_0.1-10
## [97] rstatix_0.7.2 tmvtnorm_1.6

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##	[99]	ggsignif_0.6.4	scales_1.3.0
##	[101]	memoise_2.0.1	magrittr_2.0.3
##	[103]	plyr_1.8.9	zlibbioc_1.44.0
##	[105]	compiler_4.2.1	BiocIO_1.6.0
##	[107]	RColorBrewer_1.1-3	plotrix_3.8-4
##	[109]	pcaMethods_1.90.0	clue_0.3-65
##	[111]	Rsamtools_2.14.0	cli_3.6.2
##	[113]	affy_1.76.0	XVector_0.38.0
##	[115]	formatR_1.14	htmlTable_2.4.2
##	[117]	Formula_1.2-5	MASS_7.3-60.0.1
##	[119]	tidyselect_1.2.1	vsn_3.66.0
##	[121]	stringi_1.8.3	highr_0.10
##	[123]	yaml_2.3.8	norm_1.0-11.1
##	[125]	askpass_1.2.0	locfit_1.5-9.9
##	[127]	latticeExtra_0.6-30	MALDIquant_1.22.2
##	[129]	VariantAnnotation_1.42.1	tools_4.2.1
##	[131]	parallel_4.2.1	circlize_0.4.16
##	[133]	rstudioapi_0.15.0	foreach_1.5.2
##	[135]	foreign_0.8-86	gridExtra_2.3
##	[137]	farver_2.1.1	mzID_1.36.0
##	[139]	Rtsne_0.17	digest_0.6.35
##	[141]	BiocManager_1.30.22	shiny_1.8.0
##	[143]	car_3.1-2	broom_1.0.5
##	[145]	later_1.3.2	ncdf4_1.22
##	[147]	biovizBase_1.44.0	ComplexHeatmap_2.14.0
##	[149]	colorspace_2.1-0	XML_3.99-0.16.1
##	[151]	reticulate_1.35.0	umap_0.2.10.0
##	[153]	splines_4.2.1	rematch2_2.1.2
##	[155]	gmm_1.8	xtable_1.8-4
##	[157]	jsonlite_1.8.8	Hmisc_5.1-2
##	[159]	pillar_1.9.0	htmltools_0.5.7
##	[161]	mime_0.12	glue_1.7.0
##	[163]	fastmap_1.1.1	BiocParallel_1.32.6
##	[165]	class_7.3-22	codetools_0.2-19
##	[167]	mvtnorm_1.2-4	utf8_1.2.4
##	[169]	lattice_0.22-5	tibble_3.2.1
##	[171]	curl_5.2.1	openssl_2.1.1
##	[173]	interp_1.1-6	survival_3.5-8
##	[175]	rmarkdown_2.26	InteractionSet_1.24.0
##	[177]	munsell_0.5.0	e1071_1.7-14
##	[179]	GetoptLong_1.0.5	rhdf5_2.42.1
##	[181]	GenomeInfoDbData_1.2.9	iterators_1.0.14
##	[183]	impute_1.72.3	reshape2_1.4.4
##	[185]	gtable_0.3.4	