

irCLIP-RNP dataset of CAPRIN1 during stress granule formation

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This is the pipeline used to analyze the irCLIP-RNP of CAPRIN1 during stress granule formation from the whole RNP zone ranging from 70 to 350kDa. The experiment was performed in HEK293T. Three time points (0, 30, and 60min) were collected.

1. Prepare the dataset

```
#Needed libraries
library(DEP2)
library(tidyverse)
library(ggplot2)
library(data.table)
library(pheatmap)
library(RColorBrewer)
library(gplots)
library(hrbrthemes)
library(pacman)
library(textshape)
library(ggExtra)
library(viridis)
library(purrr)
library(hexbin)
library(DESeq2)
library(ggpubr)
library(UpSetR)
library(dplyr)
library(Clipper)
library(factoextra)
library(paletteer)
library(corrplot)
library(psych)
library(ggpmisc)
library(gprofiler2)
library(viridis)
library(GGally)
library(igraph)
library(rstatix)
library(limma)
library(HMD)
library(cluster)
library(eulerr)
```

```
#Open proteinGroups.txt results from MaxQuant
data <- read.delim("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/CAPRIN1_timecourse/0_Data/
proteinGroups.txt")

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

```

#Generate unique names and ids
unique_pg <- make_unique(data, name = "Gene.names", ids = "Protein.IDs")
unique_pg <- unique_pg %>% arrange(name)

#Get the columns
ecols <- grep("LFQ.intensity.", colnames(unique_pg))

#Keep isoform with higher LFQ intensity
iso <- grep("\\.\\.\\.d+$", unique_pg$name)
rbp <- gsub("\\.1", "", c(unique_pg$name[iso]))

#Find original row name of the isoform with higher intensity
find_max_value <- function(rbp) {
  filtered_df <- unique_pg[unique_pg$name %like% rbp, grep("LFQ.intensity.", colnames(unique_pg))
]
  filtered_df$rowSums <- rowSums(filtered_df[, grep("LFQ.intensity.", colnames(filtered_df))])
  max_value <- which.max(filtered_df$rowSums)
  rownames <- rownames(filtered_df)[-max_value]
  return(rownames)
}
max_iso <- c(unlist(lapply(rbp, find_max_value)))

#Remove low intensity isoforms
unique_pg <- unique_pg[!(rownames(unique_pg) %in% max_iso),]

```

```

#Load design matrix
design <- read.delim("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/CAPRIN1_timecourse/0_Data/
Design.txt")
design

```

##	label	condition	replicate	stimulation	crosslinking
## 1	LFQ.intensity.BZ30	noUV	1	unstimulated	noUV
## 2	LFQ.intensity.BZ31	T0	1	unstimulated	UVC
## 3	LFQ.intensity.BZ32	T0	2	unstimulated	UVC
## 4	LFQ.intensity.BZ33	T30	1	stimulated	UVC
## 5	LFQ.intensity.BZ34	T30	2	stimulated	UVC
## 6	LFQ.intensity.BZ35	T60	1	stimulated	UVC
## 7	LFQ.intensity.BZ36	T60	2	stimulated	UVC

2. Determine the UVC-enriched proteins

We used ClipperR R package (PMID: 34635147) to determined the UVC-enriched proteins.

```

#Create a SummarizedExperiment
ecols <- grep("LFQ.intensity.", colnames(unique_pg))
se <- make_se(unique_pg, columns = ecols, expdesign = design)
se_UVC <- se[, se$crosslinking != "noUV" ]
se_UVC <- filter_se(se_UVC, thr = 0, filter_formula = ~ Reverse != '+' & Potential.contaminant !=
"+" & Peptides > 1 & Unique.peptides > 0)
se <- se[rownames(se) %in% rownames(se_UVC),]
write.table(se@assays@data@listData, file = "~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/
CAPRIN1_timecourse/2_DEP/CAPRIN_LFQ_intensity_raw.txt", quote = F, row.names = T, sep = "\t")

set.seed(3)
clipper <- function(flt, time) {
  se_time <- flt[, flt$condition == time]
  se <- se[rownames(se_time), se$condition %in% c("noUV", time) ]
}

```

```

imputed <- DEP2::impute(se, fun = "QRILC")
data <- as.data.frame(assay(imputed))
clipper = Clipper(score.exp = as.matrix(data[,c(2,3)]), score.back = as.matrix(data[, -c(2,3)]),
  FDR = 0.05, analysis = "e")
data$FDR <- clipper$q
data <- cbind(data, rowMeans(data[,c(2,3)])-data[1])
data$TP <- time
data$name <- rownames(data)
rownames(data) <- NULL
colnames(data)[5] <- c("logFC")
deg <- subset(data, FDR < 0.1 & logFC > log2(3))
return(list(data = data, deg = deg))
}

T0_clipper <- clipper(se, "T0")
T30_clipper <- clipper(se, "T30")
T60_clipper <- clipper(se, "T60")

write.table(T0_clipper$data, file="~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/CAPRIN1_
timecourse/2_DEP/CAPRIN_clipper_results_T0.txt", quote = F, row.names = F, sep = "\t")
write.table(T30_clipper$data, file="~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/CAPRIN1_
timecourse/2_DEP/CAPRIN_clipper_results_T30.txt", quote = F, row.names = F, sep = "\t")
write.table(T60_clipper$data, file="~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/CAPRIN1_
timecourse/2_DEP/CAPRIN_clipper_results_T60.txt", quote = F, row.names = F, sep = "\t")

CAPRIN_UVC <- rbind(T0_clipper$deg[,4:7], T30_clipper$deg[,4:7], T60_clipper$deg[,4:7])
CAPRIN_UVC_genes <- unique(CAPRIN_UVC$name)
CAPRIN_UVC

```

##	FDR	logFC	TP	name
## 3	0.03773585	2.169631	T0	ACIN1
## 8	0.03773585	2.772824	T0	ATXN2
## 17	0.03773585	2.230398	T0	CAD
## 22	0.03773585	4.694068	T0	CDCA2
## 31	0.03773585	3.085259	T0	CSDE1
## 34	0.03773585	3.280484	T0	DCD
## 44	0.03773585	2.046629	T0	DDX3X
## 51	0.03773585	2.026989	T0	DDX56
## 59	0.03773585	3.251407	T0	DNMT1
## 62	0.03773585	5.121508	T0	EEF1A1
## 67	0.03773585	1.988573	T0	ELAVL1
## 72	0.04878049	1.892060	T0	FMR1
## 75	0.03773585	3.621223	T0	FUS
## 76	0.03773585	3.748800	T0	FXR1
## 77	0.03773585	3.738969	T0	FXR2
## 78	0.03773585	4.764645	T0	G3BP1
## 86	0.03773585	2.916895	T0	HIST1H1C
## 87	0.04878049	1.745264	T0	HIST3H2A
## 88	0.03773585	2.370417	T0	HNRNPA1
## 89	0.03773585	3.353308	T0	HNRNPA2B1
## 90	0.03773585	2.945718	T0	HNRNPC
## 91	0.03773585	5.545154	T0	HNRNPF
## 92	0.03773585	4.927953	T0	HNRNPH1
## 93	0.03773585	2.312214	T0	HNRNPK
## 94	0.03773585	2.141483	T0	HNRNPL
## 95	0.03773585	2.241991	T0	HNRNPM
## 96	0.03773585	2.681650	T0	HNRNPR
## 99	0.03773585	2.821590	T0	HSP90AB1
## 100	0.03773585	2.132053	T0	HSPA1A
## 101	0.04878049	1.886912	T0	IGF2BP1
## 103	0.03773585	5.338276	T0	IGHG1
## 107	0.03773585	2.556717	T0	KHSRP
## 113	0.03773585	5.715124	T0	LRPPRC
## 119	0.03773585	2.597072	T0	MDC1
## 124	0.03773585	5.358020	T0	NCL
## 127	0.04878049	1.660961	T0	NFXL1

##	133	0.03773585	2.633291	T0	NOP14
##	137	0.03773585	4.857316	T0	NPM1
##	141	0.03773585	2.894352	T0	PABPC1
##	149	0.04878049	1.838487	T0	PHF6
##	162	0.03773585	2.861047	T0	RPS11
##	164	0.03773585	5.738320	T0	RPS2
##	166	0.04878049	1.857855	T0	RPS24
##	167	0.03773585	2.968374	T0	RPS26
##	168	0.03773585	4.191498	T0	RPS3
##	169	0.03773585	2.524558	T0	RPS3A
##	170	0.03773585	2.505033	T0	RPS4X
##	172	0.03773585	3.111346	T0	RPS7
##	176	0.03773585	3.758833	T0	RRP12
##	180	0.03773585	4.052574	T0	SERBP1
##	181	0.03773585	4.498833	T0	SFPQ
##	183	0.03773585	3.023340	T0	SMARCA5
##	184	0.03773585	3.854731	T0	SND1
##	185	0.03773585	2.065353	T0	SNRNP200
##	189	0.03773585	2.021333	T0	SRPK1
##	193	0.03773585	3.560074	T0	SYNCRIP
##	199	0.04878049	1.675527	T0	TOP1
##	210	0.03773585	4.082811	T0	UBAP2L
##	212	0.04878049	1.773150	T0	UPF3B
##	219	0.03773585	2.971594	T0	YBX1
##	221	0.04878049	1.628244	T0	ZC3H11A
##	228	0.04878049	1.819437	T0	ZNF280C
##	230	0.03773585	2.650931	T0	ZNF48
##	10	0.07692308	1.605606	T30	BAZ1B
##	11	0.02857143	3.708702	T30	BAZ2A
##	28	0.05555556	2.000559	T30	CKAP2
##	311	0.02857143	6.133811	T30	CSDE1
##	441	0.05555556	1.912250	T30	DDX3X
##	54	0.05555556	2.092090	T30	DHX30
##	591	0.02857143	3.206238	T30	DNMT1
##	671	0.02857143	4.086809	T30	ELAVL1
##	71	0.05555556	1.914577	T30	FBL
##	721	0.02857143	2.653106	T30	FMR1
##	74	0.02857143	4.560017	T30	FUBP3
##	751	0.02857143	3.113553	T30	FUS
##	761	0.05555556	2.123598	T30	FXR1
##	771	0.06896552	1.820970	T30	FXR2
##	781	0.02857143	3.232470	T30	G3BP1
##	861	0.02857143	3.189342	T30	HIST1H1C
##	881	0.06896552	1.624580	T30	HNRNPA1
##	891	0.02857143	5.895342	T30	HNRNPA2B1
##	901	0.02857143	2.823643	T30	HNRNPC
##	921	0.02857143	6.059913	T30	HNRNPH1
##	931	0.05555556	2.005732	T30	HNRNPK
##	941	0.02857143	3.038330	T30	HNRNPL
##	951	0.02857143	3.269821	T30	HNRNPM
##	991	0.05555556	2.076075	T30	HSP90AB1
##	1001	0.02857143	2.696694	T30	HSPA1A
##	1011	0.05555556	2.244918	T30	IGF2BP1
##	1031	0.02857143	4.261668	T30	IGHG1
##	1131	0.02857143	4.852124	T30	LRPPRC
##	118	0.02857143	5.313008	T30	MARS
##	1191	0.05263158	2.510758	T30	MDCl
##	1241	0.02857143	5.645755	T30	NCL
##	1271	0.02857143	3.053412	T30	NFXL1
##	1331	0.05555556	2.019744	T30	NOP14
##	1371	0.02857143	4.236133	T30	NPM1
##	1411	0.05555556	2.282763	T30	PABPC1
##	147	0.06896552	1.609178	T30	PHF2
##	1491	0.05263158	2.539820	T30	PHF6
##	153	0.02857143	2.748150	T30	PTBP1
##	1621	0.05555556	2.168289	T30	RPS11
##	1641	0.02857143	6.776957	T30	RPS2
##	1671	0.05263158	2.522224	T30	RPS26

##	1691	0.02857143	3.567995	T30	RPS3A
##	1701	0.06896552	1.702162	T30	RPS4X
##	174	0.05555556	2.055656	T30	RPS9
##	1761	0.05555556	2.325200	T30	RRP12
##	1801	0.02857143	5.191610	T30	SERBP1
##	1811	0.02857143	3.936344	T30	SFPQ
##	1831	0.02857143	2.813502	T30	SMARCA5
##	1891	0.02857143	4.760515	T30	SRPK1
##	192	0.02857143	2.633799	T30	STT3B
##	1931	0.02857143	3.459988	T30	SYNCRIP
##	195	0.02857143	4.952770	T30	TDRD3
##	198	0.02857143	2.751729	T30	TIAL1
##	1991	0.05555556	1.896960	T30	TOP1
##	202	0.05555556	2.357456	T30	TOP3B
##	207	0.02857143	4.080510	T30	U2AF2
##	2121	0.05555556	1.882293	T30	UPF3B
##	2191	0.02857143	4.543290	T30	YBX1
##	2301	0.02857143	2.750881	T30	ZNF48
##	4	0.05263158	2.935222	T60	ACTA1
##	111	0.05263158	2.867001	T60	BAZ2A
##	312	0.05263158	5.602257	T60	CSDE1
##	341	0.05263158	3.507318	T60	DCD
##	442	0.09803922	1.743731	T60	DDX3X
##	592	0.05263158	2.648051	T60	DNMT1
##	621	0.06976744	2.091934	T60	EEF1A1
##	672	0.05263158	4.085978	T60	ELAVL1
##	711	0.05263158	2.360801	T60	FBL
##	722	0.05263158	2.346995	T60	FMR1
##	741	0.05263158	3.158318	T60	FUBP3
##	752	0.05263158	3.020144	T60	FUS
##	772	0.09803922	1.895343	T60	FXR2
##	782	0.05263158	3.038115	T60	G3BP1
##	862	0.05263158	3.030668	T60	HIST1H1C
##	892	0.05263158	6.107055	T60	HNRNPA2B1
##	902	0.05263158	3.060640	T60	HNRNPC
##	922	0.05263158	6.001094	T60	HNRNPH1
##	932	0.06976744	2.204697	T60	HNRNPK
##	942	0.05263158	3.597231	T60	HNRNPL
##	952	0.05263158	3.512366	T60	HNRNPM
##	992	0.05263158	2.280296	T60	HSP90AB1
##	1002	0.09803922	1.891784	T60	HSPA1A
##	1012	0.06976744	2.165613	T60	IGF2BP1
##	1181	0.05263158	3.937063	T60	MARS
##	1192	0.05263158	2.604335	T60	MDC1
##	1242	0.05263158	5.708128	T60	NCL
##	1372	0.05263158	4.347834	T60	NPM1
##	1412	0.09803922	1.965583	T60	PABPC1
##	1492	0.05263158	2.413857	T60	PHF6
##	1531	0.05263158	2.583643	T60	PTBP1
##	1622	0.08888889	1.981721	T60	RPS11
##	1642	0.05263158	6.576949	T60	RPS2
##	1692	0.05263158	3.638107	T60	RPS3A
##	1702	0.09803922	1.856029	T60	RPS4X
##	1741	0.05263158	2.339771	T60	RPS9
##	1762	0.08888889	2.001339	T60	RRP12
##	1802	0.05263158	5.167012	T60	SERBP1
##	1812	0.05263158	3.670013	T60	SFPQ
##	1832	0.05263158	2.844567	T60	SMARCA5
##	1921	0.05263158	2.405711	T60	STT3B
##	1932	0.05263158	2.805514	T60	SYNCRIP
##	1951	0.05263158	3.674443	T60	TDRD3
##	1992	0.06976744	2.084685	T60	TOP1
##	201	0.05263158	3.107122	T60	TOP2B
##	2021	0.05263158	2.735602	T60	TOP3B
##	2071	0.05263158	4.124639	T60	U2AF2
##	2122	0.09803922	1.750227	T60	UPF3B
##	2192	0.05263158	4.426913	T60	YBX1
##	2281	0.06976744	2.137822	T60	ZNF280C

3. Differential enrichment analysis

We performed differential enrichment analysis using Limma and generate the scatterplot.

```
#Prepare the SummarizedExperiment
norm <- normalize_vsn(se_UVC)
imputed <- DEP2::impute(norm, fun = "QRILC")

model_vsn <- model.matrix(~ condition, colData(imputed))

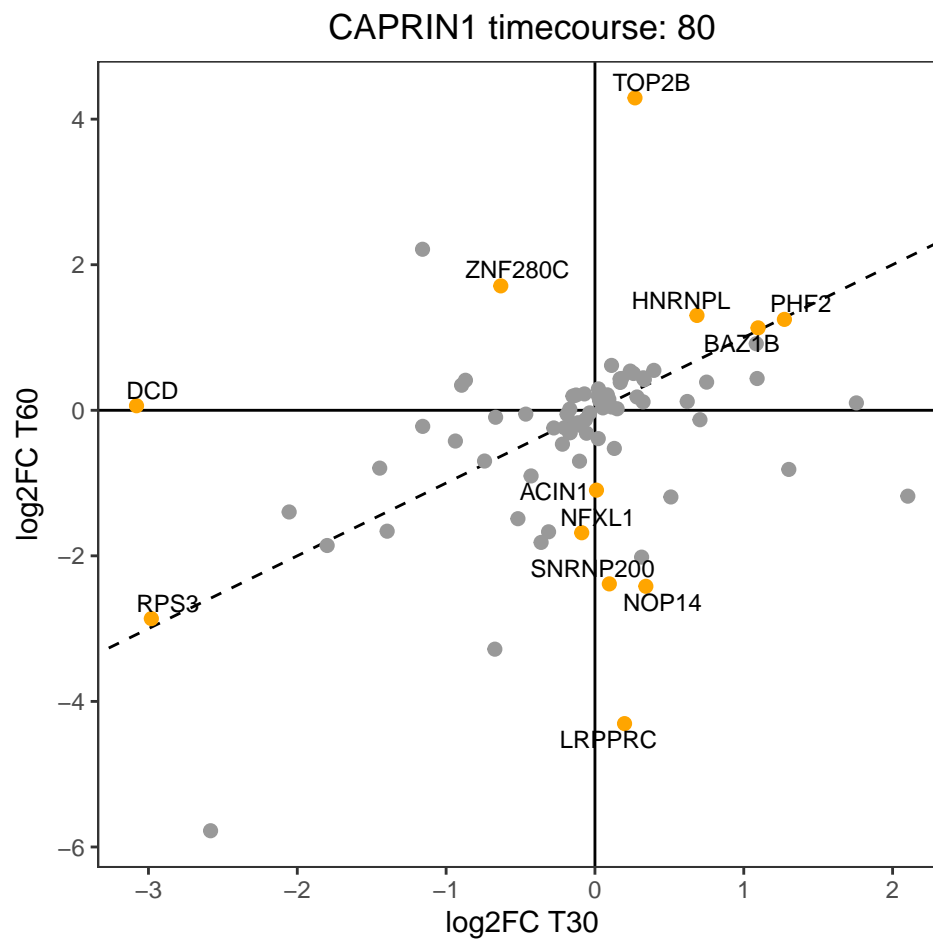
#Interaction analysis
CAPRIN1_fit1_norm_int_vsn = lmFit(assay(imputed), design = model_vsn)
CAPRIN1_fit2_norm_int_vsn <- eBayes(CAPRIN1_fit1_norm_int_vsn)
CAPRIN1_int_norm_vsn_both <- topTable(CAPRIN1_fit2_norm_int_vsn, coef = c("conditionT30",
  "conditionT60"), number = length(rownames(imputed)))
write.table(CAPRIN1_int_norm_vsn_both, file = "~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/
  CAPRIN1_timecourse/2_DEP/CAPRIN_DEP_results_onefactor.txt", quote = F, row.names = T, sep = "
  \t")
CAPRIN1_int_norm_vsn_T30 <- topTable(CAPRIN1_fit2_norm_int_vsn, coef = c("conditionT30"), number
  = length(rownames(imputed)))
CAPRIN1_int_norm_vsn_T60 <- topTable(CAPRIN1_fit2_norm_int_vsn, coef = c("conditionT60"), number
  = length(rownames(imputed)))
CAPRIN1_int_norm_vsn_both$name <- rownames(CAPRIN1_int_norm_vsn_both)
CAPRIN1_int_norm_vsn_T30$name <- rownames(CAPRIN1_int_norm_vsn_T30)
CAPRIN1_int_norm_vsn_T60$name <- rownames(CAPRIN1_int_norm_vsn_T60)
CAPRIN1_sign_prot <- subset(CAPRIN1_int_norm_vsn_both, adj.P.Val < 0.1)
CAPRIN1_sign_prot <- subset(CAPRIN1_sign_prot, abs(conditionT30) > 1 | abs(conditionT60) > 1)

#Subset with the UVC proteins
CAPRIN1_int_norm_vsn_both <- subset(CAPRIN1_int_norm_vsn_both, rownames(CAPRIN1_int_norm_vsn_both
  ) %in% CAPRIN_UVC_genes)

#Scatterplot
CAPRIN1_int_norm_vsn_both$int_sign <- CAPRIN1_int_norm_vsn_both$name %in% rownames(CAPRIN1_sign_
  prot)
CAPRIN1_colors <- c("FALSE" = "#999999", "TRUE" = "orange")
```

Volcano plot

```
ggplot <- ggplot(data=CAPRIN1_int_norm_vsn_both, aes(x=conditionT30, y=conditionT60)) + geom_
  vline(xintercept = 0) + geom_hline(yintercept = 0) + geom_abline(intercept = 0, linetype=2)
  +
  geom_point(shape=19, size=2, aes(col = int_sign)) +
  labs(title = paste("CAPRIN1 timecourse:", nrow(CAPRIN1_int_norm_vsn_both)), x = expression("
    log2FC T30"), y = expression("log2FC T60")) +
  scale_color_manual(values = CAPRIN1_colors) +
  ggrepel::geom_text_repel(data = CAPRIN1_int_norm_vsn_both[CAPRIN1_int_norm_vsn_both$int_sign ==
    "TRUE"], aes(label = name), size = 3, box.padding = unit(0.1, "lines"), point.padding =
    unit(0.1, "lines"), segment.size = 0.5, max.overlaps = Inf) +
  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))# +
```



```
#Save the plot as pdf
pdf("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/CAPRIN1_timecourse/2_DEP/CAPRIN1_DEP_plot.
pdf", height = 5, width = 5)
ggplot
dev.off()
```

Heatmap of significant UVC-enriched proteins

```
#Heatmap
CAPRIN1_sign_T30 <- subset(CAPRIN1_int_norm_vsn_T30, adj.P.Val < 0.1 & abs(logFC) > 1)
CAPRIN1_sign_T60 <- subset(CAPRIN1_int_norm_vsn_T60, adj.P.Val < 0.1 & abs(logFC) > 1)

CAPRIN1_sign_T30 <- subset(CAPRIN1_sign_T30, rownames(CAPRIN1_sign_T30) %in% CAPRIN_UVC_genes)
CAPRIN1_sign_T60 <- subset(CAPRIN1_sign_T60, rownames(CAPRIN1_sign_T60) %in% CAPRIN_UVC_genes)

lt.tsk = list(T30 = rownames(CAPRIN1_sign_T30),
              T60 = rownames(CAPRIN1_sign_T60))

fromList <- function (input) {
```

```

# Same as original fromList()...
elements <- unique(unlist(input))
data <- unlist(lapply(input, function(x) {
  x <- as.vector(match(elements, x))
}))
data[is.na(data)] <- as.integer(0)
data[data != 0] <- as.integer(1)
data <- data.frame(matrix(data, ncol = length(input), byrow = F))
data <- data[which(rowSums(data) != 0), ]
names(data) <- names(input)
# ... Except now it conserves your original value names!
row.names(data) <- elements
return(data)
}

#Get significant genes
sign.proteins <- fromList(lt.tsk)

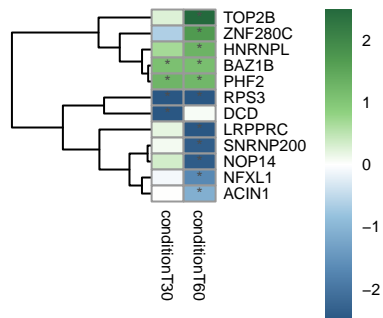
#Matrix
CAPRIN1_sign_prot_HM <- CAPRIN1_sign_prot[,c(1,2)]
CAPRIN1_sign_prot_HM <- subset(CAPRIN1_sign_prot_HM, rownames(CAPRIN1_sign_prot_HM) %in% CAPRIN_
  UVC_genes)

#Annotation and color
my.breaks <- c(seq(-2.5, 2.5, by=0.01))
my.colors <- c(rev(paletteer_c("ggthemes::Green-Blue-White Diverging", length(my.breaks))))

labels <- sign.proteins
colnames(labels) <- colnames(CAPRIN1_sign_prot_HM)
labels[labels == 1] <- "*"
labels[labels == 0] <- ""
labels <- labels[match(rownames(CAPRIN1_sign_prot_HM), rownames(labels)),]

pheatmap(
  mat = CAPRIN1_sign_prot_HM,
  cellwidth = 12,
  cellheight = 6,
  display_numbers = labels,
  fontsize_number=5.5,
  color = my.colors,
  breaks = my.breaks,
  clustering_distance_cols = "euclidean",
  show_colnames = TRUE,
  show_rownames = TRUE,
  drop_levels = TRUE,
  fontsize = 5.5,
  cluster_rows = TRUE,
  cluster_cols = FALSE
)

```

```
#Save the plot as pdf
pheatmap(
  mat              = CAPRIN1_sign_prot_HM,
  cellwidth = 12,
  cellheight = 6,
  display_numbers = labels,
  fontsize_number=5.5,
  color = my.colors,
  breaks = my.breaks,
  clustering_distance_cols = "euclidean",
  show_colnames      = TRUE,
  show_rownames      = TRUE,
  drop_levels        = TRUE,
  fontsize            = 5.5,
  cluster_rows       = TRUE,
  cluster_cols        = FALSE,
  filename = "~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/CAPRIN1_timecourse/2_DEP/CAPRIN1_
    sign_HM.pdf",
  width = 5,
  height = 5
)
```

Overlap with existing databases.

```
#Load files
```

```

res2 <- read.delim("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/CAPRIN1_timecourse/3_SG_
Database/1_Venn_Diagram/MSGP_database_proteins.txt", header = TRUE, dec = ".")
res3 <- read.delim("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/CAPRIN1_timecourse/3_SG_
Database/1_Venn_Diagram/RNP_granule_database.txt", header = TRUE, dec = ".")

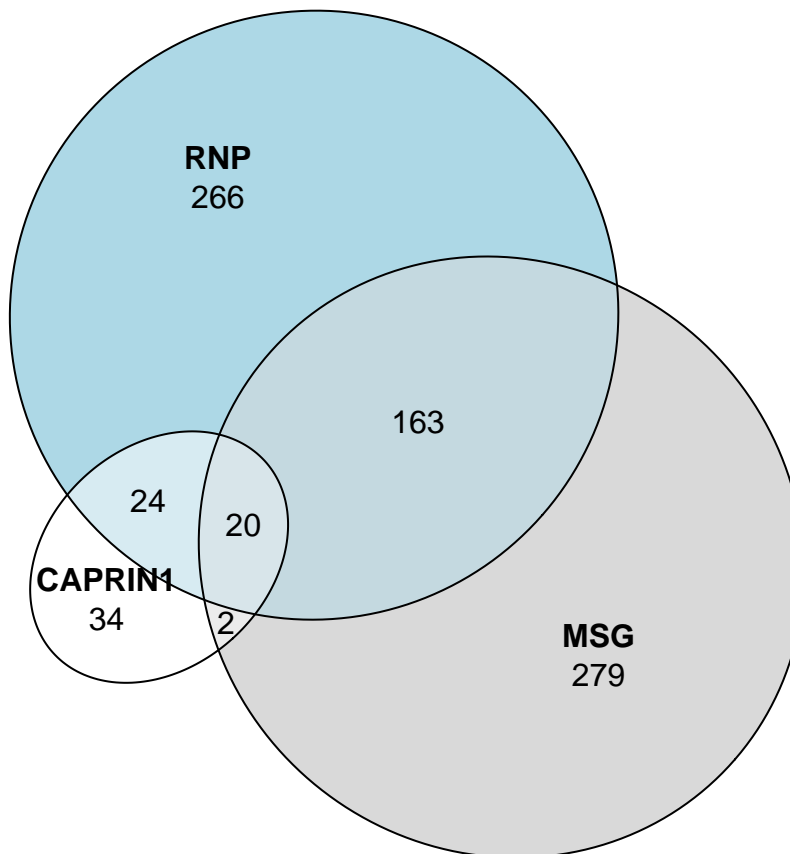
s4 <- list(CAPRIN1 = CAPRIN_UVC_genes,
          MSG = res2$Gene,
          RNP = res3$Gene)

fromList <- function (input) {
  elements <- unique(unlist(input))
  data <- unlist(lapply(input, function(x) {
    x <- as.vector(match(elements, x))
  }))
  data[is.na(data)] <- as.integer(0)
  data[data != 0] <- as.integer(1)
  data <- data.frame(matrix(data, ncol = length(input), byrow = F))
  data <- data[which(rowSums(data) != 0), ]
  names(data) <- names(input)
  row.names(data) <- elements
  return(data)
}

sign.proteins <- fromList(s4)
write.table(sign.proteins, file = "~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/CAPRIN1_
timecourse/3_SG_Database/1_Venn_Diagram/CAPRIN1_upset_database.txt", row.names = TRUE, sep =
"\t")

plot(euler(s4, shape = "ellipse"), quantities = TRUE)

```



```
pdf("~/Documents/Postdoc/PD_Projects/3_irCLIP-RNP/MS/CAPRIN1_timecourse/3_SG_Database/1_Venn_
Diagram/CAPRIN1_MSGP_venn.pdf", height = 7, width = 7)
plot(euler(s4, shape = "ellipse"), quantities = TRUE)
dev.off()
```

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

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] stats4      stats      graphics  grDevices  utils      datasets  methods
## [8] base
##
## other attached packages:
## [1] eulerr_7.0.1           cluster_2.1.6
## [3] HDMD_1.2              MASS_7.3-60.0.1
## [5] rstatix_0.7.2         igraph_2.0.3
## [7] GGally_2.2.1          gprofiler2_0.2.3
## [9] ggpmisc_0.5.5         ggpp_0.5.6
## [11] psych_2.4.3           corrplot_0.92
## [13] paletteer_1.6.0       factoextra_1.0.7
## [15] Clipper_0.0.0.9000     UpSetR_1.4.0
## [17] ggpubr_0.6.0          DESeq2_1.38.3
## [19] hexbin_1.28.3         viridis_0.6.5
## [21] viridisLite_0.4.2     ggExtra_0.10.1
## [23] textshape_1.7.3       pacman_0.5.1
## [25] hrbrthemes_0.8.7      gplots_3.1.3.1
## [27] RColorBrewer_1.1-3    pheatmap_1.0.12
## [29] data.table_1.15.2     lubridate_1.9.3
## [31] forcats_1.0.0         stringr_1.5.1
## [33] dplyr_1.1.4           purrr_1.0.2
## [35] readr_2.1.5           tidyr_1.3.1
## [37] tibble_3.2.1          ggplot2_3.5.0
## [39] tidyverse_2.0.0       DEP2_0.4.8.24
## [41] R6_2.5.1              limma_3.54.2
## [43] MSnbase_2.24.2        ProtGenerics_1.30.0
## [45] mzR_2.32.0            Rcpp_1.0.12
## [47] MsCoreUtils_1.10.0    SummarizedExperiment_1.28.0
## [49] Biobase_2.58.0        GenomicRanges_1.50.2
## [51] GenomeInfoDb_1.34.9   IRanges_2.32.0
## [53] S4Vectors_0.36.2      BiocGenerics_0.44.0
## [55] MatrixGenerics_1.10.0 matrixStats_1.2.0
##
## loaded via a namespace (and not attached):
## [1] SparseM_1.81          ggthemes_5.1.0
## [3] missForest_1.5        bit64_4.0.5
## [5] knitr_1.45            DelayedArray_0.24.0
```

##	[7]	KEGGREST_1.38.0	RCurl_1.98-1.14
##	[9]	AnnotationFilter_1.22.0	doParallel_1.0.17
##	[11]	generics_0.1.3	preprocessCore_1.60.2
##	[13]	RSQLite_2.3.5	proxy_0.4-27
##	[15]	bit_4.0.5	tzdb_0.4.0
##	[17]	httpuv_1.6.14	assertthat_0.2.1
##	[19]	TCseq_1.22.6	xfun_0.42
##	[21]	hms_1.1.3	evaluate_0.23
##	[23]	promises_1.2.1	fansi_1.0.6
##	[25]	caTools_1.18.2	htmlwidgets_1.6.4
##	[27]	DBI_1.2.2	geneplotter_1.76.0
##	[29]	ellipsis_0.3.2	RSpectra_0.16-1
##	[31]	QFeatures_1.8.0	backports_1.4.1
##	[33]	fontLiberation_0.1.0	prismatic_1.1.1
##	[35]	annotate_1.76.0	fontBitstreamVera_0.1.1
##	[37]	vctrs_0.6.5	imputeLCMD_2.1
##	[39]	quantreg_5.97	abind_1.4-5
##	[41]	cachem_1.0.8	withr_3.0.0
##	[43]	itertools_0.1-3	GenomicAlignments_1.34.1
##	[45]	fdrtool_1.2.17	MultiAssayExperiment_1.24.0
##	[47]	mnormt_2.1.1	lazyeval_0.2.2
##	[49]	crayon_1.5.2	crul_1.4.0
##	[51]	labeling_0.4.3	glmnet_4.1-8
##	[53]	edgeR_3.40.2	pkgconfig_2.0.3
##	[55]	nlme_3.1-164	rlang_1.1.3
##	[57]	lifecycle_1.0.4	miniUI_0.1.1.1
##	[59]	sandwich_3.1-0	MatrixModels_0.5-3
##	[61]	downloader_0.4	fontquiver_0.2.1
##	[63]	httpcode_0.3.0	affyio_1.68.0
##	[65]	extrafontdb_1.0	polyclip_1.10-6
##	[67]	randomForest_4.7-1.1	rngtools_1.5.2
##	[69]	Matrix_1.6-5	carData_3.0-5
##	[71]	zoo_1.8-12	GlobalOptions_0.1.2
##	[73]	png_0.1-8	rjson_0.2.21
##	[75]	bitops_1.0-7	KernSmooth_2.23-22
##	[77]	Biostrings_2.66.0	blob_1.2.4
##	[79]	doRNG_1.8.6	shape_1.4.6.1
##	[81]	tmvtnorm_1.6	ggsignif_0.6.4
##	[83]	scales_1.3.0	memoise_2.0.1
##	[85]	magrittr_2.0.3	plyr_1.8.9
##	[87]	zlibbioc_1.44.0	compiler_4.2.1
##	[89]	pcaMethods_1.90.0	clue_0.3-65
##	[91]	Rsamtools_2.14.0	cli_3.6.2
##	[93]	affy_1.76.0	XVector_0.38.0
##	[95]	tidyselect_1.2.1	vsu_3.66.0
##	[97]	stringi_1.8.3	highr_0.10
##	[99]	yaml_2.3.8	norm_1.0-11.1
##	[101]	askpass_1.2.0	locfit_1.5-9.9
##	[103]	MALDIquant_1.22.2	ggrepel_0.9.5
##	[105]	grid_4.2.1	ggstats_0.5.1
##	[107]	polynom_1.4-1	tools_4.2.1
##	[109]	timechange_0.3.0	parallel_4.2.1
##	[111]	circlize_0.4.16	rstudioapi_0.15.0
##	[113]	foreach_1.5.2	gridExtra_2.3
##	[115]	farver_2.1.1	mzID_1.36.0
##	[117]	Rtsne_0.17	digest_0.6.35
##	[119]	BiocManager_1.30.22	shiny_1.8.0
##	[121]	gfonts_0.2.0	car_3.1-2
##	[123]	broom_1.0.5	later_1.3.2
##	[125]	ncdf4_1.22	httr_1.4.7
##	[127]	gdtools_0.3.5	AnnotationDbi_1.60.2
##	[129]	ComplexHeatmap_2.14.0	colorspace_2.1-0
##	[131]	polylabelr_0.2.0	XML_3.99-0.16.1
##	[133]	reticulate_1.35.0	umap_0.2.10.0
##	[135]	splines_4.2.1	rematch2_2.1.2
##	[137]	gmm_1.8	plotly_4.10.4
##	[139]	systemfonts_1.0.5	xtable_1.8-4
##	[141]	jsonlite_1.8.8	pillar_1.9.0

##	[143]	htmltools_0.5.7	mime_0.12
##	[145]	glue_1.7.0	fastmap_1.1.1
##	[147]	BiocParallel_1.32.6	class_7.3-22
##	[149]	codetools_0.2-19	mvtnorm_1.2-4
##	[151]	utf8_1.2.4	lattice_0.22-5
##	[153]	curl_5.2.1	gtools_3.9.5
##	[155]	openssl_2.1.1	Rttf2pt1_1.3.12
##	[157]	survival_3.5-8	rmarkdown_2.26
##	[159]	munsell_0.5.0	e1071_1.7-14
##	[161]	GetoptLong_1.0.5	GenomeInfoDbData_1.2.9
##	[163]	iterators_1.0.14	impute_1.72.3
##	[165]	reshape2_1.4.4	gtable_0.3.4
##	[167]	extrafont_0.19	