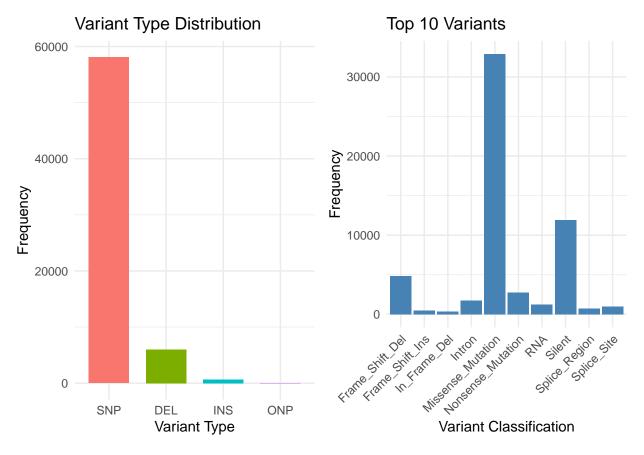
```
raw.clinical.patients <- read.table("data_clinical_patient.txt", sep = "\t",
                                      header = TRUE)
raw.data.mutations <- read.table("data_mutations.txt", sep = "\t",</pre>
                                   header = TRUE)
raw.data.RNAseq <- read.csv("RNAseq_BRCA.csv", row.names=1)</pre>
#Filter data where you only have 0 or 1 read count across all samples.
raw.data.RNAseq <- raw.data.RNAseq[rowSums(raw.data.RNAseq)>1,]
colnames(raw.data.RNAseq) <- make.unique(sapply(colnames(raw.data.RNAseq), function(name) {</pre>
  segments <- strsplit(name, "\\.")[[1]]</pre>
  paste(segments[1:3], collapse = "-")
}))
#Unique Patients in each data set
unique.clinical <- as.data.frame(unique(raw.clinical.patients PATIENT_ID))
unique.mutations <- as.data.frame(unique</pre>
                                    (raw.data.mutations$Tumor_Sample_Barcode))
unique.RNA <- as.data.frame(colnames(raw.data.RNAseq[,1:length(raw.data.RNAseq)]))
#Addition patient ID's to Mutation data
mutation.patients <- as.data.frame(raw.data.mutations$Tumor_Sample_Barcode)</pre>
colnames(mutation.patients) <- "Patient_ID"</pre>
mutation.patients$Patient_ID <- substr(mutation.patients$Patient_ID, 1, 12)</pre>
raw.data.mutations <- cbind(mutation.patients, raw.data.mutations)</pre>
colnames(unique.clinical) <- "Patient_ID"</pre>
colnames(unique.mutations) <- "Patient_ID"</pre>
colnames(unique.RNA) <- "Patient_ID"</pre>
unique.mutations$Patient_ID <- substr(unique.mutations$Patient_ID, 1, 12)
#Finding common patients
common patient ids <- Reduce(intersect, list(</pre>
  unique.clinical$Patient_ID,
 unique.mutations $Patient_ID,
  unique.RNA$Patient_ID
#3 data sets with all 975 common patients
clinical.data <- raw.clinical.patients[raw.clinical.patients$PATIENT_ID</pre>
                                          %in% common_patient_ids, ]
mutation.data <- raw.data.mutations[raw.data.mutations$Patient_ID</pre>
                                      %in% common_patient_ids, ]
seq.data <- raw.data.RNAseq[,names(raw.data.RNAseq)</pre>
                             %in% clinical.data$PATIENT_ID]
library(ggplot2)
library(gridExtra)
data_counts_mutation <- data.frame(table(mutation.data$Hugo_Symbol))</pre>
colnames(data counts mutation) <- c("Gene", "Count")</pre>
data_counts_mutation$Percentage <- round((data_counts_mutation$Count /</pre>
                                               sum(data counts mutation$Count)) * 100, 4)
```

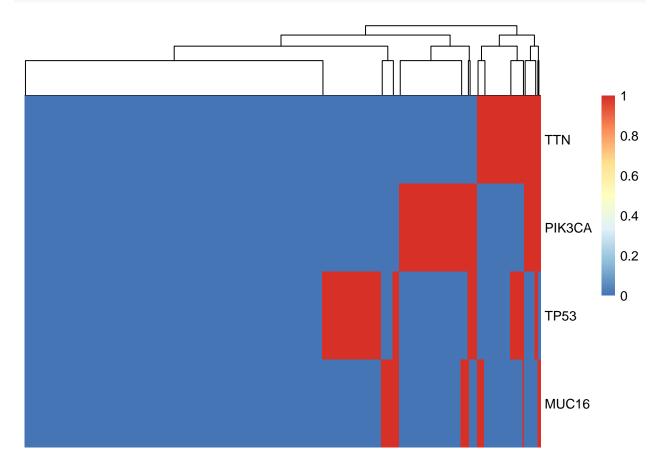
```
top_50_mutations <- data_counts_mutation[order(data_counts_mutation$Count,
                                        decreasing = TRUE), ][1:30, ]
# Reorder the gene factor for the top 50
top_50_mutations$Gene <- factor(top_50_mutations$Gene,</pre>
                                     levels = top 50 mutations$Gene
                                     [order(top 50 mutations$Count,
                                            decreasing = TRUE)])
top_Mutations_plt <- ggplot(top_50_mutations, aes(x = Gene, y=Count)) +</pre>
  geom_bar(stat="identity", fill = "steelblue") +
  labs(title = "Top 30 Gene Variants", x = "Gene", y = "Count") +
  theme_minimal()+
  theme(plot.title = element_text(size = 20)) +
  theme(axis.text.x = element_text(angle = 45, hjust = 1),)
ggsave("top_30_mutation.png", top_Mutations_plt, width = 8, height =6)
# Create data frame for Variant Type
var.type <- as.data.frame(table(mutation.data$Variant_Type))</pre>
var.type <- var.type[order(var.type$Freq, decreasing = TRUE), ]</pre>
# Convert Var1 to a factor with levels in the desired order
var.type$Var1 <- factor(var.type$Var1, levels = var.type$Var1)</pre>
# Plot 1: Variant Type Distribution
plt1 <- ggplot(data = var.type, aes(x = Var1, y = Freq)) +</pre>
  geom_col(aes(fill = Var1), width = 0.7) +
  labs(title = "Variant Type Distribution", x = "Variant Type", y = "Frequency") +
  theme_minimal()+
  theme(legend.position = "none")
# Create data frame for Variant Classification
var.class <- as.data.frame(table(mutation.data$Variant_Classification))</pre>
# Plot 2: Top 10 Variants
plt2 <- ggplot(data = var.class[var.class$Freq > 100, ], aes(x = Var1, y = Freq)) +
  geom_col(fill = "steelblue") +
  labs(title = "Top 10 Variants", x = "Variant Classification", y = "Frequency") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
plt_var_class <- grid.arrange(plt1, plt2, nrow = 1)</pre>
```



```
# Save Combined Plot
ggsave("plt_var_class.png", plot = plt_var_class, width = 6, height = 4)
```

```
library(pheatmap)
cnv_events = unique(mutation.data$Variant_Classification)
oncomat = reshape2::dcast(
  data = mutation.data,
  formula = Hugo_Symbol ~ Tumor_Sample_Barcode,
  fun.aggregate = function(x, cnv = cnv_events) {
    x = as.character(x) # >= 2 same/distinct variant classification = Multi_Hit
    xad = x[x \%in\% cnv]
    xvc = x[!x \%in\% cnv]
    if (length(xvc) > 0) {
      xvc = ifelse(test = length(xvc) > 1,
                   yes = 'Multi_Hit',
                   no = xvc)
    }
    x = ifelse(
      test = length(xad) > 0,
      yes = paste(xad, xvc, sep = ';'),
      no = xvc
    x = gsub(pattern = ';$',
             replacement = '',
```

```
x = x
    x = gsub(pattern = '^;',
              replacement = '',
              x = x
    return(x)
  },
  value.var = 'Variant_Classification',
  fill = '',
  drop = FALSE
rownames(oncomat) = oncomat$Hugo_Symbol
oncomat <- oncomat[,-1]</pre>
hugo <- as.data.frame(table(mutation.data$Hugo_Symbol))</pre>
oncomat.ordered <- oncomat[order(-hugo$Freq),]</pre>
mat <- oncomat.ordered</pre>
mat[mat!=""]=1
mat[mat==""]=0
mat <- apply(mat, 2 ,as.numeric)</pre>
mat <- as.matrix(mat)</pre>
rownames(mat) <- row.names(oncomat.ordered)</pre>
reduce.mat <- mat[1:4,]</pre>
res <- pheatmap(reduce.mat,</pre>
         cluster_rows = F,
          show colnames=FALSE)
```

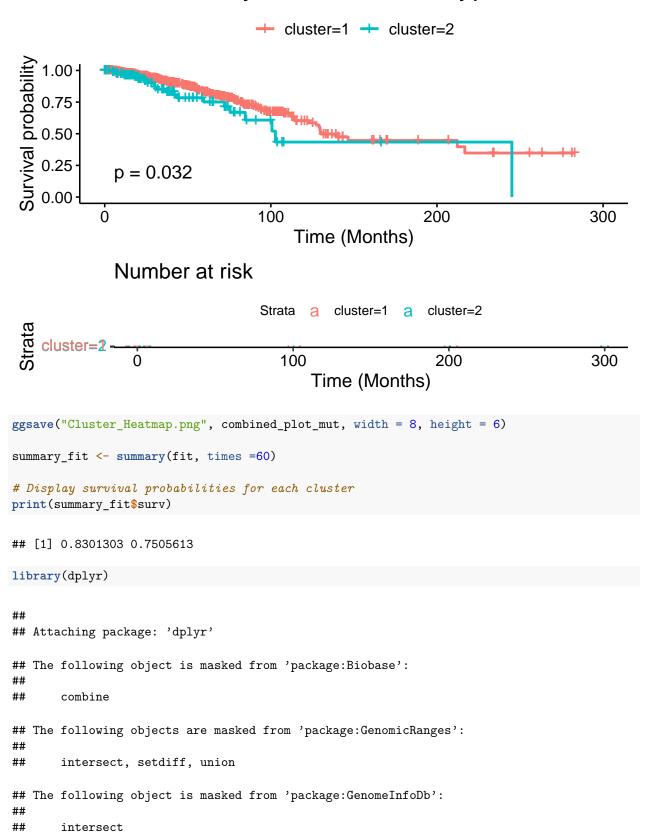


```
ggsave("Mutation_Heatmap.png", res$gtable, width = 7, height = 4)
cluster <- as.data.frame(cutree(res$tree_col,k = 2))</pre>
library("TCGAbiolinks")
library("survival")
library("survminer")
## Loading required package: ggpubr
##
## Attaching package: 'survminer'
## The following object is masked from 'package:survival':
##
##
       myeloma
library("SummarizedExperiment")
## Loading required package: MatrixGenerics
## Loading required package: matrixStats
## Attaching package: 'MatrixGenerics'
## The following objects are masked from 'package:matrixStats':
##
##
       colAlls, colAnyNAs, colAnys, colAvgsPerRowSet, colCollapse,
       colCounts, colCummaxs, colCummins, colCumprods, colCumsums,
##
       colDiffs, colIQRDiffs, colIQRs, colLogSumExps, colMadDiffs,
##
##
       colMads, colMaxs, colMeans2, colMedians, colMins, colOrderStats,
##
       colProds, colQuantiles, colRanges, colRanks, colSdDiffs, colSds,
##
       colSums2, colTabulates, colVarDiffs, colVars, colWeightedMads,
##
       colWeightedMeans, colWeightedMedians, colWeightedSds,
       colWeightedVars, rowAlls, rowAnyNAs, rowAnys, rowAvgsPerColSet,
##
##
       rowCollapse, rowCounts, rowCummaxs, rowCummins, rowCumprods,
##
       rowCumsums, rowDiffs, rowIQRDiffs, rowIQRs, rowLogSumExps,
##
       rowMadDiffs, rowMads, rowMaxs, rowMeans2, rowMedians, rowMins,
##
       rowOrderStats, rowProds, rowQuantiles, rowRanges, rowRanks,
       rowSdDiffs, rowSds, rowSums2, rowTabulates, rowVarDiffs, rowVars,
##
##
       rowWeightedMads, rowWeightedMeans, rowWeightedMedians,
##
       rowWeightedSds, rowWeightedVars
## Loading required package: GenomicRanges
## Loading required package: stats4
## Loading required package: BiocGenerics
```

```
##
## Attaching package: 'BiocGenerics'
## The following object is masked from 'package:gridExtra':
##
##
       combine
## The following objects are masked from 'package:stats':
##
##
       IQR, mad, sd, var, xtabs
## The following objects are masked from 'package:base':
##
##
       anyDuplicated, aperm, append, as.data.frame, basename, cbind,
##
       colnames, dirname, do.call, duplicated, eval, evalq, Filter, Find,
       get, grep, grepl, intersect, is.unsorted, lapply, Map, mapply,
##
       match, mget, order, paste, pmax, pmax.int, pmin, pmin.int,
##
##
       Position, rank, rbind, Reduce, rownames, sapply, setdiff, table,
##
       tapply, union, unique, unsplit, which.max, which.min
## Loading required package: S4Vectors
## Attaching package: 'S4Vectors'
## The following object is masked from 'package:utils':
##
       findMatches
##
## The following objects are masked from 'package:base':
##
##
       expand.grid, I, unname
## Loading required package: IRanges
## Loading required package: GenomeInfoDb
## Loading required package: Biobase
## Welcome to Bioconductor
##
##
       Vignettes contain introductory material; view with
##
       'browseVignettes()'. To cite Bioconductor, see
       'citation("Biobase")', and for packages 'citation("pkgname")'.
##
## Attaching package: 'Biobase'
## The following object is masked from 'package:MatrixGenerics':
##
##
       rowMedians
```

```
## The following objects are masked from 'package:matrixStats':
##
       anyMissing, rowMedians
##
rownames(cluster) <- substr(rownames(cluster), 1, 12)</pre>
clinical.data$deceased = clinical.data$OS_STATUS == "1:DECEASED"
clinical_info <- clinical.data[, c("PATIENT_ID", "OS_MONTHS", "deceased")]</pre>
clinical_info$cluster <- cluster$`cutree(res$tree_col, k = 2)`</pre>
Surv(clinical_info$0S_MONTHS, clinical_info$deceased) ~ cluster
## Surv(clinical_info$OS_MONTHS, clinical_info$deceased) ~ cluster
fit = survfit(Surv(OS_MONTHS, deceased) ~ cluster, data=clinical_info)
mut_surve <- ggsurvplot(fit, data=clinical_info, pval=T, risk.table=T, risk.table.col="strata", risk.ta</pre>
mut_surve$plot <- mut_surve$plot +</pre>
 theme(
    plot.title = element_text(size = 21), # Title size
    legend.text = element_text(size = 12),
                                                          # Legend text size
    legend.title = element_text(size = 12)
                                                         # Legend title size
 labs(color = NULL, fill = NULL, linetype = NULL)
 mut_surve$table <- mut_surve$table</pre>
 theme(legend.position = "none")
## List of 1
## $ legend.position: chr "none"
## - attr(*, "class")= chr [1:2] "theme" "gg"
## - attr(*, "complete")= logi FALSE
## - attr(*, "validate")= logi TRUE
combined_plot_mut <- grid.arrange(mut_surve$plot, mut_surve$table, ncol = 1, heights = c(2, 1))
```

Survival Anlaysis of Mutation Type



```
## The following objects are masked from 'package: IRanges':
##
##
       collapse, desc, intersect, setdiff, slice, union
## The following objects are masked from 'package:S4Vectors':
##
##
       first, intersect, rename, setdiff, setequal, union
## The following objects are masked from 'package:BiocGenerics':
##
##
       combine, intersect, setdiff, union
## The following object is masked from 'package:matrixStats':
##
##
       count
## The following object is masked from 'package:gridExtra':
##
##
       combine
##
  The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
clinical.data$deceased = clinical.data$OS_STATUS == "1:DECEASED"
filtered_mutations <- mutation.data[mutation.data$Hugo_Symbol %in% top_50_mutations$Gene[1:5], ]
clinical_info <- clinical.data[, c("PATIENT_ID", "OS_MONTHS", "deceased")]</pre>
colnames(filtered_mutations)[colnames(filtered_mutations) == "Patient_ID"] <- "PATIENT_ID"</pre>
# Merge the filtered mutations with the clinical data based on Patient_ID
merged_data <- merge(filtered_mutations, clinical_info, by = "PATIENT_ID", all.x = TRUE)
unique_merged_data <- merged_data %>%
  distinct(PATIENT_ID, Hugo_Symbol, .keep_all = TRUE)
table(unique_merged_data$Hugo_Symbol)
##
   KMT2C MUC16 PIK3CA
                                   TTN
##
                          TP53
##
       57
              72
                    178
                           176
                                   120
Surv(unique_merged_data$0S_MONTHS, unique_merged_data$deceased) ~ unique_merged_data$Hugo_Symbol
```

```
## Surv(unique_merged_data$0S_MONTHS, unique_merged_data$deceased) ~

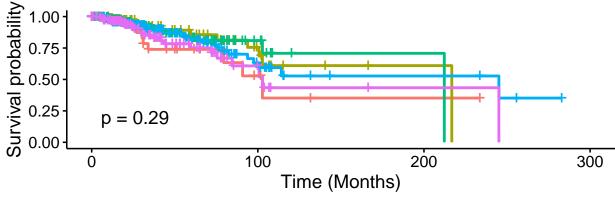
## unique_merged_data$Hugo_Symbol

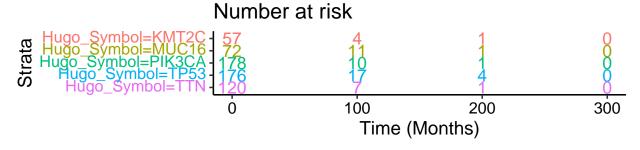
fit = survfit(Surv(0S_MONTHS, deceased) ~ Hugo_Symbol, data=unique_merged_data)

ggsurvplot(fit, data=unique_merged_data, pval=T, risk.table=T, risk.table.col="strata", risk.table.heig

Survival Anlaysis of Mutation Type

o_Symbol=KMT2C + Hugo_Symbol=MUC16 + Hugo_Symbol=PIK3CA + Hugo_Symbol=TP5:
```

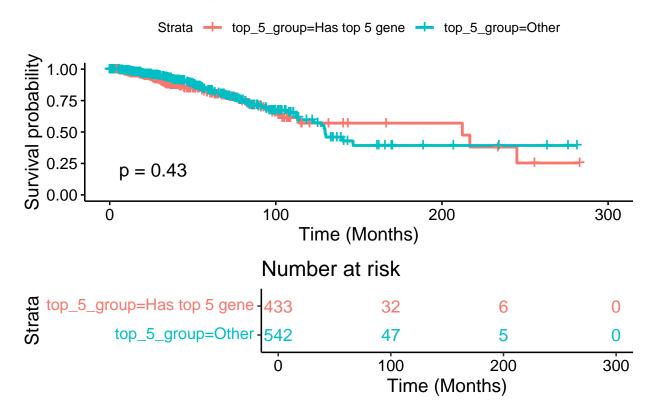


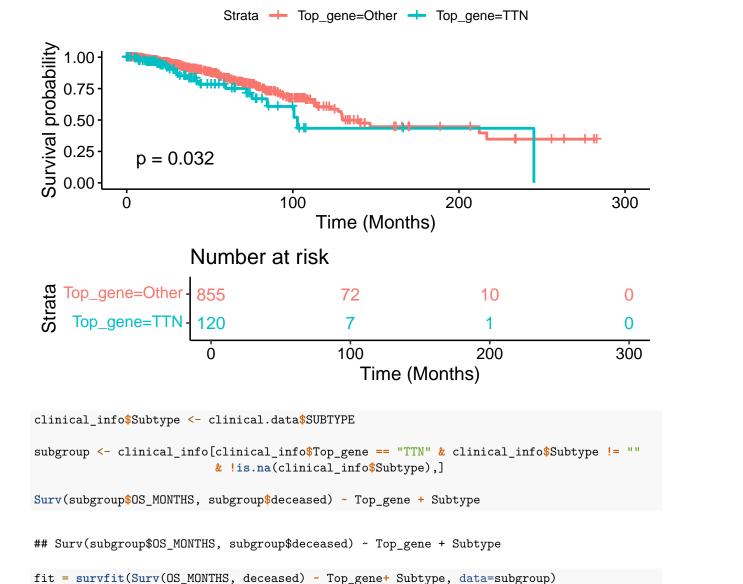


```
## Surv(clinical_info$0S_MONTHS, clinical_info$deceased) ~ top_5_group

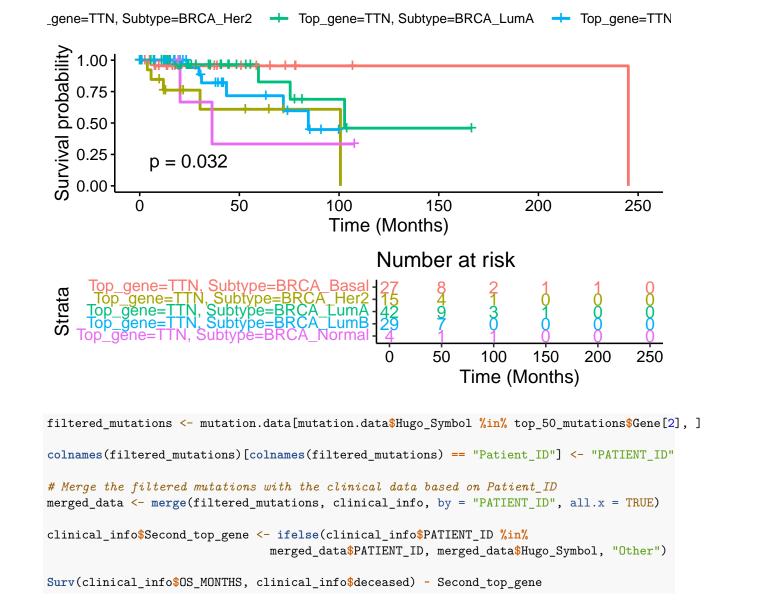
fit = survfit(Surv(OS_MONTHS, deceased) ~ top_5_group, data=clinical_info)

ggsurvplot(fit, data=clinical_info, pval=T, risk.table=T, risk.table.col="strata", risk.table.height=0...
```





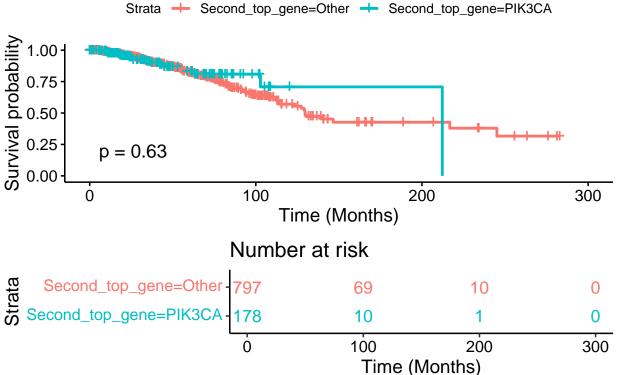
ggsurvplot(fit, data=subgroup, pval=T, risk.table=T, risk.table.col="strata", risk.table.height=0.35, t

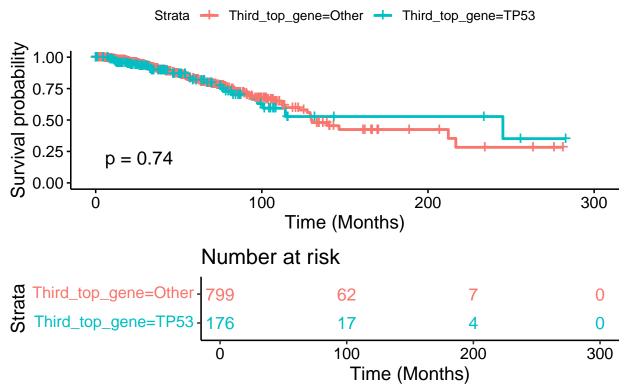


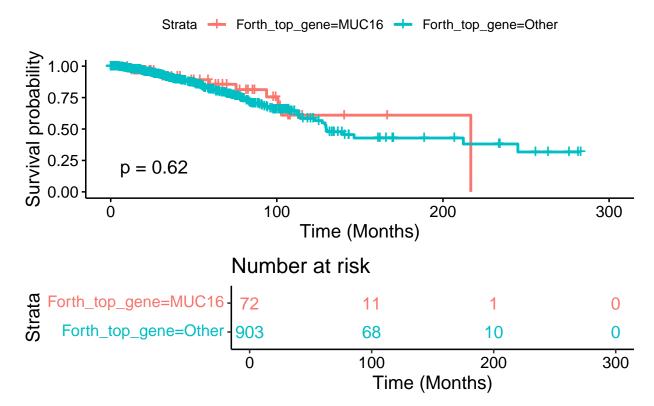
ggsurvplot(fit, data=clinical_info, pval=T, risk.table=T, risk.table.col="strata", risk.table.height=0..

Surv(clinical_info\$OS_MONTHS, clinical_info\$deceased) ~ Second_top_gene

fit = survfit(Surv(OS_MONTHS, deceased) ~ Second_top_gene, data=clinical_info)







#TTN