# R markdown Survival Analysis

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### Introduction

The following code was used to perform the various survival analyses for this research. Please see .... for the code to download and process the Forest Service data. The UMN-PALS data (Fleischman et al., 2020) can be downloaded from here: https://conservancy.umn.edu/handle/11299/211669

#### Load libraries

```
library(tidyverse)
## -- Attaching packages ------ tidyverse 1.3.0 --
## v ggplot2 3.3.3
                     v purrr
                              0.3.4
## v tibble 3.1.0
                     v dplyr
                              1.0.5
## v tidyr 1.1.3
                     v stringr 1.4.0
## v readr
          1.4.0
                    v forcats 0.5.1
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                   masks stats::lag()
library(forcats)
library(survminer)
## Loading required package: ggpubr
library(survival)
library(lubridate)
## Attaching package: 'lubridate'
## The following objects are masked from 'package:base':
##
##
      date, intersect, setdiff, union
library(reshape2)
## Attaching package: 'reshape2'
## The following object is masked from 'package:tidyr':
##
##
      smiths
```

```
library(ggplot2)
library(ggpubr)

# Set your working directory to wherever you placed your processed FS-PALS dataset.

setwd("/Users/kathrynmurenbeeld/Desktop/Survival_Analysis/Oct_presentation/")

# Set up to see more rows of output if desired

options(max.print = 10000)
```

#### Load the data

Data from post-processing. See S3.1 and S3.2.

```
## Load the data for survival analysis
df_fin <- read.csv("df_c20201020_v02.csv")</pre>
df_fin <- df_fin %>%
  select(-X)
## Replace "Inf" in the project delay column with NA
df_fin$proj.delay[is.infinite(df_fin$proj.delay)] <- NA</pre>
## Bring in binomial data as a factor
df_fin$REGION <- as.factor(df_fin$REGION)</pre>
df_fin$LITIGATED <- as.factor(df_fin$LITIGATED)</pre>
df_fin$APPEALED <- as.factor(df_fin$APPEALED)</pre>
df_fin$size <- as.factor(df_fin$size)</pre>
## Relevel the size data in order from small to extra-large
df_fin <- df_fin %>%
  mutate(size = fct_relevel(size,
            "small", "medium", "large", "x-large"))
## Create different data frames with Regions releveled.
## This is required to test the impact of using different
## Regions as a baseline or reference in the Cox proportional
## analysis.
df_fin2 <- df_fin %>%
  mutate(REGION = fct_relevel(REGION,
            "2", "1", "3", "4", "5", "6"))
df_fin3 <- df_fin %>%
  mutate(REGION = fct_relevel(REGION,
            "3", "1", "2", "4", "5", "6"))
df_fin4 <- df_fin %>%
  mutate(REGION = fct_relevel(REGION,
            "4", "1", "2", "3", "5", "6"))
df_fin5 <- df_fin %>%
  mutate(REGION = fct_relevel(REGION,
            "5", "1", "2", "3", "4", "6"))
```

#### Check for data correlations

## Why care?

When it comes to USFS project planning and inititaion there is some anecdotal evidence for NEPA causing delays...but very few quantitative studies.

Fuel, weather, and topography comprise the three legs of the Wildland Fire Behavior Triangle. Of those components, fuel is the only factor we have the ability to manage. Yet year after year fuels accumulate in our forests as management projects go neglected, delayed or obstructed. There is a clear correlation between the decline in timber harvests experienced on our National Forests and the increase in intensity and size of wildfires over the past three decades—both of which have had lasting impacts on the economic vitality of Western, rural communities. While our forests burn, our economic, recreational, and aesthetic capital burns with it.

While we recognize the ecological role wildfires can play in ecosystems, the severity and intensity of wildfires supersede that which should be occurring. Bureaucratic processes, burdensome regulations, external pressure, and judicial activism hamstrings our federal agencies from completing work on the ground in a timely manner. We must arm our federal land agencies with the tools they require to sustain the health and productivity of our nation's forests. Doing so is compatible with efforts to reduce emissions, as well-managed forests and the buildings constructed by the sustainable wood products that come from them have the potential to sequester carbon.

Figure 1: Recent letter to Congress from Western Congressional Causcus

## Survival Analysis

"Time to Event"

- Time until an individual dies (or is cured!).
- Time until a kitten or puppy is adopted.
- Any time an outcome is a duration
- Requires a start date, end date, duration (time between), if the event occurred, and the time of observation.



#### Different approaches

- Kaplan-Meier -> non-parametric, predictive
- Cox Proportional Hazards -> semi-parametric, use to estimate effect of covariate, not predictive

## Survival Analysis - Kaplan Meier

### All data - Kaplan Meier (KM) Estimate

Using all the data pooled.

```
# Use the survfit function from the survival package to
# calculate the KM estimate

fit_delay_all <- survfit(Surv(proj.delay, INITIATED) ~ 1, data = df_fin)
fit_delay_all

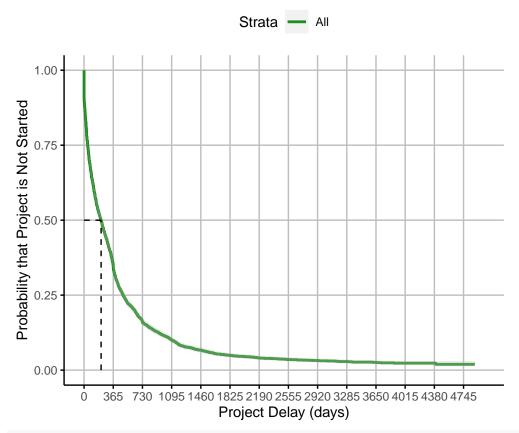
## Call: survfit(formula = Surv(proj.delay, INITIATED) ~ 1, data = df_fin)
##
## n events median 0.95LCL 0.95UCL
## 3557 3289 213 194 228

#summary(fit_delay_all)</pre>
```

### Survival Analysis - All - Kaplan Meier Curves

All data

```
## ggsurvvplot will create survival curves from the KM model created above.
km_fit_all <- ggsurvplot(fit_delay_all,</pre>
           conf.int = TRUE,
           risk.table = FALSE,
           risk.table.col = "strata",
           surv.median.line = "hv",
           break.time.by = 365,
           risk.table.y.text=FALSE,
           censor = FALSE,
           ylab = "Probability that Project is Not Started",
           xlab = "Project Delay (days)",
           palette = c("forestgreen"),
           surv.plot.height = 1,
           ggtheme = theme(aspect.ratio = 0.75,
                           axis.line = element_line(colour = "black"),
                           panel.grid.major = element_line(colour = "grey"),
                           panel.border = element_blank(),
                           panel.background = element_blank()),
           tables.theme = theme(aspect.ratio = 0.06)
print(km_fit_all)
```

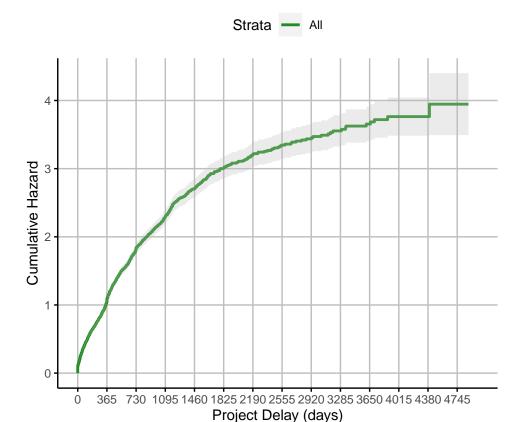


 $\#ggsave (\ "/Users/kathrynmurenbeeld/Desktop/Survival\_Analysis/WRITING/FIGS/km\_fit\_all.pdf",\ print(km\_fit\_all.pdf))$ 

#### Survival Analysis - All - Kaplan Meier - Cumulative Hazard

All data

```
# ggsurvplot will create the cumulative hazard plots based on the KM model created above.
km_haz_all <- ggsurvplot(fit_delay_all,</pre>
           conf.int = TRUE,
           risk.table = FALSE,
           risk.table.col = "strata",
           break.time.by = 365,
           risk.table.y.text=FALSE,
           censor = FALSE,
           ylab = "Cumulative Hazard",
           xlab = "Project Delay (days)",
           palette = c("forestgreen"),
           fun = "cumhaz",
           surv.plot.height = 1,
           ggtheme = theme(aspect.ratio = 0.75,
                          axis.line = element_line(colour = "black"),
                           panel.grid.major = element_line(colour = "grey"),
                           panel.border = element_blank(),
                           panel.background = element_blank()),
                           theme(aspect.ratio = 0.06)
           tables.theme =
print(km_haz_all)
```



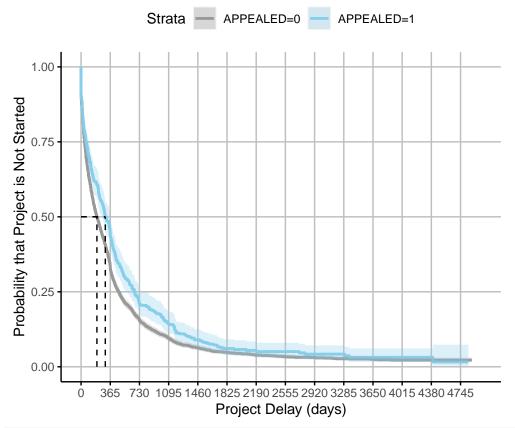
 $\#ggsave (\ "/Users/kathrynmurenbeeld/Desktop/Survival\_Analysis/WRITING/FIGS/km\_haz\_all.pdf",\ print(km\_haz_all.pdf))$ 

### Survival Analysis - Appealed? - KM Estimate

Data grouped by appealed and non-appealed projects.

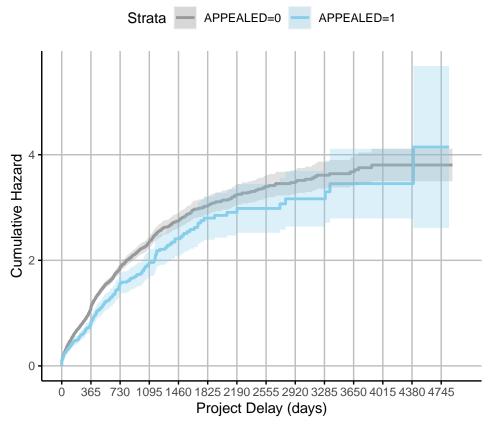
```
# Use the survfit function from the survival package to
# calculate the KM estimate for data grouped by appealed or non-appealed.
# This code chunk and the next two act as a template for the K-M estimation.
fit_app <- survfit(Surv(proj.delay, INITIATED) ~ APPEALED, data = df_fin)</pre>
fit_app_table <- summary(fit_app)</pre>
fit_app
## Call: survfit(formula = Surv(proj.delay, INITIATED) ~ APPEALED, data = df_fin)
##
##
                 n events median 0.95LCL 0.95UCL
## APPEALED=0 3184
                      2937
                              198
                                      182
                                               220
## APPEALED=1 373
                      352
                              303
                                      258
                                               365
```

### Survival Analysis - Appealed? - KM Curve



 $\#ggsave (\ "/Users/kathrynmurenbeeld/Desktop/Survival\_Analysis/WRITING/FIGS/km\_curv\_app.pdf",\ print(km\_curv\_app.pdf))$ 

### Survival Analysis - Appealed? - KM Cumulative Hazard



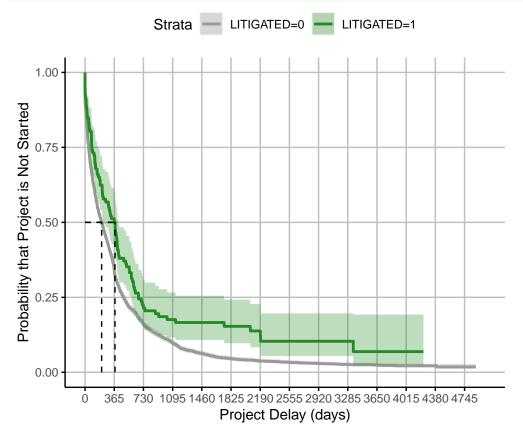
#ggsave( "/Users/kathrynmurenbeeld/Desktop/Survival\_Analysis/WRITING/FIGS/km\_haz\_app.pdf", print(km\_haz\_app.pdf)

### Survival Analysis - Litigated? - KM Estimate

Data grouped by litigated and non-litigated projects.

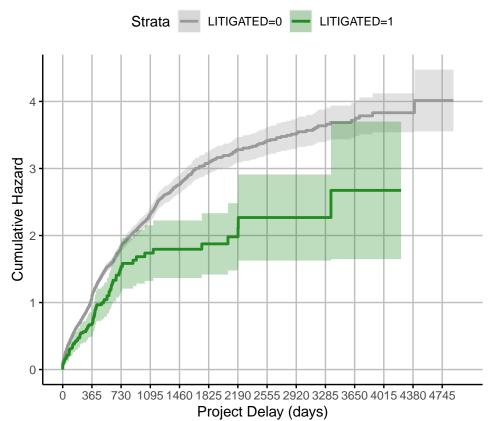
### Survival Analysis - Litigated? - KM Curve

```
km_curv_lit <- ggsurvplot(fit_lit,</pre>
           conf.int = TRUE,
           risk.table = FALSE,
           risk.table.col = "strata",
           surv.median.line = "hv",
           break.time.by = 365,
           risk.table.y.text=FALSE,
           tables.height = 0.3,
           censor = FALSE,
           ylab = "Probability that Project is Not Started",
           xlab = "Project Delay (days)",
           palette = c("#999999", "forestgreen"),
           surv.plot.height = 1,
           ggtheme = theme(aspect.ratio = 0.75,
                           axis.line = element_line(colour = "black"),
                           panel.grid.major = element_line(colour = "grey"),
                           panel.border = element_blank(),
                           panel.background = element_blank()),
           tables.theme = theme(aspect.ratio = 0.06)
print(km_curv_lit)
```



### Survival Analysis - Litigated? - KM Cumulative Hazard

```
km_haz_lit <- ggsurvplot(fit_lit,</pre>
           conf.int = TRUE,
           risk.table = FALSE,
           risk.table.col = "strata",
           break.time.by = 365,
           risk.table.y.text=FALSE,
           censor = FALSE,
           ylab = "Cumulative Hazard",
           xlab = "Project Delay (days)",
           palette = c("#999999", "forestgreen"),
           fun = "cumhaz",
           surv.plot.height = 1,
           ggtheme = theme(aspect.ratio = 0.75,
                           axis.line = element_line(colour = "black"),
                           panel.grid.major = element_line(colour = "grey"),
                           panel.border = element_blank(),
                           panel.background = element_blank()),
           tables.theme = theme(aspect.ratio = 0.06)
print(km_haz_lit)
```



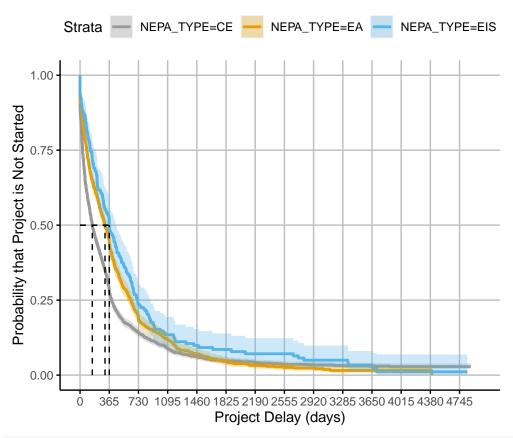
### Survival Analysis - NEPA Type - KM Estimate

Data grouped by NEPA type for projects. EIS = Environmental Impact Statement, EA = Environmental Assessment, CE = Categorical Exclusion

```
fit_delay <- survfit(Surv(proj.delay, INITIATED) ~ NEPA_TYPE, data = df_fin)</pre>
fit_nepa_table <- summary(fit_delay)</pre>
fit_delay
## Call: survfit(formula = Surv(proj.delay, INITIATED) ~ NEPA_TYPE, data = df_fin)
##
                    n events median 0.95LCL 0.95UCL
## NEPA TYPE=CE 2266
                         2110
                                 153
                                         139
## NEPA_TYPE=EA 1073
                          988
                                 312
                                         285
                                                  350
                                 365
                                         303
                                                  455
## NEPA_TYPE=EIS 218
                          191
```

#### Survival Analysis - NEPA Type - KM Curve

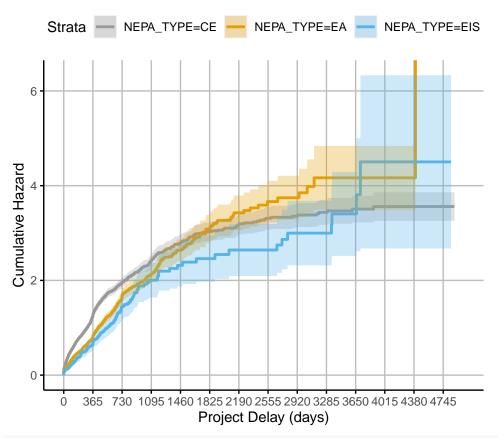
```
km_curv_nepa <- ggsurvplot(fit_delay,</pre>
           conf.int = TRUE,
           risk.table = FALSE,
           risk.table.col = "strata",
           surv.median.line = "hv",
           break.time.by = 365,
           risk.table.y.text=FALSE,
           tables.height = 0.3,
           censor = FALSE,
           ylab = "Probability that Project is Not Started",
           xlab = "Project Delay (days)",
           palette = c("#999999", "#E69F00", "#56B4E9"),
           surv.plot.height = 1,
           ggtheme = theme(aspect.ratio = 0.75,
                           axis.line = element_line(colour = "black"),
                           panel.grid.major = element line(colour = "grey"),
                           panel.border = element_blank(),
                           panel.background = element_blank()),
           tables.theme = theme(aspect.ratio = 0.06)
print(km_curv_nepa)
```



 $\#ggsave (\ "/Users/kathrynmurenbeeld/Desktop/Survival\_Analysis/WRITING/FIGS/km\_curv\_nepa.pdf",\ print(km\_curv\_nepa.pdf'',\ print(km\_curv\_nepa.pdf''),\ print(km\_curv\_nepa.pdf'')),\ print(km\_curv\_nepa.pdf''),\ print(km\_curv\_nepa.pdf'')),\ print(km\_curv\_n$ 

### Survival Analysis - NEAP Type - KM Cumulative Hazard

```
km_haz_nepa <- ggsurvplot(fit_delay,</pre>
           conf.int = TRUE,
           risk.table = FALSE,
           risk.table.col = "strata",
           break.time.by = 365,
           risk.table.y.text=FALSE,
           tables.height = 0.3,
           censor = FALSE,
           ylab = "Cumulative Hazard",
           xlab = "Project Delay (days)",
           palette = c("#999999", "#E69F00", "#56B4E9"),
           fun = "cumhaz",
           surv.plot.height = 1,
           ggtheme = theme(aspect.ratio = 0.75,
                           axis.line = element_line(colour = "black"),
                           panel.grid.major = element_line(colour = "grey"),
                           panel.border = element_blank(),
                           panel.background = element_blank()),
                           theme(aspect.ratio = 0.06)
           tables.theme =
print(km haz nepa)
```



 $\#ggsave (\ "/Users/kathrynmurenbeeld/Desktop/Survival\_Analysis/WRITING/FIGS/km\_haz\_nepa.pdf",\ print(km\_halysis/WRITING/FIGS/km\_haz\_nepa.pdf",\ print(km\_halysis/WRITING/FIGS/km_haz\_nepa.pdf",\ print(km\_halysis/WRITING/FIGS/km_haz\_nepa.pdf"),\ print(km\_halysis/WRITING/FIGS/km_haz\_nepa.pdf"),\ print(km\_halysis/WRITING/FIGS/km_haz\_nepa.pdf"),\ print(km\_halysis/WRITING/FIGS/km_haz\_nepa.pdf"),\ print(km\_halysis/WRITING/FIGS/km_haz\_nepa.pdf"),\ print(km\_halysis/WRITING/FIGS/km_haz\_nepa.pdf"),\ print(km\_halysis/WRITING/FIGS/km_haz\_nepa.pdf"),\ print(km\_halysis/WRITING/FIGS/km_haz)),\ print(km\_halysis/WRITING/FIGS/km_haz\_nepa.pdf"),\ print(km\_halysis/WRITING/FIGS/km_haz\_nepa.pdf"),\ print(km\_halysis/WRITING/FIGS/km_haz\_nepa.pdf"),\ print(km\_halysis/WRITING/FIGS/km_haz\_nepa.pdf"),\ print(km\_halysis/WRITING/FIGS/km_haz\_nepa.pdf"),\ print(km\_halysis/WRITING/FIGS/km_haz\_nepa.pdf"),\ print(km\_halysis/WRITING/FIGS/km_haz)),\ print(km\_halysis/WRITI$ 

#### Survival Analysis - Regions - KM Estimate

```
Data grouped by project location. United Stated Forest Service (USFS) Regions 1-6

fit_reg <- survfit(Surv(proj.delay, INITIATED) ~ REGION, data = df_fin)

fit_reg_table <- summary(fit_reg)

fit_reg

## Call: survfit(formula = Surv(proj.delay, INITIATED) ~ REGION, data = df_fin)

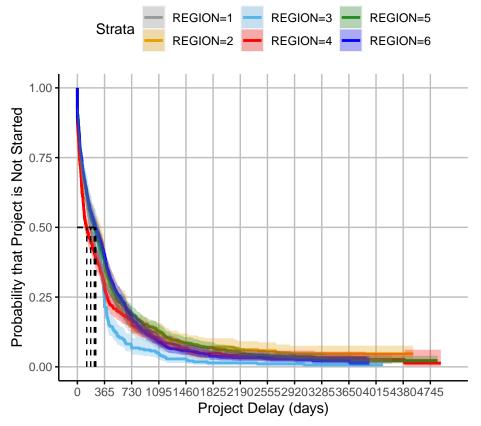
##

##

n events median 0.95LCL 0.95UCL
```

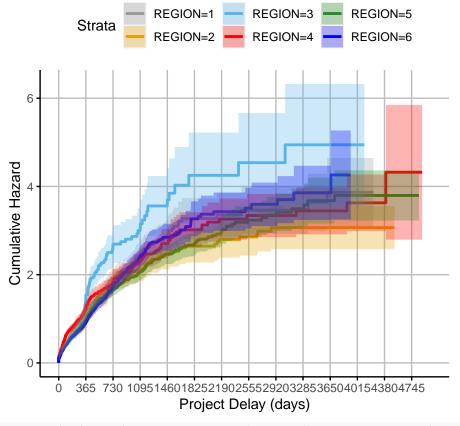
```
##
## REGION=1 538
                    503
                            179
                                    153
                                             218
                            245
                                    197
                                             323
## REGION=2 455
                    405
## REGION=3 301
                    293
                            245
                                    183
                                             287
## REGION=4 540
                    507
                            126
                                     94
                                             173
## REGION=5 928
                    845
                            228
                                    195
                                             261
## REGION=6 795
                                    205
                                             294
                    736
                            245
```

#### Survival Analysis - Regions - KM Curve



 $\#ggsave (\ "/Users/kathrynmurenbeeld/Desktop/Survival\_Analysis/WRITING/FIGS/km\_curv\_reg.pdf",\ print(km\_curv\_reg.pdf))$ 

### Survival Analysis - Regions - KM Cumulative Hazard



#ggsave( "/Users/kathrynmurenbeeld/Desktop/Survival\_Analysis/WRITING/FIGS/km\_haz\_reg.pdf", print(km\_haz

#### Survival Analysis - Size - KM Estimate

Data grouped by a project's cumulative size. Size categories = small, medium, large, and extra-large.

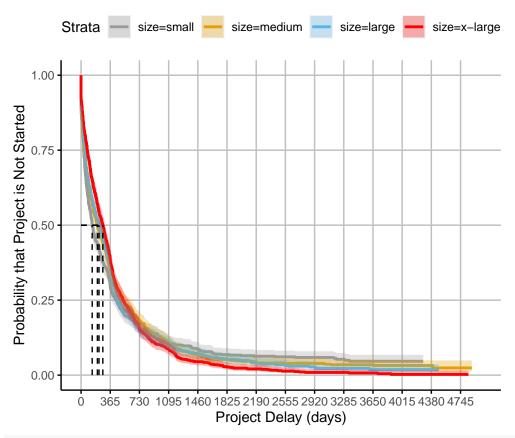
Sizes are categorized based on quartile ranges + x-large > 2670 acres + large 768-2670 acres + medium 174-768 acres + small < 174 acres

```
fit_size <- survfit(Surv(proj.delay, INITIATED) ~ size, data = df_fin)
fit_size_table <- summary(fit_size)
fit_size</pre>
```

```
## Call: survfit(formula = Surv(proj.delay, INITIATED) ~ size, data = df_fin)
##
                  n events median 0.95LCL 0.95UCL
##
## size=small
                       803
                              139
                                      123
                888
## size=medium 890
                       811
                              204
                                      171
                                              245
## size=large
              888
                       814
                              224
                                      187
                                              261
## size=x-large 891
                       861
                              272
                                      235
                                              300
```

#### Survival Analysis - Size - KM Curve

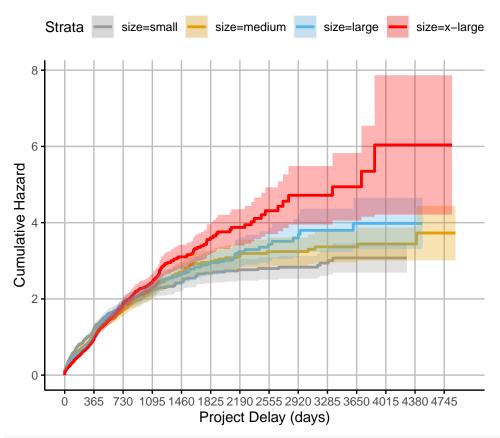
```
km_curv_size <- ggsurvplot(fit_size,</pre>
           conf.int = TRUE,
           risk.table = FALSE,
           risk.table.col = "strata",
           surv.median.line = "hv",
           break.time.by = 365,
           risk.table.y.text=FALSE,
           tables.height = 0.3,
           censor = FALSE,
           ylab = "Probability that Project is Not Started",
           xlab = "Project Delay (days)",
           palette = c("#999999", "#E69F00", "#56B4E9", "red"),
           surv.plot.height = 1,
           ggtheme = theme(aspect.ratio = 0.75,
                           axis.line = element_line(colour = "black"),
                           panel.grid.major = element_line(colour = "grey"),
                           panel.border = element_blank(),
                           panel.background = element_blank()),
           tables.theme = theme(aspect.ratio = 0.06)
print(km_curv_size)
```



 $\#ggsave (\ "/Users/kathrynmurenbeeld/Desktop/Survival\_Analysis/WRITING/FIGS/km\_curv\_size.pdf",\ print(km\_curv\_size.pdf)))$ 

#### Survival Analysis - Size - KM Cumulative Hazard

```
km_haz_size <- ggsurvplot(fit_size,</pre>
           conf.int = TRUE,
           risk.table = FALSE,
           risk.table.col = "strata",
           break.time.by = 365,
           risk.table.y.text=FALSE,
           tables.height = 0.3,
           censor = FALSE,
           ylab = "Cumulative Hazard",
           xlab = "Project Delay (days)",
           palette = c("#999999", "#E69F00", "#56B4E9", "red"),
           fun = "cumhaz",
           surv.plot.height = 1,
           ggtheme = theme(aspect.ratio = 0.75,
                      axis.line = element_line(colour = "black"),
                           panel.grid.major = element_line(colour = "grey"),
                           panel.border = element_blank(),
                           panel.background = element_blank()),
                           theme(aspect.ratio = 0.06)
           tables.theme =
print(km haz size)
```



 $\#ggsave (\ "/Users/kathrynmurenbeeld/Desktop/Survival\_Analysis/WRITING/FIGS/km\_haz\_size.pdf",\ print(km\_haultigenerated))) + (km_haultigenerated) + (km_haulti$ 

#### Comparing survival curves using a log-rank approach

- Null hypothesis: no difference in survival between the groups
- Non-parametric test

Log-Rank - Appealed?

```
# The survdiff function in the survival package does this for you.
surv_diff_delay_app <- survdiff(Surv(proj.delay, INITIATED) ~ APPEALED, data = df_fin)</pre>
print(surv_diff_delay_app)
## Call:
## survdiff(formula = Surv(proj.delay, INITIATED) ~ APPEALED, data = df_fin)
##
                 N Observed Expected (O-E)^2/E (O-E)^2/V
                        2937
                                           1.58
## APPEALED=0 3184
                                 2870
                                                      12.6
## APPEALED=1 373
                         352
                                  419
                                          10.83
                                                      12.6
##
   Chisq= 12.6 on 1 degrees of freedom, p= 4e-04
Log-Rank - Litigated?
surv_diff_delay_lit <- survdiff(Surv(proj.delay, INITIATED) ~ LITIGATED, data = df_fin)</pre>
print(surv_diff_delay_lit)
## Call:
## survdiff(formula = Surv(proj.delay, INITIATED) ~ LITIGATED, data = df_fin)
```

```
##
##
                  N Observed Expected (0-E)^2/E (0-E)^2/V
                                  3154
## LITIGATED=0 3445
                        3194
                                           0.504
## LITIGATED=1 112
                          95
                                   135
                                          11.787
                                                       12.4
##
  Chisq= 12.4 on 1 degrees of freedom, p= 4e-04
Log-Rank - NEPA
surv_diff_delay_nepa <- survdiff(Surv(proj.delay, INITIATED) ~ NEPA_TYPE, data = df_fin)</pre>
print(surv diff delay nepa)
## Call:
## survdiff(formula = Surv(proj.delay, INITIATED) ~ NEPA_TYPE, data = df_fin)
##
##
                    N Observed Expected (O-E)^2/E (O-E)^2/V
## NEPA_TYPE=CE 2266
                                    1898
                           2110
                                              23.7
                                                         57.1
## NEPA_TYPE=EA 1073
                            988
                                    1135
                                              19.1
                                                         29.7
                                     256
## NEPA_TYPE=EIS 218
                            191
                                              16.4
                                                         18.0
##
  Chisq= 60.3 on 2 degrees of freedom, p= 8e-14
Log-Rank - Region
surv_diff_delay_reg <- survdiff(Surv(proj.delay, INITIATED) ~ REGION, data = df_fin)</pre>
print(surv_diff_delay_reg)
## Call:
## survdiff(formula = Surv(proj.delay, INITIATED) ~ REGION, data = df_fin)
##
##
              N Observed Expected (0-E)^2/E (0-E)^2/V
## REGION=1 538
                     503
                               477
                                        1.47
                                                  1.74
## REGION=2 455
                     405
                                        2.93
                                                  3.43
                               441
## REGION=3 301
                     293
                               241
                                       11.44
                                                 12.59
## REGION=4 540
                     507
                                        5.41
                               457
                                                  6.37
## REGION=5 928
                     845
                               905
                                        3.95
                                                  5.52
## REGION=6 795
                     736
                               769
                                        1.41
                                                  1.87
##
## Chisq= 27.1 on 5 degrees of freedom, p= 6e-05
Log-Rank - size
surv_diff_delay_size <- survdiff(Surv(proj.delay, INITIATED) ~ size, data = df_fin)</pre>
print(surv_diff_delay_size)
## Call:
## survdiff(formula = Surv(proj.delay, INITIATED) ~ size, data = df_fin)
##
##
                  N Observed Expected (O-E)^2/E (O-E)^2/V
## size=small
                888
                         803
                                   772
                                         1.25516 1.666100
## size=medium 890
                         811
                                   827
                                         0.29103 0.393744
## size=large
                888
                         814
                                   813
                                         0.00047 0.000633
                                         0.30047 0.415975
## size=x-large 891
                         861
                                   877
## Chisq= 1.9 on 3 degrees of freedom, p= 0.6
```

## Cox Proportional Hazards (Cox ph) Model

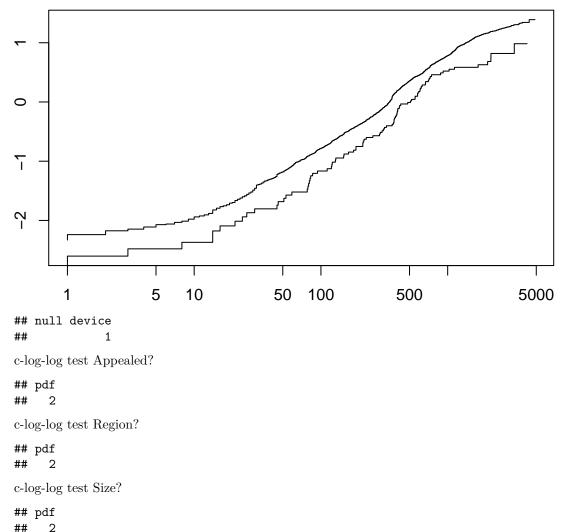
### Test Cox proportional hazards assumptions

There are 5 assumptions for using a Cox PH model

- 1. Non-informative censoring
- 2. Survival times, t, are independent
- 3. Hazards are proportional
- 4. ln(HR) is a linear function of numerical covariates (non in this study)
- 5. Covariate values don't change over time (eg. changing a treatment or dosage)
- 6. The baseline hazard (ie. hazard if all covariates were 0) is unspecified

### Cox PH - Test proportionality assumption

c-log-log test Litigated?



# Determining the Cox Proportional Hazards Model

Here we used a likelihood ratios test (LRT) to determine the final form of the Cox proportional hazards model.

```
## Call: coxph(formula = Surv(proj.delay, INITIATED) ~ 1, data = df_fin)
##
## Null model
    log likelihood= -23950.14
    n = 3557
Should we add NEPA type?
## Call:
## coxph(formula = Surv(proj.delay, INITIATED) ~ NEPA_TYPE, data = df_fin)
##
                   coef exp(coef) se(coef)
## NEPA_TYPEEA -0.24888
                          0.77967 0.03865 -6.439 1.21e-10
                          0.66841 0.07563 -5.327 1.00e-07
## NEPA_TYPEEIS -0.40285
##
## Likelihood ratio test=62.32 on 2 df, p=2.934e-14
## n= 3557, number of events= 3289
# Here we use the anova function to complete the LRT test.
anova(cox.cat2, cox.test, test="LRT")
## Analysis of Deviance Table
## Cox model: response is Surv(proj.delay, INITIATED)
## Model 1: ~ NEPA_TYPE
## Model 2: ~ 1
## loglik Chisq Df P(>|Chi|)
## 1 -23919
## 2 -23950 62.32 2 2.934e-14 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Yes
Should we include Litigated?
## Analysis of Deviance Table
## Cox model: response is Surv(proj.delay, INITIATED)
## Model 1: ~ NEPA_TYPE + LITIGATED
## Model 2: ~ NEPA TYPE
   loglik Chisq Df P(>|Chi|)
## 1 -23916
## 2 -23919 6.2551 1 0.01238 *
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Yes
Should we include Appealed?
## Analysis of Deviance Table
## Cox model: response is Surv(proj.delay, INITIATED)
## Model 1: ~ NEPA TYPE + LITIGATED + APPEALED
## Model 2: ~ NEPA_TYPE + LITIGATED
    loglik Chisq Df P(>|Chi|)
## 1 -23916
## 2 -23916 0.4222 1
                        0.5159
```

#### Should we include Region?

```
## Analysis of Deviance Table
## Cox model: response is Surv(proj.delay, INITIATED)
## Model 1: ~ NEPA_TYPE + LITIGATED + REGION
## Model 2: ~ NEPA_TYPE + LITIGATED
## loglik Chisq Df P(>|Chi|)
## 1 -23905
## 2 -23916 22.477 5 0.0004248 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Yes
```

#### Should we include size?

```
## Analysis of Deviance Table
## Cox model: response is Surv(proj.delay, INITIATED)
## Model 1: ~ NEPA_TYPE + LITIGATED + REGION + size
## Model 2: ~ NEPA_TYPE + LITIGATED + REGION
## loglik Chisq Df P(>|Chi|)
## 1 -23898
## 2 -23905 12.284 3 0.00647 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

The final Cox proportional hazards model should include all covariates except for appealed.

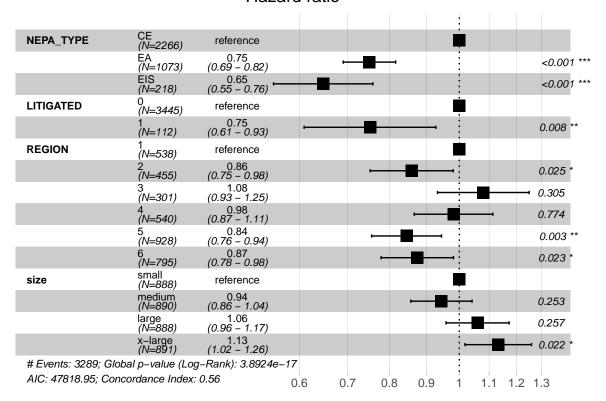
Need to create Cox models using each of the dataframes made with a different region set as the reference.

#### Create forest plots of Cox proportional hazards models.

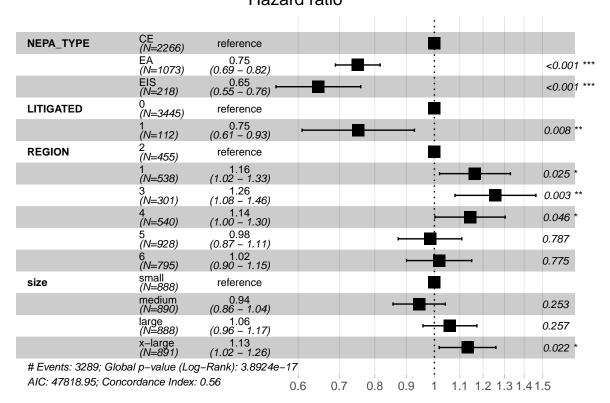
#### Cox PH Model - Hazard Ratios - Region 1 as Ref

```
## coxph(formula = Surv(proj.delay, INITIATED) ~ NEPA_TYPE + LITIGATED +
       REGION + size, data = df fin)
##
##
                    coef exp(coef) se(coef)
                                                 7.
## NEPA TYPEEA -0.28722
                           0.75035 0.04272 -6.722 1.79e-11
## NEPA TYPEEIS -0.43501
                           0.64726
                                    0.08107 -5.366 8.06e-08
## LITIGATED1
                -0.28542
                           0.75170
                                    0.10743 - 2.657
                                                    0.00789
## REGION2
                -0.15222
                           0.85880
                                    0.06776 -2.246
                                                    0.02468
## REGION3
                0.07668
                           1.07970
                                    0.07474 1.026
                                                    0.30493
## REGION4
                -0.01845
                           0.98172
                                    0.06413 -0.288
                                                    0.77357
## REGION5
                -0.16863
                           0.84482
                                    0.05698 -2.960
                                                    0.00308
## REGION6
                -0.13439
                           0.87425
                                    0.05901 -2.277
                                                    0.02277
## sizemedium
                -0.05720
                           0.94441
                                    0.04999 -1.144
                                                    0.25251
                 0.05836
                           1.06010
                                    0.05146 1.134
## sizelarge
                                                    0.25677
## sizex-large
                 0.12441
                           1.13248
                                    0.05425 2.293
## Likelihood ratio test=103.3 on 11 df, p=< 2.2e-16
## n= 3557, number of events= 3289
```

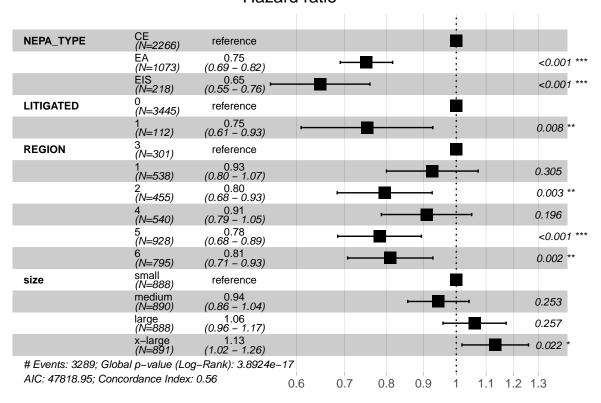
### Hazard ratio



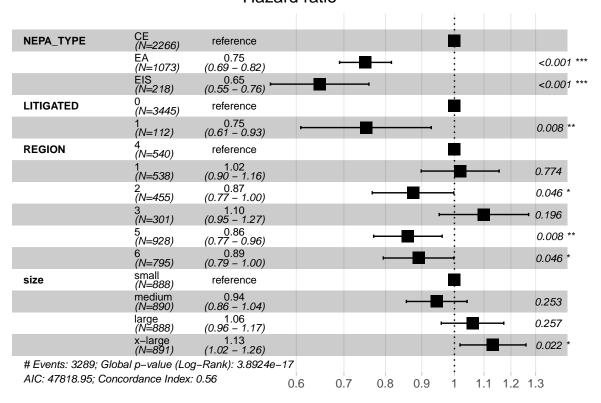
Cox PH Model - Hazard Ratios - Region 2 as Ref Hazard ratio



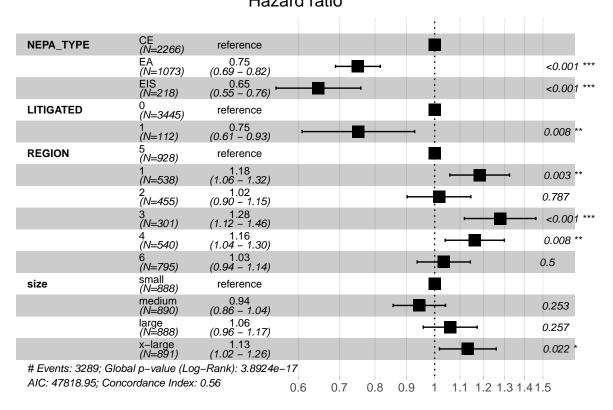
## Cox PH Model - Hazard Ratios - Region 3 as Ref Hazard ratio



## Cox PH Model - Hazard Ratios - Region 4 as Ref Hazard ratio



## Cox PH Model - Hazard Ratios - Region 5 as Ref Hazard ratio



## Cox PH Model - Hazard Ratios - Region 6 as Ref Hazard ratio

