

Homework 3: Survival Analysis + CLV (R)

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Libraries

```
library(survival)
library(flexsurv)

## Warning: package 'flexsurv' was built under R version 4.4.3

library(survminer)

## Warning: package 'survminer' was built under R version 4.4.3

## Loading required package: ggplot2

## Loading required package: ggpunr

##
## Attaching package: 'survminer'

## The following object is masked from 'package:survival':
## 
##     myeloma

library(ggplot2)
library(dplyr)

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
## 
##     filter, lag

## The following objects are masked from 'package:base':
## 
##     intersect, setdiff, setequal, union
```

```
library(tidyverse)
```

Loading the Dataset

```
df <- read.csv("telco.csv", stringsAsFactors = FALSE)

df$churn_flag <- ifelse(tolower(df$churn) == "yes", 1, 0)
df$tenure <- as.numeric(df$tenure)
```

Factoring the Categorical Variables

```
df$gender <- factor(df$gender)
df$ed <- factor(df$ed, levels = c("Did not complete high school", "High school degree", "Some college", "College graduate"))
df$custcat <- factor(df$custcat)
df$region <- factor(df$region)
```

Creating Survival Object

```
surv_obj <- Surv(time = df$tenure, event = df$churn_flag)
```

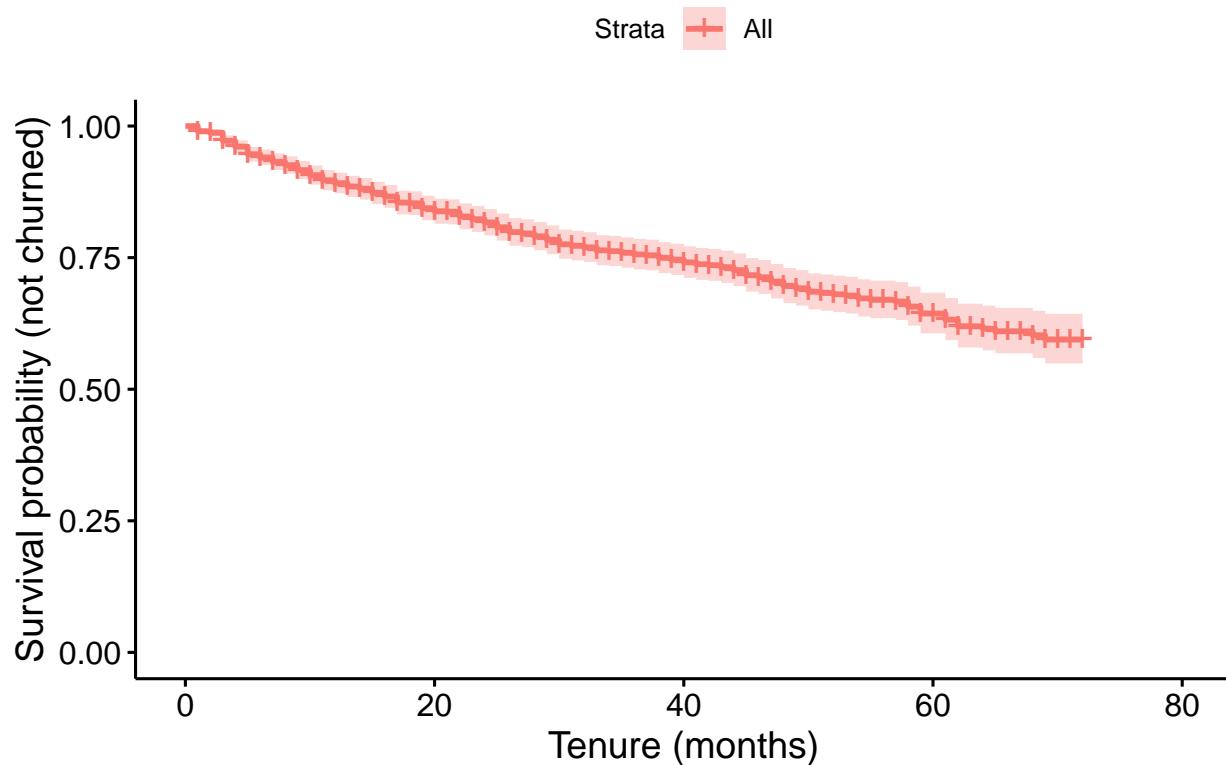
Kaplan-Meier

```
km <- survfit(surv_obj ~ 1, data = df)
km

## Call: survfit(formula = surv_obj ~ 1, data = df)
##
##      n events median 0.95LCL 0.95UCL
## [1,] 1000     274      NA      NA      NA

ggsurvplot(km, data = df, xlab = "Tenure (months)", ylab = "Survival probability (not churned)", title =
```

KM – All customers

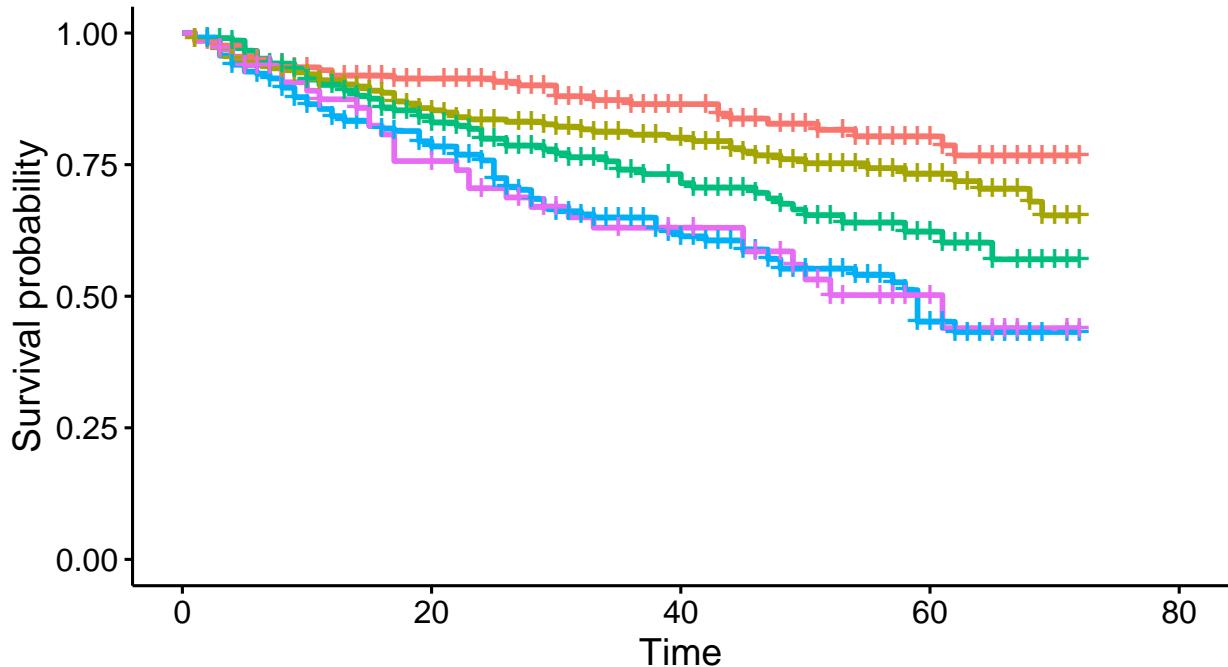


KM by Education

```
if(nlevels(df$ed) > 1){  
  ggsurvplot(survfit(surv_obj ~ ed, data = df), data = df,  
             legend.title = "Education",  
             title = "Kaplan-Meier by education (small sample)")  
}
```

Kaplan–Meier by education (small sample)

ot complete high school + ed=High school degree + ed=Some college + ed=College degree -



AFT Parametric Models

```
dists <- c("exponential", "weibull", "lognormal", "loglogistic") # survreg supports these names
models <- list()
for(dist in dists){
  formula <- as.formula("surv_obj ~ gender + ed + income")
  m <- try(survreg(formula, data = df, dist = dist), silent = TRUE)
  if(inherits(m, "try-error")){
    models[[dist]] <- NULL
  } else {
    models[[dist]] <- m
  }
}

flex_wb <- try(flexsurvreg(formula = surv_obj ~ gender + ed + income, data = df, dist = "weibull"), silent = TRUE)
if(!inherits(flex_wb, "try-error")) models[["flex_weibull"]] <- flex_wb
```

Model Comparison

```
aic_df <- data.frame(dist = character(0), AIC = numeric(0), stringsAsFactors = FALSE)
for(nm in names(models)){
  m <- models[[nm]]
```

```

    if(!is.null(m)){
      aic_df <- rbind(aic_df, data.frame(dist = nm, AIC = AIC(m)))
    }
  }
  aic_df <- aic_df %>% arrange(AIC)
  aic_df

```

```

##           dist      AIC
## 1 exponential 3142.414
## 2 flex_weibull 3144.414
## 3      weibull 3144.414
## 4 loglogistic 3144.505
## 5   lognormal 3145.703

```

Visualizing Survival Curves

```

plot_df_all <- data.frame()
time_grid <- seq(0, 80, by = 1)

get_surv_probs <- function(fit_obj, newdata, times){
  if(inherits(fit_obj, "flexsurvflex")){
    s <- summary(fit_obj, newdata = newdata, t = times, type = "survival")
    probs <- s[[1]]$est
    return(probs)
  } else if(inherits(fit_obj, "survreg")){
    lp <- predict(fit_obj, newdata = newdata, type = "response")
    distname <- fit_obj$dist
    require(flexsurv)
    fs <- try(flexsurvreg(formula = update(fit_obj$call$formula, . ~ gender + ed + income),
                           data = eval(fit_obj$call$data),
                           dist = distname), silent = TRUE)
    if(!inherits(fs, "try-error")){
      s <- summary(fs, newdata = newdata, t = times, type = "survival")
      return(s[[1]]$est)
    } else {
      return(rep(NA, length(times)))
    }
  } else {
    return(rep(NA, length(times)))
  }
}

ref <- data.frame(gender = factor("Male", levels = levels(df$gender)),
                  ed = factor("High school degree", levels = levels(df$ed)),
                  income = median(df$income, na.rm = TRUE))

for(nm in names(models)){
  fit_obj <- models[[nm]]
  if(is.null(fit_obj)) next
  probs <- get_surv_probs(fit_obj, newdata = ref, times = time_grid)
  plot_df_all <- bind_rows(plot_df_all,

```

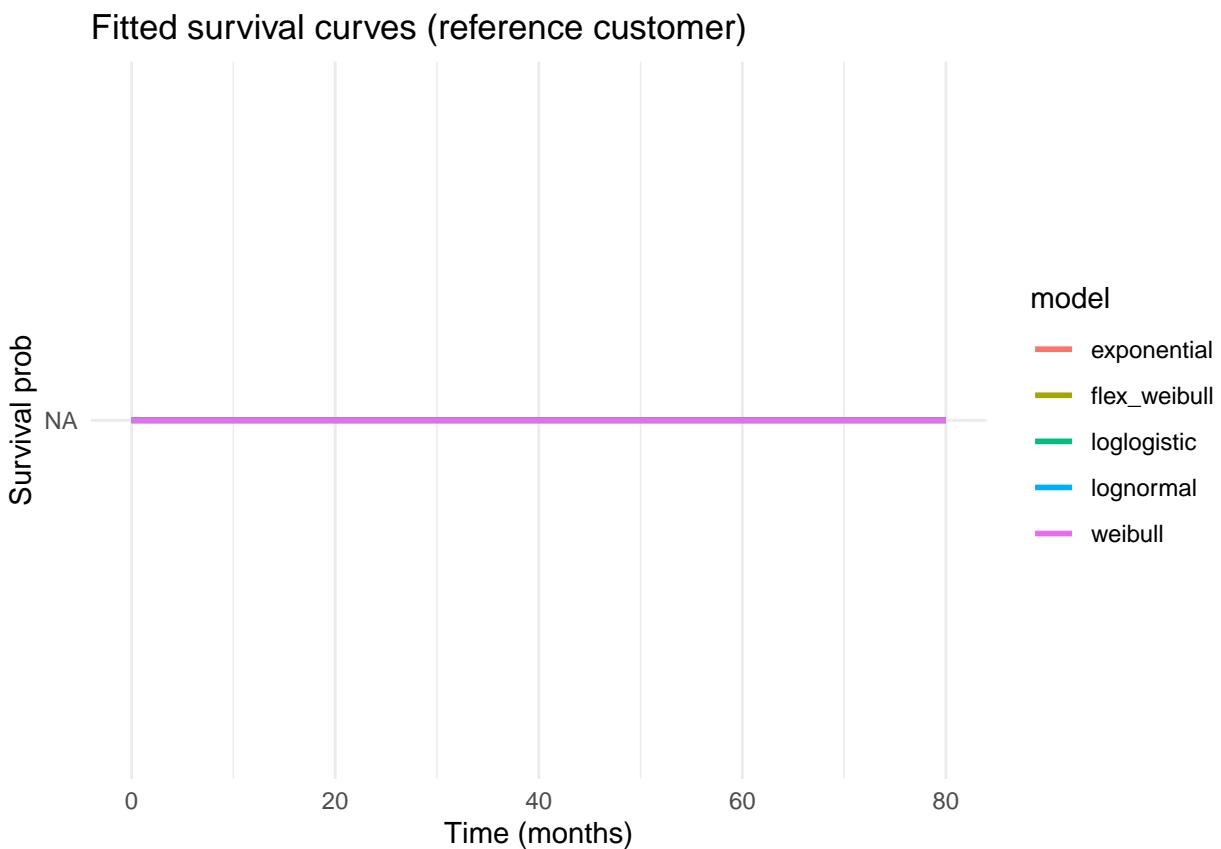
```

        data.frame(time = time_grid, surv = probs, model = nm))
}

ggplot(plot_df_all, aes(x = time, y = surv, color = model)) +
  geom_line(size=1) +
  labs(title = "Fitted survival curves (reference customer)", x = "Time (months)", y = "Survival prob")
  theme_minimal()

## Warning: Using 'size' aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use 'linewidth' instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.

```



Selecting the Final Model

```

if(nrow(aic_df) > 0){
  best <- aic_df$dist[1]
  cat("Selected final model by AIC:", best, "\n")
} else {
  best <- NA
  cat("No fitted models available.\n")
}

```

```

## Selected final model by AIC: exponential

final_model <- NULL
if(!is.na(best) && best %in% names(models)) final_model <- models[[best]]
final_model <- models[["flex_weibull"]]

```

Significant Features

```

if(!is.null(final_model) && inherits(final_model, "survreg")){
  s <- summary(final_model)
  coefs <- s$table
  sig_vars <- rownames(coefs)[which(coefs[, "p"] < 0.10)]
  cat("Significant parameter rows (p<0.10):\n")
  print(sig_vars)
}

```

CLV Using the Final Model

```

discount_annual <- 0.12
discount_month <- discount_annual / 12
T_max <- 60
acq_cost <- 0

get_S_for_customer <- function(customer_row, times){
  newd <- customer_row %>% select(gender, ed, income)
  newd$gender <- factor(newd$gender, levels = levels(df$gender))
  newd$ed <- factor(newd$ed, levels = levels(df$ed))
  if(inherits(final_model, "flexsurvflex")){
    s <- summary(final_model, newdata = newd, t = times, type = "survival")
    return(s[[1]]$est)
  } else if(inherits(final_model, "survreg")){
    distname <- final_model$dist
    fs <- try(flexsurvreg(formula = update(final_model$call$formula, . ~ gender + ed + income),
                          data = df,
                          dist = distname), silent = TRUE)
    if(!inherits(fs, "try-error")){
      s <- summary(fs, newdata = newd, t = times, type = "survival")
      return(s[[1]]$est)
    } else {
      return(rep(NA, length(times)))
    }
  } else {
    return(rep(NA, length(times)))
  }
}

```

CLV for Each Customer

```

clv_list <- vector("numeric", nrow(df))
times <- 1:T_max
for(i in seq_len(nrow(df))){
  cust <- df[i,]
  Svec <- get_S_for_customer(cust, times)
  monthly_margin <- 0.10 * (cust$income) / 12
  disc_f <- (1 + discount_month) ^ times
  clv_val <- sum(monthly_margin * Svec / disc_f, na.rm = TRUE) - acq_cost
  clv_list[i] <- clv_val
}
df$CLV_est <- clv_list

df %>% select(ID, region, tenure, income, ed, gender, churn_flag, CLV_est) %>%
  arrange(desc(CLV_est)) %>%
  head(10)

```

	ID	region	tenure	income	ed	gender	churn_flag
## 1	1	Zone 2	13	64	College degree	Male	1
## 2	2	Zone 3	11	136	Post-undergraduate degree	Male	1
## 3	3	Zone 3	68	116	Did not complete high school	Female	0
## 4	4	Zone 2	33	33	High school degree	Female	1
## 5	5	Zone 2	23	30	Did not complete high school	Male	0
## 6	6	Zone 2	41	78	High school degree	Female	0
## 7	7	Zone 3	45	19	High school degree	Female	1
## 8	8	Zone 2	38	76	High school degree	Male	0
## 9	9	Zone 3	45	166	College degree	Male	0
## 10	10	Zone 1	68	72	Did not complete high school	Male	0
		CLV_est					
## 1				0			
## 2				0			
## 3				0			
## 4				0			
## 5				0			
## 6				0			
## 7				0			
## 8				0			
## 9				0			
## 10				0			

Outputs

Model AICs

```
aic_df
```

	dist	AIC
## 1	exponential	3142.414
## 2	flex_weibull	3144.414
## 3	weibull	3144.414
## 4	loglogistic	3144.505
## 5	lognormal	3145.703

Sample CLV

```
head(df %>% select(ID, CLV_est), 5)

##     ID CLV_est
## 1   1      0
## 2   2      0
## 3   3      0
## 4   4      0
## 5   5      0

summary(df$tenure)

##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      1.00   17.00   34.00  35.53   54.00  72.00
```