

# Survival Analysis Exam Arezki BALI:

2025-12-21

```
library(tidyverse)

## Warning: package 'tidyverse' was built under R version 4.5.2

## Warning: package 'readr' was built under R version 4.5.2

## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr     1.1.4     v readr     2.1.6
## v forcats   1.0.1     v stringr   1.6.0
## v ggplot2   4.0.0     v tibble    3.3.0
## v lubridate 1.9.4     v tidyverse 1.3.2
## v purrr    1.1.0
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()   masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(survival)
library(broom)
```

```
## Warning: package 'broom' was built under R version 4.5.2
```

```
library(survminer) # For Kaplan-Meier plots
```

```
## Loading required package: ggpubr
##
## Attaching package: 'survminer'
##
## The following object is masked from 'package:survival':
##
##     myeloma
```

```
library(survivalROC) # For Time-Dependent ROC
```

```
# Data Loading
dat <- read.csv("/Users/arezkibali/Downloads/WA_Fn-UseC_-Telco-Customer-Churn.csv")
```

```
# Convert 'TotalCharges' to numeric (this introduces NAs for blank strings)
dat$TotalCharges <- as.numeric(dat$TotalCharges)
```

```

# Handle Missing Values (The "Emicvaz" improvement: Impute 0 instead of dropping)
# Logic: If tenure is 0, TotalCharges should be 0.
dat$TotalCharges[is.na(dat$TotalCharges)] <- 0

# Convert Target 'Churn' to Binary (1 = Yes, 0 = No)
dat$Churn <- ifelse(dat$Churn == "Yes", 1, 0)

# Convert Categorical Predictors to Factors
dat$Contract <- as.factor(dat$Contract)
dat$PaymentMethod <- as.factor(dat$PaymentMethod)
dat$InternetService <- as.factor(dat$InternetService)
dat$gender <- as.factor(dat$gender)

# Kaplan-Meier Plots)

# 1. Fit the Kaplan-Meier overall
km_fit_contract <- survfit(Surv(tenure, Churn) ~ 1, data = dat)

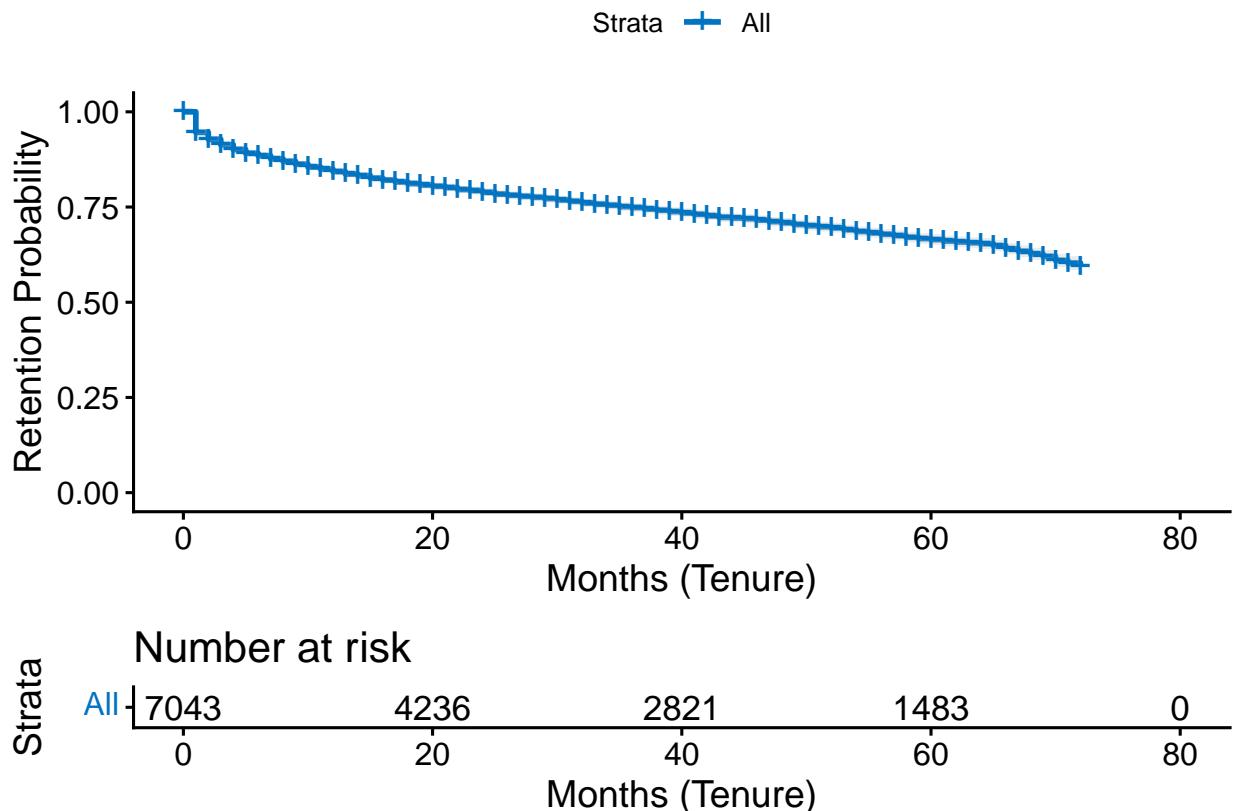
# 2. Plot
ggsurvplot(km_fit_contract,
            data = dat,
            pval = TRUE,                      # Show p-value
            risk.table = TRUE,                 # Show table of people at risk
            conf.int = TRUE,                  # Show confidence intervals
            palette = "jco",                  # Nice color palette
            main = "Kaplan-Meier: Actual Retention by Contract Type",
            xlab = "Months (Tenure)",         # X-axis label
            ylab = "Retention Probability")  # Y-axis label

## Warning: Using 'size' aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use 'linewidth' instead.
## i The deprecated feature was likely used in the ggpibr package.
## Please report the issue at <https://github.com/kassambara/ggpibr/issues>.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.

## Warning in .pvalue(fit, data = data, method = method, pval = pval, pval.coord = pval.coord, : There
## This is a null model.

## Ignoring unknown labels:
## * fill : "Strata"
## Ignoring unknown labels:
## * fill : "Strata"
## Ignoring unknown labels:
## * fill : "Strata"
## Ignoring unknown labels:
## * fill : "Strata"
## Ignoring unknown labels:
## * fill : "Strata"
## Ignoring unknown labels:
## * colour : "Strata"

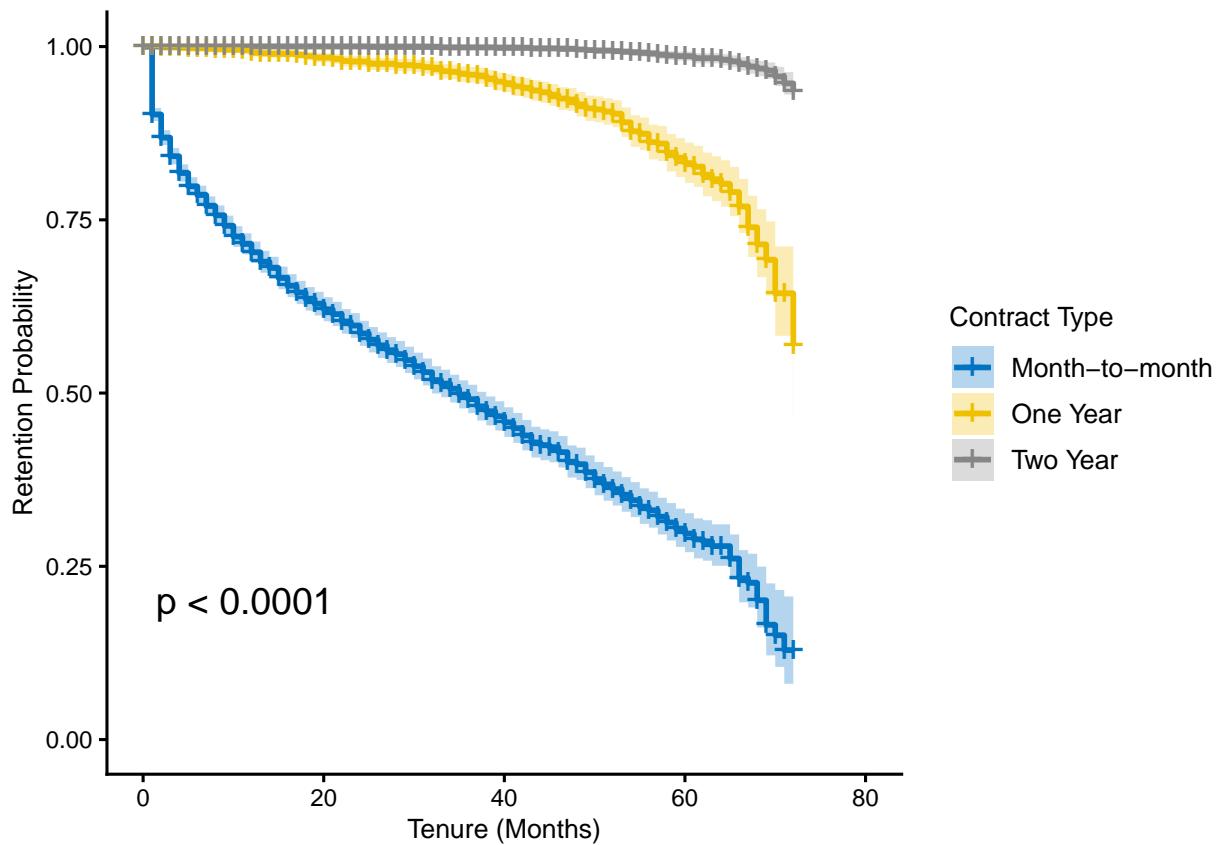
```



```
# Kaplan-Meier by contract
km_fit_contract <- survfit(Surv(tenure, Churn) ~ Contract, data = dat)

ggsurvplot(km_fit_contract,
            data = dat,
            pval = TRUE,
            conf.int = TRUE,
            palette = "jco",

            # --- VISUAL FIXES ---
            legend = "right",
            legend.title = "Contract Type",
            legend.labs = c("Month-to-month", "One Year", "Two Year"),
            xlab = "Tenure (Months)",
            ylab = "Retention Probability",
            font.x = 10,
            font.y = 10,
            font.tickslab = 9
        )
```



```

# Cox: MODEL BUILDING

# 1. Define the Full Model WITHOUT 'TotalCharges'
# Total Charges is mainly Monthly Charge* Tenure...sort of colinearity that will effect the model.

Mfull <- coxph(Surv(tenure, Churn) ~ MonthlyCharges + Contract +
                 InternetService + PaymentMethod + PaperlessBilling +
                 gender + Partner + Dependents,
                 data = dat)

# 2. Run Standard Stepwise Selection
MAIC <- step(Mfull, direction = "both", trace = 0)

print("Best Model Selected")

## [1] "Best Model Selected"

summary(MAIC)

## Call:
## coxph(formula = Surv(tenure, Churn) ~ MonthlyCharges + Contract +
##        InternetService + PaymentMethod + PaperlessBilling + gender +
##        Partner, data = dat)
##
```

```

##   n= 7043, number of events= 1869
##
##                               coef  exp(coef)    se(coef)      z
## MonthlyCharges          -0.030634  0.969830  0.002138 -14.327
## ContractOne year         -1.715405  0.179891  0.087037 -19.709
## ContractTwo year         -3.415544  0.032859  0.162976 -20.957
## InternetServiceFiber optic  1.467957  4.340359  0.100783  14.566
## InternetServiceNo        -1.232081  0.291685  0.124228 -9.918
## PaymentMethodCredit card (automatic) -0.073819  0.928840  0.090607 -0.815
## PaymentMethodElectronic check     0.666345  1.947107  0.070686  9.427
## PaymentMethodMailed check       0.639425  1.895391  0.088069  7.260
## PaperlessBillingYes          0.175693  1.192073  0.056092  3.132
## genderMale                  -0.075324  0.927443  0.046401 -1.623
## PartnerYes                  -0.575206  0.562589  0.050268 -11.443
##                               Pr(>|z|)
## MonthlyCharges            < 2e-16 ***
## ContractOne year           < 2e-16 ***
## ContractTwo year           < 2e-16 ***
## InternetServiceFiber optic < 2e-16 ***
## InternetServiceNo          < 2e-16 ***
## PaymentMethodCredit card (automatic) 0.41523
## PaymentMethodElectronic check     < 2e-16 ***
## PaymentMethodMailed check       3.86e-13 ***
## PaperlessBillingYes          0.00173 **
## genderMale                  0.10452
## PartnerYes                  < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##                               exp(coef)  exp(-coef) lower .95 upper .95
## MonthlyCharges          0.96983   1.0311   0.96577  0.97390
## ContractOne year         0.17989   5.5589   0.15168  0.21335
## ContractTwo year         0.03286  30.4335   0.02387  0.04522
## InternetServiceFiber optic  4.34036   0.2304   3.56237  5.28825
## InternetServiceNo        0.29168   3.4284   0.22865  0.37210
## PaymentMethodCredit card (automatic) 0.92884   1.0766   0.77771  1.10934
## PaymentMethodElectronic check     1.94711   0.5136   1.69520  2.23645
## PaymentMethodMailed check       1.89539   0.5276   1.59490  2.25249
## PaperlessBillingYes          1.19207   0.8389   1.06797  1.33060
## genderMale                  0.92744   1.0782   0.84682  1.01574
## PartnerYes                  0.56259   1.7775   0.50980  0.62084
##
## Concordance= 0.852  (se = 0.004 )
## Likelihood ratio test= 3245 on 11 df,  p=<2e-16
## Wald test                 = 1813 on 11 df,  p=<2e-16
## Score (logrank) test = 2929 on 11 df,  p=<2e-16

# VISUALIZATION: FOREST PLOT (Hazard Ratios)

library(survminer)

# Create the Forest Plot
p <- ggforest(MAIC,
  data = dat,

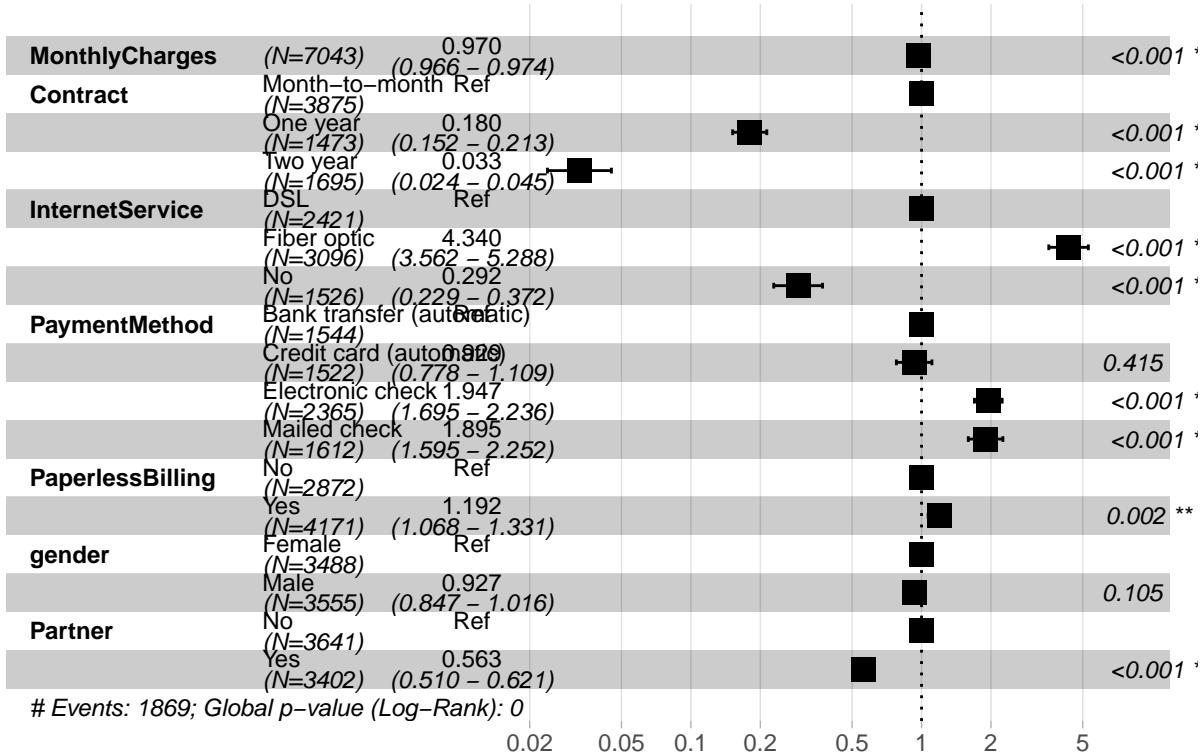
```

```

    main = "Hazard Ratios: Predictors of Churn",
    fontsize = 0.8,
    refLabel = "Ref",
    cpositions = c(0.02, 0.22, 0.4)
)
p

```

## Hazard Ratios: Predictors of Churn



```

# Save with specific dimensions (Width x Height)
ggsave("ForestPlot_Tall.png", plot = p, width = 10, height = 12)

```

#VALIDATION: TRAIN/TEST SPLIT

```

set.seed(123) # For reproducibility
train_index <- sample(1:nrow(dat), size = 0.5 * nrow(dat)) # 50% split
d_train <- dat[train_index, ]
d_test <- dat[-train_index, ]

# Fit model on TRAINING data only
M_train <- coxph(Surv(tenure, Churn) ~ MonthlyCharges + Contract +
                    InternetService + PaymentMethod + PaperlessBilling,
                    data = d_train)

# Predict risk scores on TESTING data
d_test$risk_score <- predict(M_train, newdata = d_test, type = "lp")

```

```

# Calculate C-Statistic (Concordance) on Test Data
# 0.5 = Random, 1.0 = Perfect
perf_test <- coxph(Surv(tenure, Churn) ~ risk_score, data = d_test)
c_index <- summary(perf_test)$concordance[1]

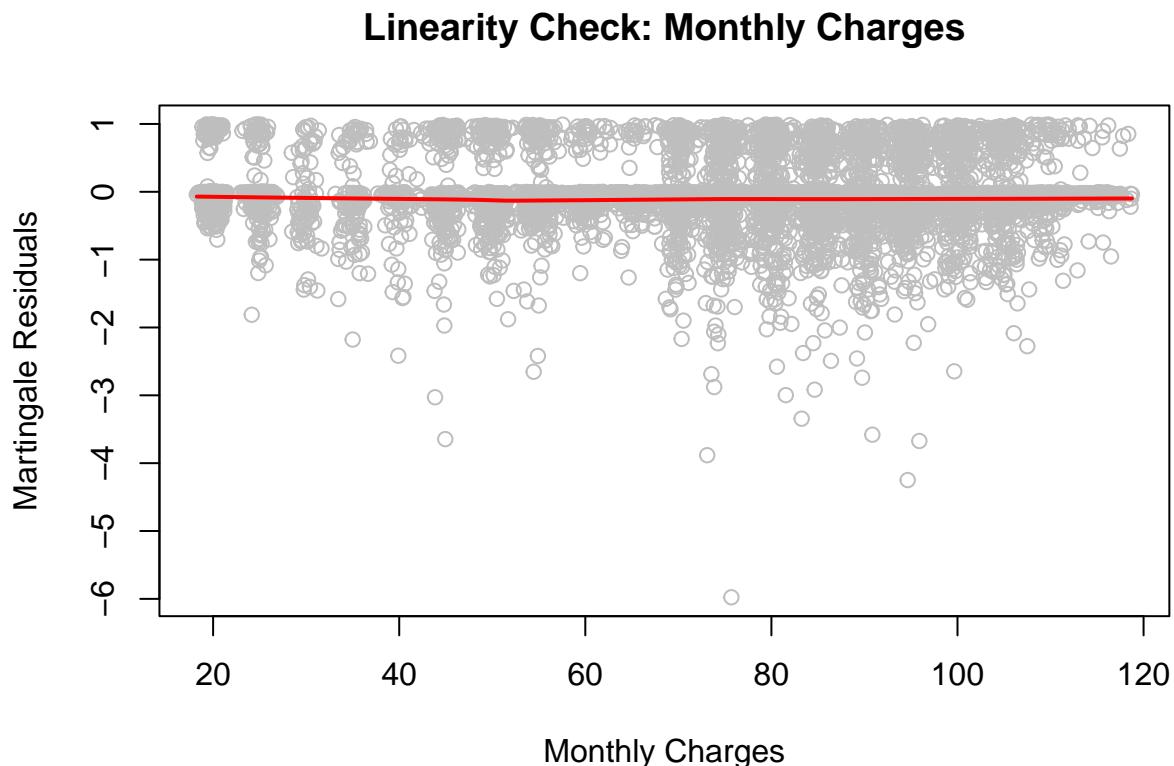
print(paste("C-Statistic (Test Set):", round(c_index, 3)))

## [1] "C-Statistic (Test Set): 0.847"

# DIAGNOSTICS: CHECKING ASSUMPTIONS

# A. Martingale Residuals (Check Linearity of MonthlyCharges)
dat$residuals <- residuals(MAIC, type = "martingale")
plot(dat$MonthlyCharges, dat$residuals, col = "gray",
     main = "Linearity Check: Monthly Charges",
     xlab = "Monthly Charges", ylab = "Martingale Residuals")
lines(lowess(dat$MonthlyCharges, dat$residuals), col = "red", lwd = 2)

```



```

# B. Proportional Hazards Test (Schoenfeld Residuals)
# Checks if risk ratios are constant over time
test_ph <- cox.zph(MAIC)
print(" Proportional Hazards Test ")

## [1] " Proportional Hazards Test "

```

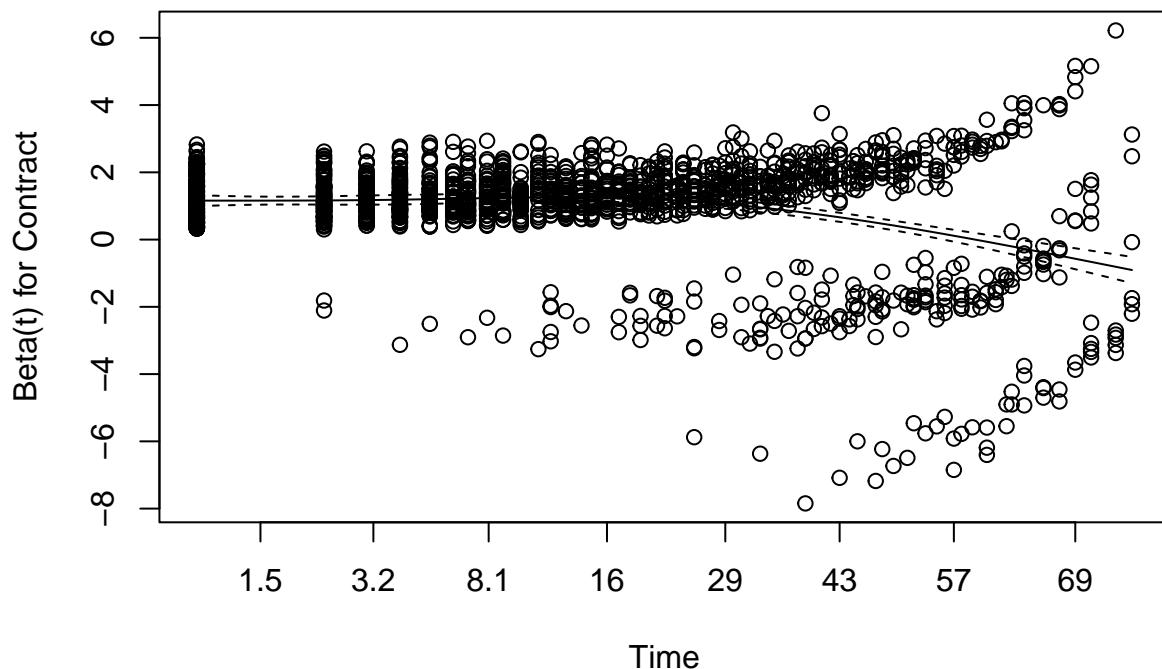
```

print(test_ph)

##          chisq df      p
## MonthlyCharges 105.393 1 < 2e-16
## Contract       89.719  2 < 2e-16
## InternetService 41.496  2 9.8e-10
## PaymentMethod   48.453  3 1.7e-10
## PaperlessBilling  3.732  1  0.053
## gender          0.012  1  0.913
## Partner         20.289  1 6.7e-06
## GLOBAL          253.807 11 < 2e-16

# Visual check for 'Contract' (Likely violator)
plot(test_ph, var = "Contract")

```



```

library(survivalROC)
library(dplyr)
library(purrr)
library(tidyr)
library(ggplot2)

times_to_check <- c(12, 24, 36)

# If Churn and tenure are already correct, keep it simple:

```

```

d_test <- dat %>%
  mutate(lp = predict(MAIC, newdata = ., type = "lp"))

roc_df <- map_df(times_to_check, function(t) {
  roc <- survivalROC(
    Stime      = d_test$tenure,
    status     = d_test$Churn,
    marker     = d_test$lp,
    predict.time = t,
    method     = "NNE",
    span       = 0.25 * nrow(d_test)^(-0.20)
  )

  tibble(
    t   = t,
    auc = roc$AUC,
    FP  = roc$FP,
    TP  = roc$TP
  )
})

ggplot(roc_df, aes(FP, TP)) +
  geom_line() +
  geom_abline(linetype = "dashed") +
  geom_label(
    data = roc_df %>% distinct(t, auc),
    aes(x = 0.6, y = 0.2, label = sprintf("AUC = %.3f", auc)),
    inherit.aes = FALSE
  ) +
  facet_wrap(~t, labeller = labeller(t = \((x)\) paste0(x, " months")))) +
  theme_bw()

```

