

harinris_Homework6

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Lab 11.8

11.8.1-Brain Cancer Data

```
library(ISLR2)

names(BrainCancer)

## [1] "sex"      "diagnosis" "loc"      "ki"      "gtv"      "stereo"
## [7] "status"   "time"
```

```
attach(BrainCancer)
table(sex)
```

```
## sex
## Female  Male
##      45    43
```

```
table(diagnosis)
```

```
## diagnosis
## Meningioma  LG glioma  HG glioma    Other
##           42         9        22        14
```

```
table(status)
```

```
## status
##  0  1
## 53 35
```

```
length(BrainCancer$time)
```

```
## [1] 88
```

```
length(BrainCancer$status)
```

```
## [1] 88
```

```
class(BrainCancer$time)
```

```
## [1] "numeric"
```

```
class(BrainCancer$status)
```

```
## [1] "integer"
```

```
str(BrainCancer)
```

```
## 'data.frame': 88 obs. of 8 variables:
## $ sex : Factor w/ 2 levels "Female","Male": 1 2 1 1 2 1 2 2 1 2 ...
## $ diagnosis: Factor w/ 4 levels "Meningioma","LG glioma",...: 1 3 1 2 3 1 1 2 1 3
## ...
## $ loc : Factor w/ 2 levels "Infratentorial",...: 1 2 1 2 2 2 2 2 2 2 ...
## $ ki : int 90 90 70 80 90 80 80 80 70 100 ...
## $ gtv : num 6.11 19.35 7.95 7.61 5.06 ...
## $ stereo : Factor w/ 2 levels "SRS","SRT": 1 2 1 2 2 1 2 2 2 2 ...
## $ status : int 0 1 0 1 1 0 0 0 0 0 ...
## $ time : num 57.64 8.98 26.46 47.8 6.3 ...
```

```
head(BrainCancer)
```

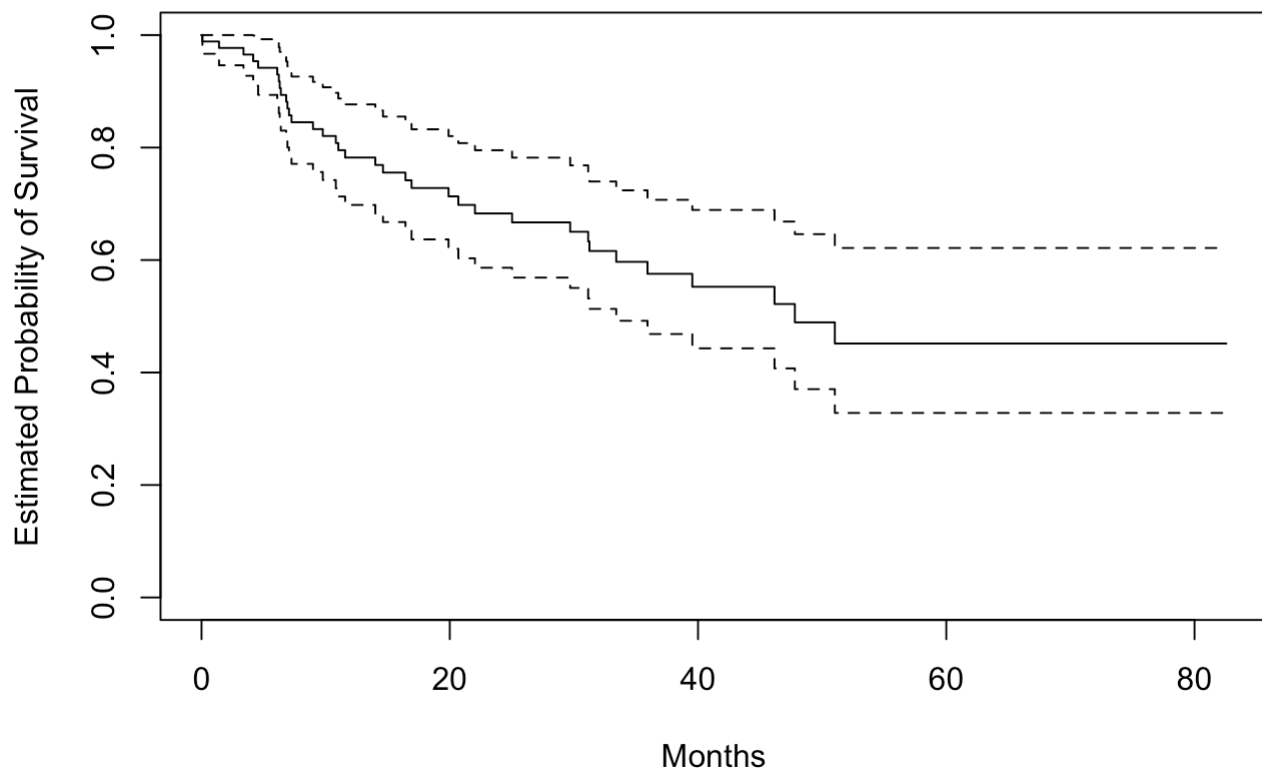
```
##      sex diagnosis      loc ki   gtv stereo status  time
## 1 Female Meningioma Infratentorial 90 6.11    SRS      0 57.64
## 2  Male  HG glioma Supratentorial 90 19.35    SRT      1 8.98
## 3 Female Meningioma Infratentorial 70 7.95    SRS      0 26.46
## 4 Female  LG glioma Supratentorial 80 7.61    SRT      1 47.80
## 5  Male  HG glioma Supratentorial 90 5.06    SRT      1 6.30
## 6 Female Meningioma Supratentorial 80 4.82    SRS      0 52.75
```

```
library(survival)
```

```
surv_obj <- Surv(BrainCancer$time, BrainCancer$status)
```

```
fit.surv <- survfit(surv_obj ~ 1)
```

```
plot(fit.surv, xlab = "Months",
     ylab = "Estimated Probability of Survival")
```

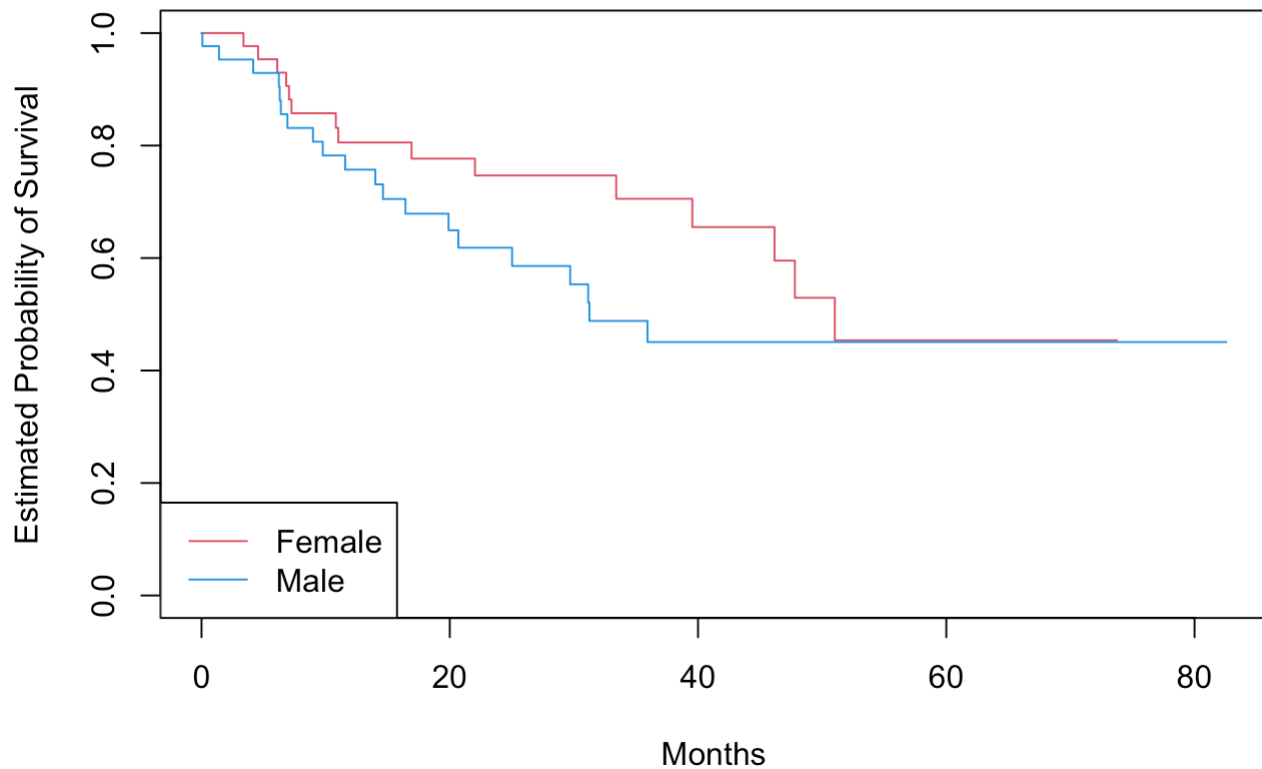


```
BrainCancer$sex <- factor(BrainCancer$sex)

fit.sex <- survfit(Surv(time, status) ~ sex, data = BrainCancer)

plot(fit.sex, xlab = "Months", ylab = "Estimated Probability of Survival", col = c(2, 4))

legend("bottomleft", legend = levels(BrainCancer$sex), col = c(2, 4), lty = 1)
```



```
logrank.test <- survdiff(Surv(time, status) ~ sex, data = BrainCancer)
logrank.test
```

```
## Call:
## survdiff(formula = Surv(time, status) ~ sex, data = BrainCancer)
##
##              N Observed Expected (O-E)^2/E (O-E)^2/V
## sex=Female  45      15      18.5    0.676    1.44
## sex=Male   43      20      16.5    0.761    1.44
##
##  Chisq= 1.4  on 1 degrees of freedom, p= 0.2
```

```
fit.cox <- coxph(Surv(time, status) ~ sex, data = BrainCancer)
summary(fit.cox)
```

```
## Call:
## coxph(formula = Surv(time, status) ~ sex, data = BrainCancer)
##
##      n= 88, number of events= 35
##
##              coef exp(coef) se(coef)      z Pr(>|z|)
## sexMale 0.4077      1.5033   0.3420 1.192   0.233
##
##              exp(coef) exp(-coef) lower .95 upper .95
## sexMale      1.503      0.6652   0.769      2.939
##
## Concordance= 0.565 (se = 0.045 )
## Likelihood ratio test= 1.44 on 1 df,  p=0.2
## Wald test              = 1.42 on 1 df,  p=0.2
## Score (logrank) test = 1.44 on 1 df,  p=0.2
```

```
summary(fit.cox)$logtest[1]
```

```
##      test
## 1.438822
```

```
summary(fit.cox)$waldtest[1]
```

```
## test
## 1.42
```

```
summary(fit.cox)$sctest[1]
```

```
##      test
## 1.440495
```

```
logrank.test$chisq
```

```
## [1] 1.440495
```

```
fit.all <- coxph(Surv(time, status) ~ sex + diagnosis + loc + ki + gtv + stereo, data = BrainCancer)

fit.all
```

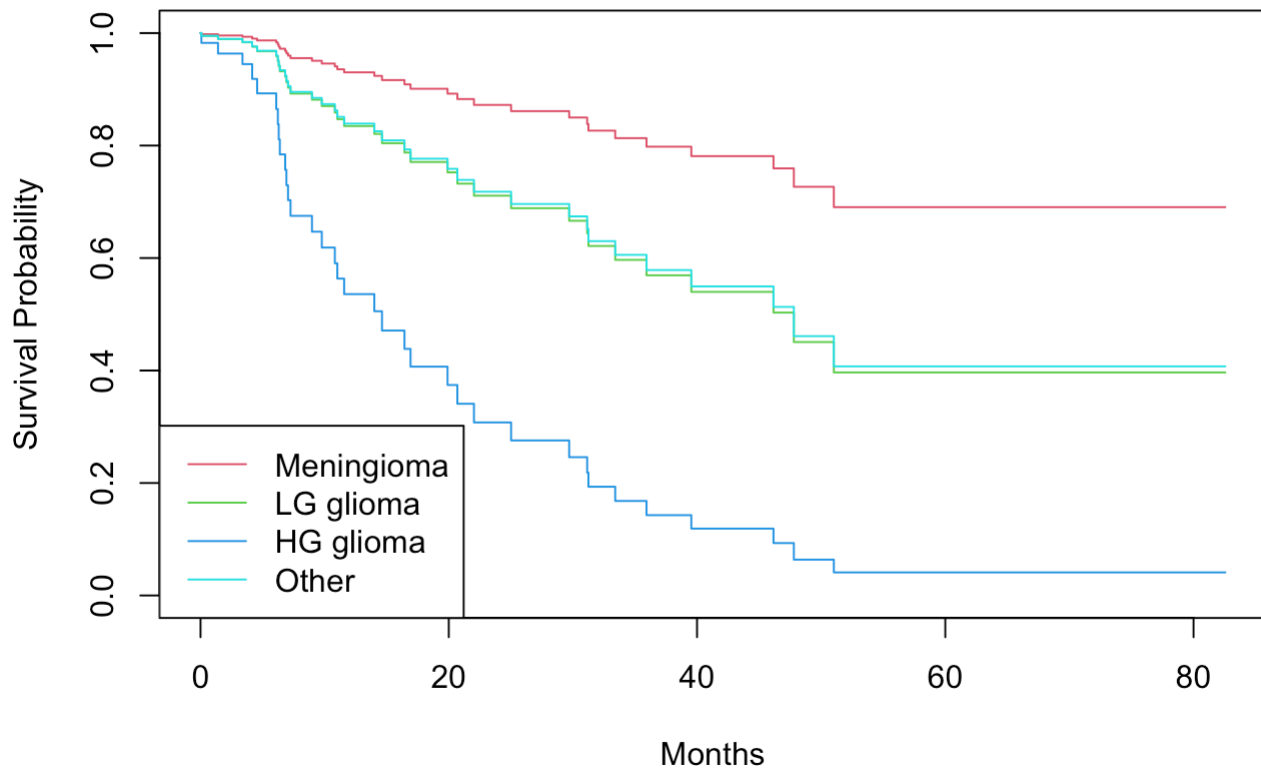
```
## Call:
## coxph(formula = Surv(time, status) ~ sex + diagnosis + loc +
##       ki + gtv + stereo, data = BrainCancer)
##
```

	coef	exp(coef)	se(coef)	z	p
## sexMale	0.18375	1.20171	0.36036	0.510	0.61012
## diagnosisLG glioma	0.91502	2.49683	0.63816	1.434	0.15161
## diagnosisHG glioma	2.15457	8.62414	0.45052	4.782	1.73e-06
## diagnosisOther	0.88570	2.42467	0.65787	1.346	0.17821
## locSupratentorial	0.44119	1.55456	0.70367	0.627	0.53066
## ki	-0.05496	0.94653	0.01831	-3.001	0.00269
## gtv	0.03429	1.03489	0.02233	1.536	0.12466
## stereoSRT	0.17778	1.19456	0.60158	0.296	0.76760

```
##
## Likelihood ratio test=41.37 on 8 df, p=1.776e-06
## n= 87, number of events= 35
## (1 observation deleted due to missingness)
```

```
modaldata <- data.frame(
  diagnosis = levels(diagnosis),
  sex = rep("Female", 4),
  loc = rep("Supratentorial", 4),
  ki = rep(mean(ki), 4),
  gtv = rep(mean(gtv), 4),
  stereo = rep("SRT", 4)
)
```

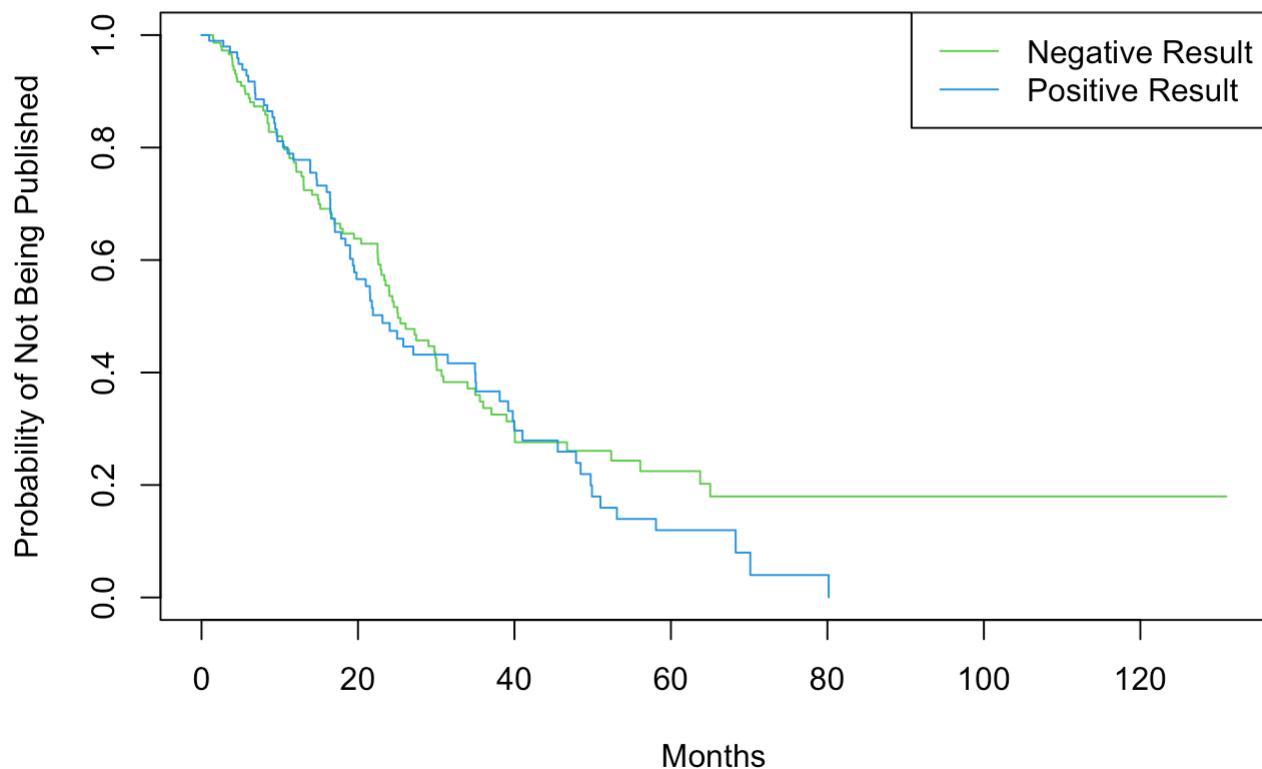
```
survplots <- survfit(fit.all, newdata = modaldata)
plot(survplots, xlab = "Months",
     ylab = "Survival Probability", col = 2:5)
legend("bottomleft", levels(diagnosis), col = 2:5, lty = 1)
```



11.8.2 - Publication Data

```
library(survival)
library(ISLR2)
fit.posres <- survfit(Surv(time, status) ~ posres, data = Publication)
```

```
plot(fit.posres, xlab = "Months",
     ylab = "Probability of Not Being Published", col = 3:4)
legend("topright", c("Negative Result", "Positive Result"),
     col = 3:4, lty = 1)
```



```
fit.pub <- coxph(Surv(time, status) ~ posres, data = Publication)
summary(fit.pub)
```

```
## Call:
## coxph(formula = Surv(time, status) ~ posres, data = Publication)
##
##    n= 244, number of events= 156
##
##              coef exp(coef) se(coef)      z Pr(>|z|)
## posres  0.1481    1.1596   0.1616  0.916    0.36
##
##      exp(coef) exp(-coef) lower .95 upper .95
## posres      1.16    0.8624   0.8448    1.592
##
## Concordance= 0.503  (se = 0.022 )
## Likelihood ratio test= 0.83  on 1 df,   p=0.4
## Wald test               = 0.84  on 1 df,   p=0.4
## Score (logrank) test = 0.84  on 1 df,   p=0.4
```

```
logrank.test <- survdiff(Surv(time, status) ~ posres, data = Publication)
logrank.test
```



```
## Call:
## survdiff(formula = Surv(time, status) ~ posres, data = Publication)
##
##              N Observed Expected (O-E)^2/E (O-E)^2/V
## posres=0 146      87      92.6      0.341      0.844
## posres=1  98      69      63.4      0.498      0.844
##
##  Chisq= 0.8  on 1 degrees of freedom, p= 0.4
```

```
fit.pub2 <- coxph(Surv(time, status) ~ . - mech, data = Publication)
summary(fit.pub2)
```

```
## Call:
## coxph(formula = Surv(time, status) ~ . - mech, data = Publication)
##
##    n= 244, number of events= 156
##
##              coef exp(coef) se(coef)      z Pr(>|z|)
## posres    5.708e-01  1.770e+00  1.760e-01  3.244  0.00118 **
## multi    -4.086e-02  9.600e-01  2.512e-01 -0.163  0.87079
## clinend   5.462e-01  1.727e+00  2.620e-01  2.085  0.03710 *
## sampsize  4.678e-06  1.000e+00  1.472e-05  0.318  0.75070
## budget    4.385e-03  1.004e+00  2.465e-03  1.779  0.07518 .
## impact    5.832e-02  1.060e+00  6.676e-03  8.735  < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##              exp(coef) exp(-coef) lower .95 upper .95
## posres          1.770      0.5651    1.2534    2.498
## multi           0.960      1.0417    0.5867    1.571
## clinend         1.727      0.5792    1.0332    2.885
## sampsize        1.000      1.0000    1.0000    1.000
## budget          1.004      0.9956    0.9996    1.009
## impact          1.060      0.9433    1.0463    1.074
##
## Concordance= 0.801 (se = 0.016 )
## Likelihood ratio test= 149.2 on 6 df,  p=<2e-16
## Wald test              = 159.3 on 6 df,  p=<2e-16
## Score (logrank) test = 233.7 on 6 df,  p=<2e-16
```

11.8.3 - Call Center Data

```
set.seed(4)

N <- 2000

Operators <- sample(5:15, N, replace = TRUE)

Center <- sample(c("A", "B", "C"), N, replace = TRUE)

Time <- sample(c("Morn.", "After.", "Even."), N, replace = TRUE)

X <- model.matrix(~ Operators + Center + Time)[, -1]
```

```
X[1:5, ]
```

```
##   Operators CenterB CenterC TimeEven. TimeMorn.
## 1         12        1        0         0         1
## 2         15        0        0         0         0
## 3          7        0        1         1         0
## 4          7        0        0         0         0
## 5         11        0        1         0         1
```

```
true.beta <- c(0.04, -0.3, 0, 0.2, -0.2)
h.fn <- function(x) return(0.00001 * x)
```

```
library(rms)
```

```
## Warning: package 'rms' was built under R version 4.3.2
```

```
## Loading required package: Hmisc
```

```
## Warning: package 'Hmisc' was built under R version 4.3.2
```

```
##
## Attaching package: 'Hmisc'
```

```
## The following objects are masked from 'package:base':
##
##   format.pval, units
```

```
library(coxed)
```

```
## Loading required package: mgcv
```

```
## Loading required package: nlme
```

```
## This is mgcv 1.9-1. For overview type 'help("mgcv-package")'.
```

```
queuing <- sim.survdata(N = N, T = 1000, X = X, beta = true.beta, hazard.fun = h.fn)
```

```
## Warning in FUN(X[[i]], ...): 9 additional observations right-censored because the
user-supplied hazard function
##                               is nonzero at the latest timepoint. To avoid the
se extra censored observations, increase T
```

```
names(queuing)
```

```
## [1] "data"          "xdata"          "baseline"       "xb"
## [5] "exp.xb"         "betas"          "ind.survive"    "marg.effect"
## [9] "marg.effect.data"
```

```
head(queuing$data)
```

```
##   Operators CenterB CenterC TimeEven. TimeMorn.   y failed
## 1         12        1        0         0         1 344  TRUE
## 2         15        0        0         0         0 241  TRUE
## 3          7        0        1         1         0 187  TRUE
## 4          7        0        0         0         0 279  TRUE
## 5         11        0        1         0         1 954  TRUE
## 6          7        1        0         0         1 455  TRUE
```

```
mean(queuing$data$failed)
```

```
## [1] 0.89
```

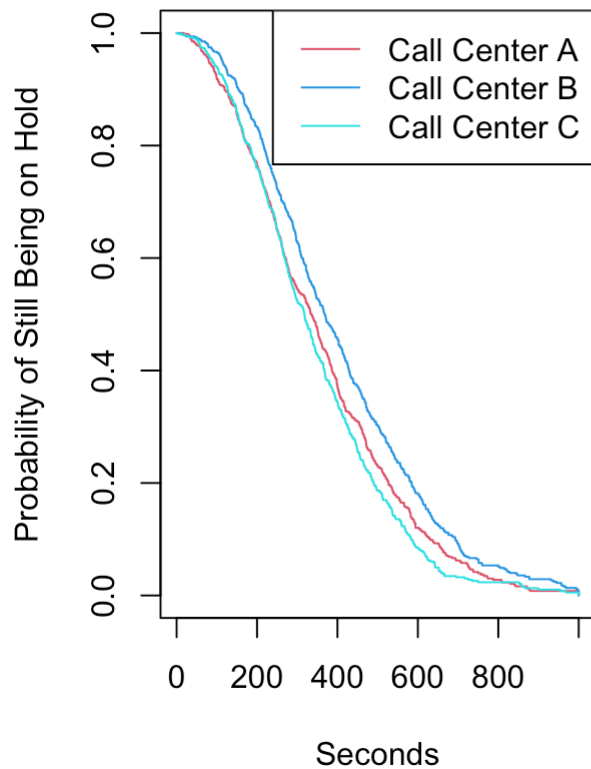
```
par(mfrow = c(1, 2))
```

```
fit.Center <- survfit(Surv(y, failed) ~ Center, data = queuing$data)
```

```
plot(fit.Center, xlab = "Seconds",
     ylab = "Probability of Still Being on Hold",
     col = c(2, 4, 5)) # Use different colors for each 'Center'
```

```
legend("topright",
     c("Call Center A", "Call Center B", "Call Center C"), # Legend labels
     col = c(2, 4, 5), # Colors corresponding to each 'Center'
     lty = 1) # Line type (solid)
```

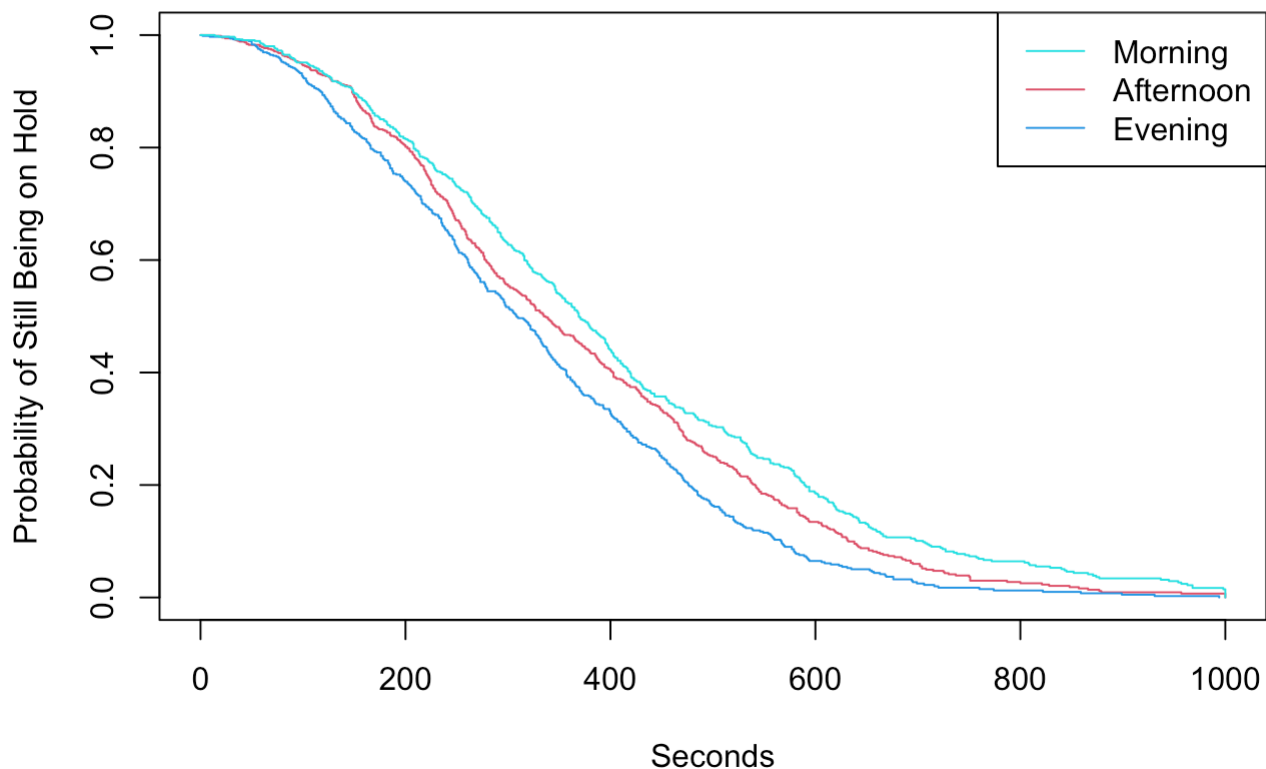
```
par(mfrow = c(1, 1))
```



```
fit.Time <- survfit(Surv(y, failed) ~ Time, data = queuing$data)

plot(fit.Time, xlab = "Seconds",
     ylab = "Probability of Still Being on Hold",
     col = c(2, 4, 5)) # Use different colors for each 'Time'

legend("topright",
     c("Morning", "Afternoon", "Evening"), # Legend labels
     col = c(5, 2, 4), # Colors corresponding to each 'Time'
     lty = 1) # Line type (solid)
```



```
survdif(Surv(y, failed) ~ Center, data = queuing$data)
```

```
## Call:
## survdif(formula = Surv(y, failed) ~ Center, data = queuing$data)
##
##           N Observed Expected (O-E)^2/E (O-E)^2/V
## Center=A 683      603      579    0.971    1.45
## Center=B 667      600      701   14.641   24.64
## Center=C 650      577      499   12.062   17.05
##
##  Chisq= 28.3  on 2 degrees of freedom, p= 7e-07
```

```
survdif(Surv(y, failed) ~ Time, data = queuing$data)
```

```
## Call:
## survdif(formula = Surv(y, failed) ~ Time, data = queuing$data)
##
##           N Observed Expected (O-E)^2/E (O-E)^2/V
## Time=After. 688      616      619    0.0135    0.021
## Time=Even. 653      582      468   27.6353   38.353
## Time=Morn. 659      582      693   17.7381   29.893
##
##  Chisq= 46.8  on 2 degrees of freedom, p= 7e-11
```

```
fit.queuing <- coxph(Surv(y, failed) ~ ., data = queuing$data)
```

```
fit.queuing
```

```
## Call:
```

```
## coxph(formula = Surv(y, failed) ~ ., data = queuing$data)
```

```
##
```

```
##
```

	coef	exp(coef)	se(coef)	z	p
## Operators	0.04174	1.04263	0.00759	5.500	3.8e-08
## CenterB	-0.21879	0.80349	0.05793	-3.777	0.000159
## CenterC	0.07930	1.08253	0.05850	1.356	0.175256
## TimeEven.	0.20904	1.23249	0.05820	3.592	0.000328
## TimeMorn.	-0.17352	0.84070	0.05811	-2.986	0.002828

```
##
```

```
## Likelihood ratio test=102.8 on 5 df, p=< 2.2e-16
```

```
## n= 2000, number of events= 1780
```

Applied Problem 10

10. a.

```
library(ISLR2)
```

```
x <- Surv(BrainCancer$time, BrainCancer$status)
```

```
plot(survfit(x ~ 1),
```

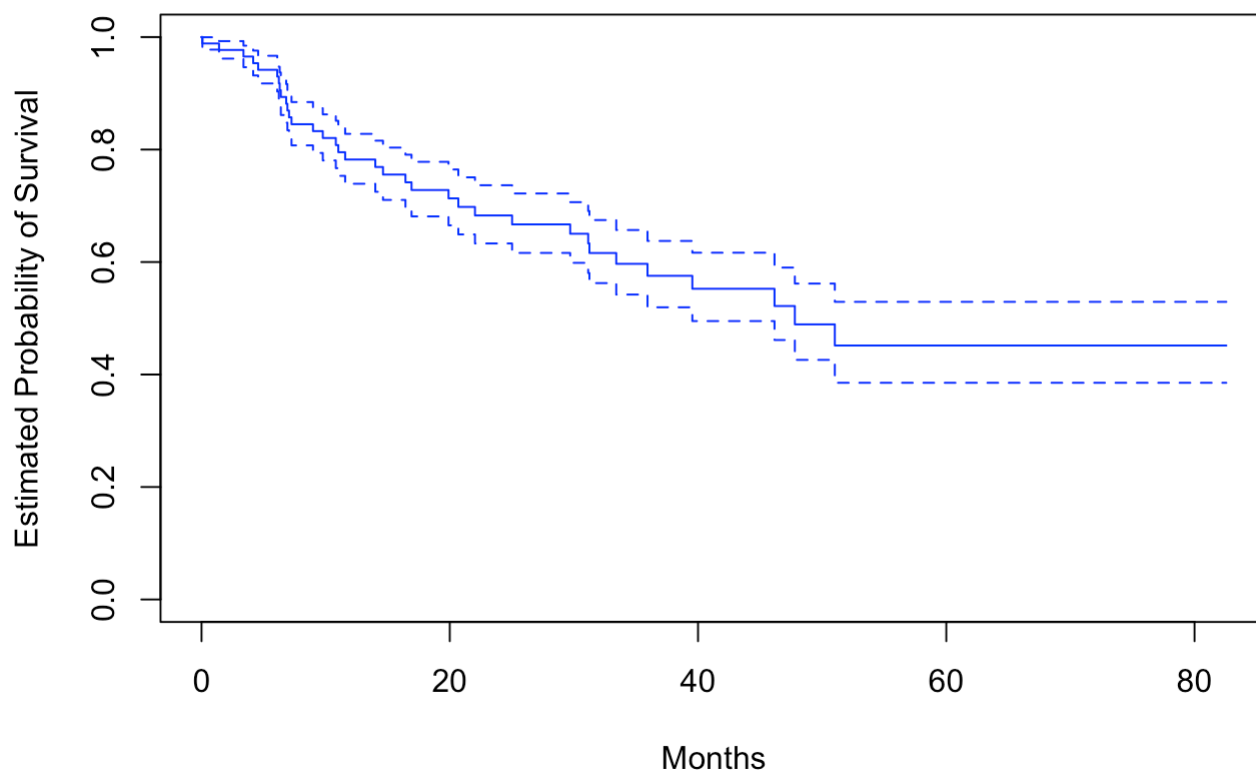
```
  xlab = "Months",
```

```
  ylab = "Estimated Probability of Survival",
```

```
  col = "blue",
```

```
  conf.int = 0.67
```

```
)
```



10. b.

```
library(survival)
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
```

```
## The following object is masked from 'package:nlme':
##
## collapse
```

```
## The following objects are masked from 'package:Hmisc':
##
## src, summarize
```

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

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

```
library(ggplot2)
```

```
## Warning: package 'ggplot2' was built under R version 4.3.2
```

```
set.seed(123) # for reproducibility
n <- 100
time <- rexp(n, rate = 0.1)
status <- rbinom(n, size = 1, prob = 0.7)
data <- data.frame(time = time, status = status)

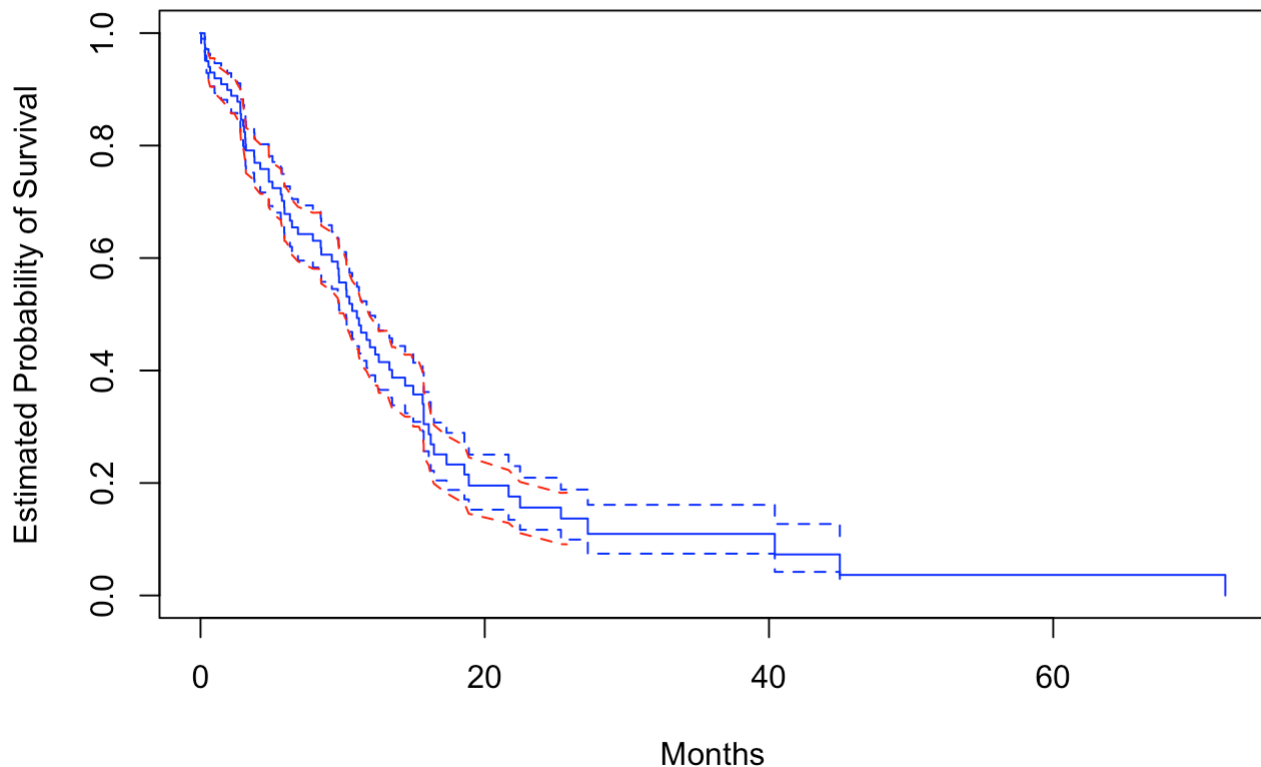
fit <- survfit(Surv(time, status) ~ 1)

n_bootstraps <- 200
bootstrap_results <- replicate(n_bootstraps, {
  boot_data <- sample_n(data, size = n, replace = TRUE)
  survfit(Surv(boot_data$time, boot_data$status) ~ 1)
}, simplify = FALSE)

bootstrap_surv_probs <- sapply(bootstrap_results, function(boot_result) {
  approx(boot_result$time, boot_result$surv, xout = fit$time)$y
})

se <- apply(bootstrap_surv_probs, 1, sd)

plot(fit, xlab = "Months", ylab = "Estimated Probability of Survival", col = "blue",
conf.int = 0.67)
lines(fit$time, fit$surv - se, lty = 2, col = "red")
lines(fit$time, fit$surv + se, lty = 2, col = "red")
```

10. c.

diagnosisHG and ki seem to be significant.

```
fit <- coxph(Surv(time, status) ~ sex + diagnosis + loc + ki + gtv + stereo, data = BrainCancer)
fit
```

```
## Call:
## coxph(formula = Surv(time, status) ~ sex + diagnosis + loc +
##       ki + gtv + stereo, data = BrainCancer)
##
##              coef exp(coef) se(coef)      z      p
## sexMale          0.18375    1.20171  0.36036  0.510 0.61012
## diagnosisLG glioma  0.91502    2.49683  0.63816  1.434 0.15161
## diagnosisHG glioma  2.15457    8.62414  0.45052  4.782 1.73e-06
## diagnosisOther      0.88570    2.42467  0.65787  1.346 0.17821
## locSupratentorial   0.44119    1.55456  0.70367  0.627 0.53066
## ki                 -0.05496    0.94653  0.01831 -3.001 0.00269
## gtv                 0.03429    1.03489  0.02233  1.536 0.12466
## stereoSRT           0.17778    1.19456  0.60158  0.296 0.76760
##
## Likelihood ratio test=41.37  on 8 df, p=1.776e-06
## n= 87, number of events= 35
## (1 observation deleted due to missingness)
```

10. d.

In order to account for additional predictors, we construct a model that includes those predictors. We then utilize this model to forecast new, synthesized data points. Here, we allow `ki` to assume every possible value, while maintaining the other predictors at either their mode or mean values.

```
library(ggfortify)
```

```
## Warning: package 'ggfortify' was built under R version 4.3.2
```

```
modeldata <- data.frame(
  sex = rep("Female", 5),
  diagnosis = rep("Meningioma", 5),
  loc = rep("Supratentorial", 5),
  ki = c(60, 70, 80, 90, 100),
  gtv = rep(mean(BrainCancer$gtv), 5),
  stereo = rep("SRT", 5)
)
survplots <- survfit(fit, newdata = modeldata)
plot(survplots, xlab = "Months", ylab = "Survival Probability", col = 2:6)
legend("bottomleft", c("60", "70", "80", "90", "100"), col = 2:6, lty = 1)
```

