Untitled

```
library(ISLR2)
library(survival)
library(coxed)
Loading required package: rms
Loading required package: Hmisc
Attaching package: 'Hmisc'
The following objects are masked from 'package:base':
    format.pval, units
Loading required package: mgcv
Loading required package: nlme
This is mgcv 1.9-1. For overview type 'help("mgcv-package")'.
\# labs \ \# 11.8.1 Brain Cancer Data
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)

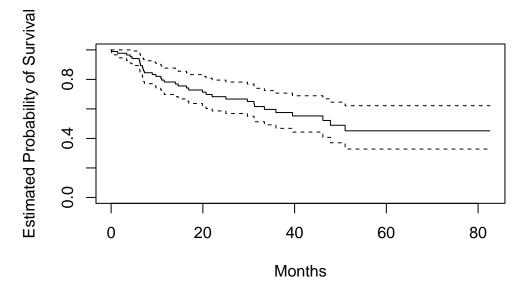
status

0 1

53 35

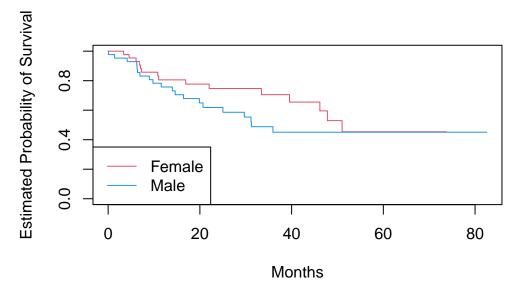
#create the Kaplan-Meier survival curve

```
fit.surv <- survfit(Surv(time, status) ~ 1)
plot(fit.surv, xlab = "Months",
    ylab = "Estimated Probability of Survival")</pre>
```



#stratified by sex, Figure 11.3.

```
fit.sex <- survfit(Surv(time, status) ~ sex)
plot(fit.sex, xlab = "Months",
    ylab = "Estimated Probability of Survival", col = c(2,4))
legend("bottomleft", levels(sex), col = c(2,4), lty = 1)</pre>
```



#a log-rank test to compare the survival of males to females

```
logrank.test <- survdiff(Surv(time, status) ~ sex)
logrank.test</pre>
```

Call:

survdiff(formula = Surv(time, status) ~ sex)

```
N Observed Expected (0-E)^2/E (0-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 proportional hazards models

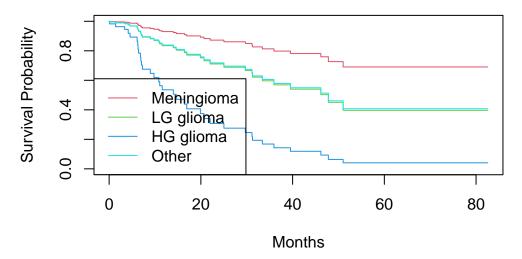
```
fit.cox <- coxph(Surv(time, status) ~ sex)
summary(fit.cox)</pre>
```

```
Call:
coxph(formula = Surv(time, status) ~ sex)
 n= 88, number of events= 35
         coef exp(coef) se(coef) z Pr(>|z|)
                1.5033 0.3420 1.192
sexMale 0.4077
       exp(coef) exp(-coef) lower .95 upper .95
           1.503
                   0.6652
                              0.769
sexMale
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 a model that makes use of additional predictors.
```

```
fit.all <- coxph(
Surv(time, status) ~ sex + diagnosis + loc + ki + gtv +
 stereo)
fit.all
Call:
coxph(formula = Surv(time, status) ~ sex + diagnosis + loc +
   ki + gtv + stereo)
                  coef exp(coef) se(coef)
                                         Z
sexMale
                2.49683 0.63816 1.434 0.15161
diagnosisLG glioma 0.91502
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
               gtv
stereoSRT
                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(</pre>
    diagnosis = levels(diagnosis),
    sex = rep("Female", 4),
   loc = rep("Supratentorial", 4),
   ki = rep(mean(ki), 4),
```

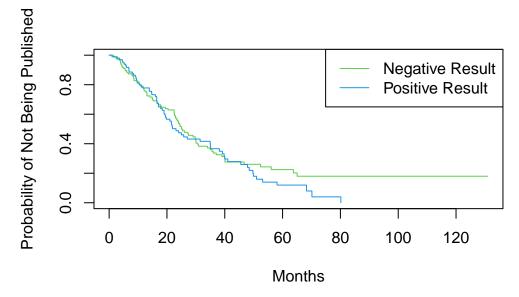
```
modaldata <- data.frame(
    diagnosis = levels(diagnosis),
    sex = rep("Female", 4),
    loc = rep("Supratentorial", 4),
    ki = rep(mean(ki), 4),
    gtv = rep(mean(gtv), 4),
    stereo = rep("SRT", 4)
    )

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)</pre>
```



#11.8.2 Publication Data #plotting the Kaplan-Meier curves stratified on the posres

```
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)</pre>
```



#fitting Cox's proportional hazards model to the posres variable

```
fit.pub <- coxph(Surv(time, status) ~ posres,</pre>
    data = Publication)
fit.pub
Call:
coxph(formula = Surv(time, status) ~ posres, data = Publication)
         coef exp(coef) se(coef)
posres 0.1481
                 1.1596
                          0.1616 0.916 0.36
Likelihood ratio test=0.83 on 1 df, p=0.3611
n= 244, number of events= 156
#log-rank test
logrank.test <- survdiff(Surv(time, status) ~ posres,</pre>
    data = Publication)
logrank.test
Call:
survdiff(formula = Surv(time, status) ~ posres, data = Publication)
           N Observed Expected (0-E)^2/E (0-E)^2/V
                          92.6
                   87
                                    0.341
posres=0 146
                                              0.844
                          63.4
                                    0.498
posres=1 98
                   69
                                              0.844
 Chisq= 0.8 on 1 degrees of freedom, p= 0.4
#include other predictors in the model
fit.pub2 <- coxph(Surv(time, status) ~ . - mech,</pre>
    data = Publication)
fit.pub2
Call:
coxph(formula = Surv(time, status) ~ . - mech, data = Publication)
               coef exp(coef)
                                  se(coef)
          5.708e-01 1.770e+00 1.760e-01 3.244 0.00118
posres
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 Likelihood ratio test=149.2 on 6 df, p=< 2.2e-16 n= 244, number of events= 156
```

#11.8.3 Call Center Data

```
set.seed(4)
N <- 2000
Operators <- sample(5:15, N, replace = T)
Center <- sample(c("A", "B", "C"), N, replace = T)
Time <- sample(c("Morn.", "After.", "Even."), N, replace = T)
X <- model.matrix( ~ Operators + Center + Time)[, -1]

#specify the coefficients and the hazard function
true.beta <- c(0.04, -0.3, 0, 0.2, -0.2)
h.fn <- function(x) return(0.00001 * x)</pre>
```

#Cox proportional hazards model

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

Warning in FUN(X[[i]], ...): 9 additional observations right-censored because the user-suppl is nonzero at the latest timepoint. To avoid these extra contact the supplementary of the supplementary o

```
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
```

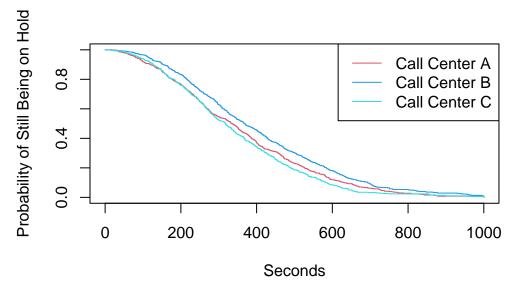
```
3
           7
                    0
                                                   0 187
                                                            TRUE
                             1
                                        1
4
           7
                    0
                             0
                                        0
                                                   0 279
                                                            TRUE
                    0
                                        0
5
          11
                             1
                                                   1 954
                                                            TRUE
           7
                    1
                             0
                                        0
                                                   1 455
                                                            TRUE
```

mean(queuing\$data\$failed)

[1] 0.89

#plot Kaplan-Meier survival curves. stratify by Center.

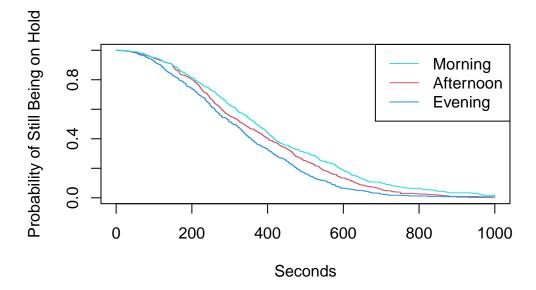
```
fit.Center <- survfit(Surv(y, failed) ~ Center,
    data = queuing$data)
plot(fit.Center, xlab = "Seconds",
    ylab = "Probability of Still Being on Hold",
    col = c(2, 4, 5))
legend("topright",
    c("Call Center A", "Call Center B", "Call Center C"),
    col = c(2, 4, 5), lty = 1)</pre>
```



#stratify by Time.

```
fit.Time <- survfit(Surv(y, failed) ~ Time,
    data = queuing$data)
plot(fit.Time, xlab = "Seconds",</pre>
```

```
ylab = "Probability of Still Being on Hold",
col = c(2, 4, 5))
legend("topright", c("Morning", "Afternoon", "Evening"),
col = c(5, 2, 4), lty = 1)
```



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

Call:

survdiff(formula = Surv(y, failed) ~ Center, data = queuing\$data)

N Observed Expected $(0-E)^2/E (0-E)^2/V$ 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

survdiff(Surv(y, failed) ~ Time, data = queuing\$data)

Call:

survdiff(formula = Surv(y, failed) ~ Time, data = queuing\$data)

N Observed Expected (0-E)^2/E (0-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 Cox's proportional hazards model to the data.

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

Call:

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

coefexp(coef)se(coef)zpOperators0.041741.042630.007595.5003.8e-08CenterB-0.218790.803490.05793-3.7770.000159CenterC0.079301.082530.058501.3560.175256TimeEven.0.209041.232490.058203.5920.000328TimeMorn.-0.173520.840700.05811-2.9860.002828

Likelihood ratio test=102.8 on 5 df, p=< 2.2e-16 n= 2000, number of events= 1780