

PH model checking

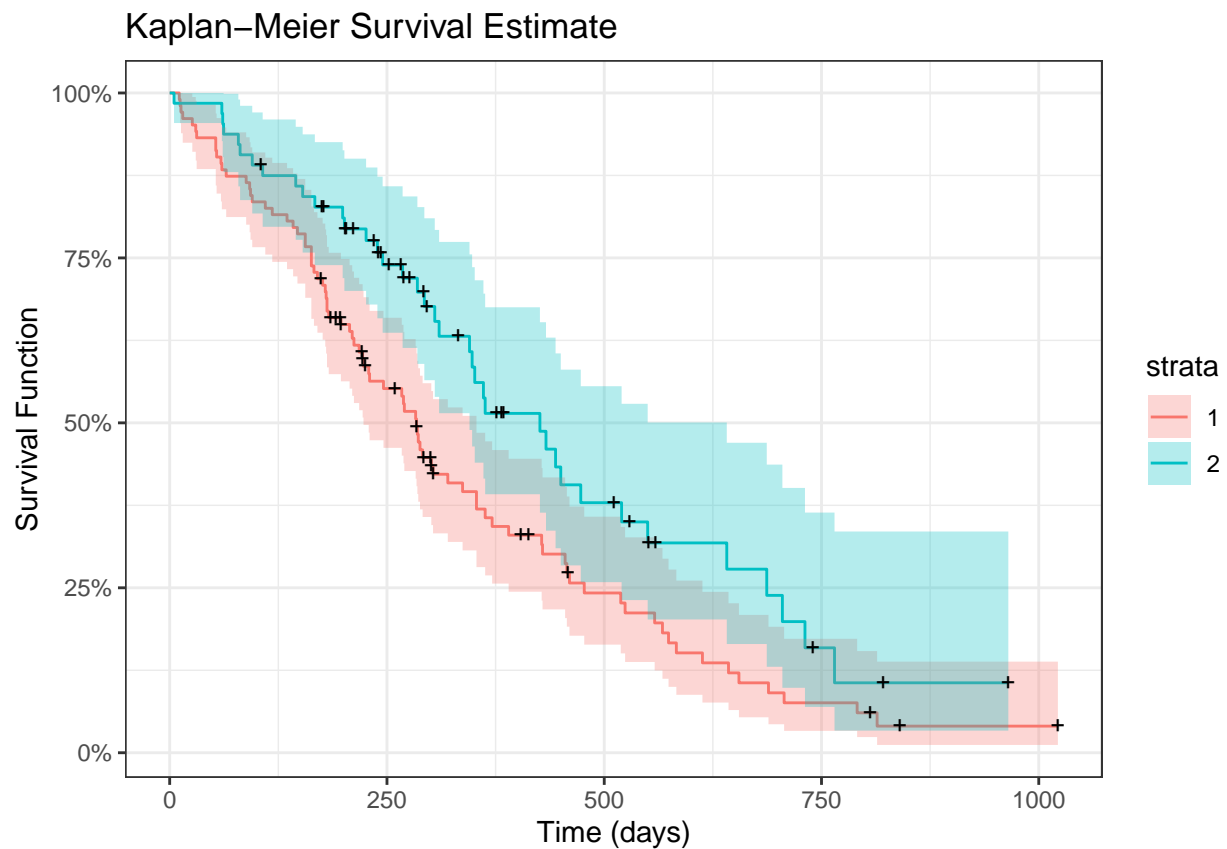
Qihang Wu

2022-December-01

Model checking I (graphical methods)

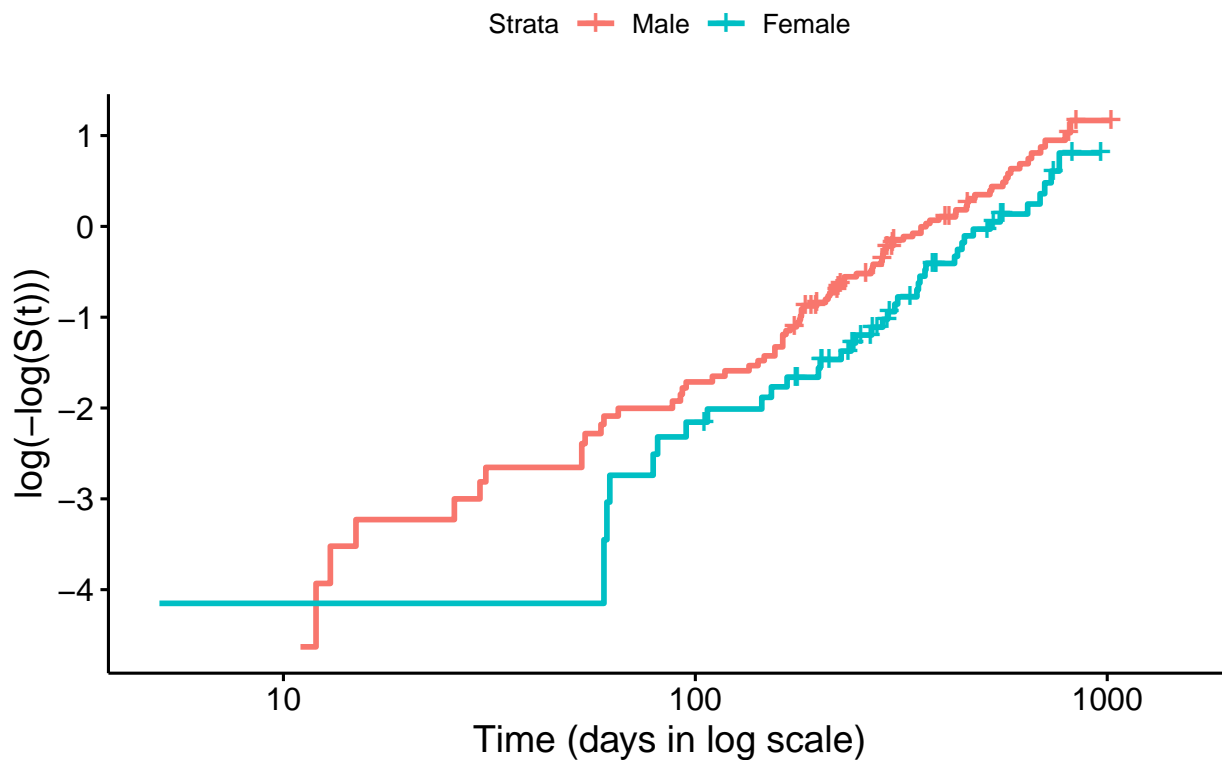
```
# import data
dat_lung_raw <- survival::lung
dat_lung <- dat_lung_raw %>%
  mutate_at(c(1, 3, 5, 6), .funs = ~as.factor(.)) %>% na.omit()

# --- sex ---
# km plot
fit_km_sex <- survfit(Surv(time, status == 2) ~ sex, dat_lung)
autoplot(fit_km_sex) + theme_bw() +
  labs(x = "Time (days)", y = "Survival Function",
       title = "Kaplan-Meier Survival Estimate")
```



```
# loglog vs. log time
ggsurvplot(fit_km_sex, fun = "cloglog", xlim = c(5, 1500),
  xlab = "Time (days in log scale)",
  title = "Log of Negative Log of Estimated Survival Functions",
  legend.labs = c("Male", "Female"))
```

Log of Negative Log of Estimated Survival Functions

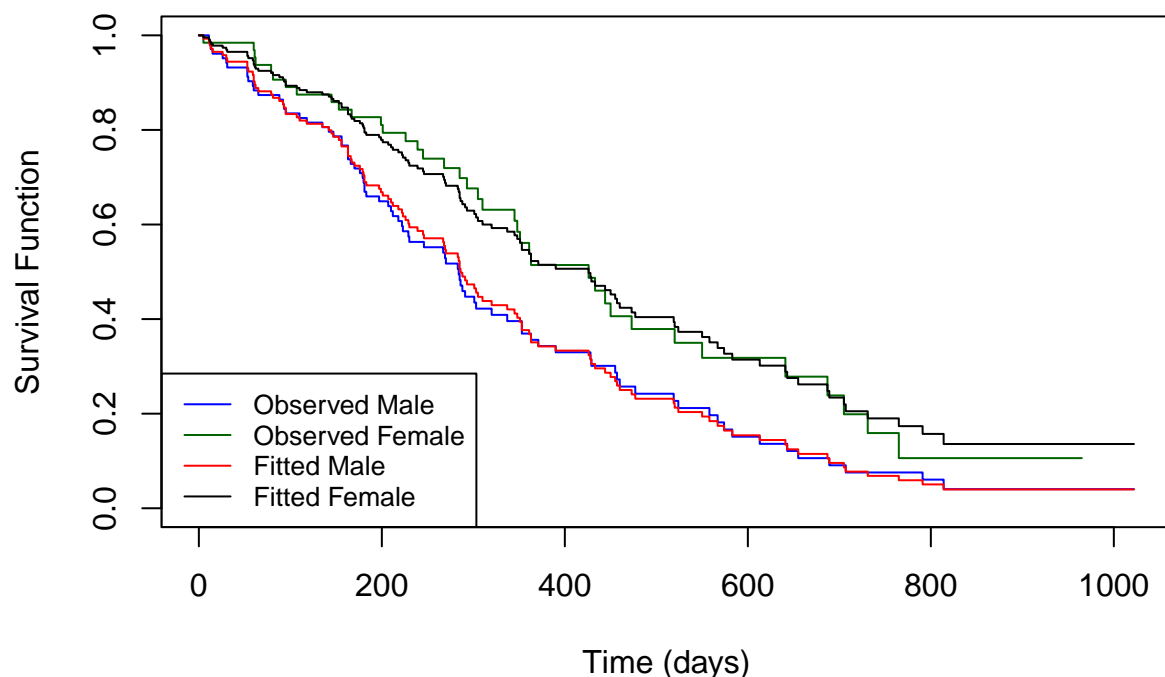


```
# ggsave("ph_checking_1.png", width = 10, height = 8)

# observed vs. fitted
fit_ph_sex <- coxph(Surv(time, status == 2) ~ sex, dat_lung)

# png("ph_checking_2.png", width = 500, height = 400)
plot(fit_km_sex, col = c("blue", "darkgreen"),
  xlab = "Time (days)", ylab = "Survival Function",
  main = "Observed vs. Fitted")
lines(survfit(fit_ph_sex, newdata = data.frame(sex = as.factor(1))), # male
  col = "red", conf.int = FALSE)
lines(survfit(fit_ph_sex, newdata = data.frame(sex = as.factor(2))), # female
  col = "black", conf.int = FALSE)
legend("bottomleft", legend = c("Observed Male", "Observed Female",
  "Fitted Male", "Fitted Female"),
  col = c("blue", "darkgreen", "red", "black"), lty = 1, cex = .8)
```

Observed vs. Fitted



Interpretation: The above two figures demonstrate the proportional hazards assumption is hold given there is only one indicator variable `sex` in the model.

Model checking II

```
# --- to be updated ---
# stepwise selection
stepw_res <- StepReg::stepwiseCox(Surv(time, status == 2) ~., dat_lung %>% na.omit(),
                                selection = "bidirection", select = "AIC")
stepw_res
```

```
## $'Basic Information'
##
## 1      Response Variable =  Surv(time, status == 2)
## 2      Included Variable =                               NULL
## 3      Selection Method =                               bidirection
## 4      Select Criterion =                               AIC
## 5 Multicollinearity Terms =                             efron
## 6      Method =                                           NULL
##
## $'Variable Class'
##      class                                variable
## 1  factor:                                inst sex ph.ecog
## 2 nmatrix.2:                               Surv(time, status == 2)
## 3  numeric: age ph.karno pat.karno meal.cal wt.loss
##
## $Process
```

```
## Step EnteredEffect RemovedEffect DF NumberIn AIC
## 2 1 ph.ecog 3 3 1008.58427913433
## 3 2 sex 1 4 1003.84247093824
## 4 3 wt.loss 1 5 1003.06920656013
## 5 4 ph.karno 1 6 1002.15358992223
##
## $Variables
## [1] "ph.ecog" "sex" "wt.loss" "ph.karno"
##
## $Coefficients
## coef exp(coef) se(coef) z Pr(>|z|)
## ph.ecog2 0.65012440 1.9157791 0.280695973 2.316116 0.0205519425
## ph.ecog3 1.67694368 5.3491822 0.441824118 3.795501 0.0001473459
## ph.ecog4 2.88359035 17.8783476 1.121914073 2.570242 0.0101627579
## sex2 -0.56468797 0.5685375 0.199666717 -2.828153 0.0046817462
## wt.loss -0.01279297 0.9872885 0.007676834 -1.666438 0.0956262406
## ph.karno 0.01853809 1.0187110 0.011153914 1.662025 0.0965077398

# interaction
fit_ph_1 <- coxph(Surv(time, status == 2) ~ sex + wt.loss + ph.karno +
  sex * log(time) + wt.loss * log(time) + ph.karno * log(time),
  dat_lung %>% na.omit())
summary(fit_ph_1)

## Call:
## coxph(formula = Surv(time, status == 2) ~ sex + wt.loss + ph.karno +
## sex * log(time) + wt.loss * log(time) + ph.karno * log(time),
## data = dat_lung %>% na.omit())
##
## n= 167, number of events= 120
##
## coef exp(coef) se(coef) z Pr(>|z|)
## sex2 -1.675e+00 1.874e-01 3.603e-01 -4.648 3.35e-06 ***
## wt.loss 8.469e-02 1.088e+00 1.208e-02 7.009 2.40e-12 ***
## ph.karno 7.232e-03 1.007e+00 1.390e-02 0.520 0.6028
## log(time) -1.786e+02 2.651e-78 2.613e+01 -6.835 8.18e-12 ***
## sex2:log(time) 2.974e-01 1.346e+00 6.319e-02 4.706 2.52e-06 ***
## wt.loss:log(time) -1.540e-02 9.847e-01 2.139e-03 -7.201 5.96e-13 ***
## ph.karno:log(time) -4.522e-03 9.955e-01 2.440e-03 -1.853 0.0638 .
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## exp(coef) exp(-coef) lower .95 upper .95
## sex2 1.874e-01 5.337e+00 9.249e-02 3.797e-01
## wt.loss 1.088e+00 9.188e-01 1.063e+00 1.114e+00
## ph.karno 1.007e+00 9.928e-01 9.802e-01 1.035e+00
## log(time) 2.651e-78 3.773e+77 1.511e-100 4.651e-56
## sex2:log(time) 1.346e+00 7.427e-01 1.190e+00 1.524e+00
## wt.loss:log(time) 9.847e-01 1.016e+00 9.806e-01 9.889e-01
## ph.karno:log(time) 9.955e-01 1.005e+00 9.907e-01 1.000e+00
##
## Concordance= 1 (se = 0 )
## Likelihood ratio test= 950.1 on 7 df, p=<2e-16
## Wald test = 195.2 on 7 df, p=<2e-16
```

```
## Score (logrank) test = 398.7 on 7 df, p=<2e-16
```

```
# residual
```

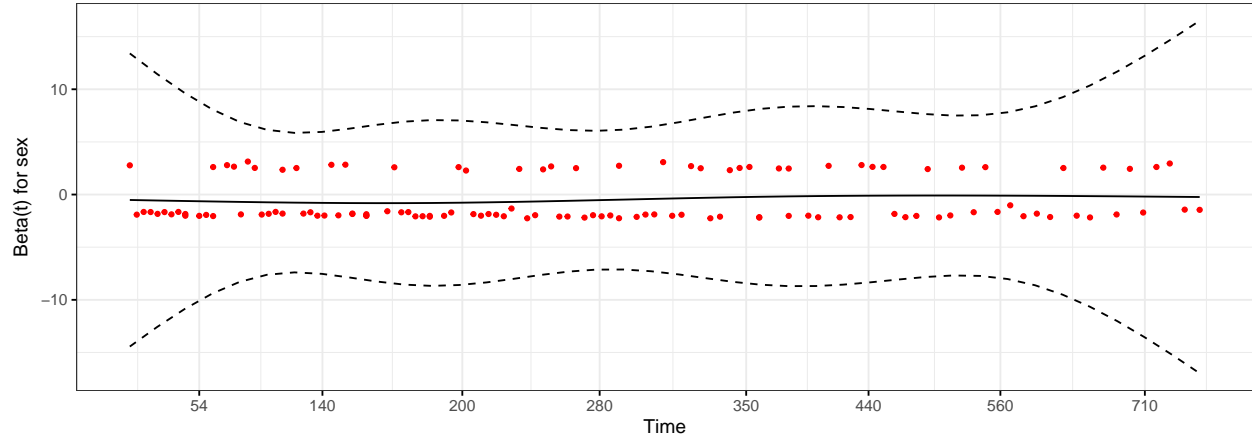
```
fit_ph_2 <- coxph(Surv(time, status == 2) ~ sex + wt.loss + ph.karno, dat_lung %>% na.omit())  
test_ph <- cox.zph(fit_ph_2)  
test_ph
```

```
##           chisq df      p  
## sex         1.405  1 0.2359  
## wt.loss      0.199  1 0.6559  
## ph.karno     9.488  1 0.0021  
## GLOBAL     11.406  3 0.0097
```

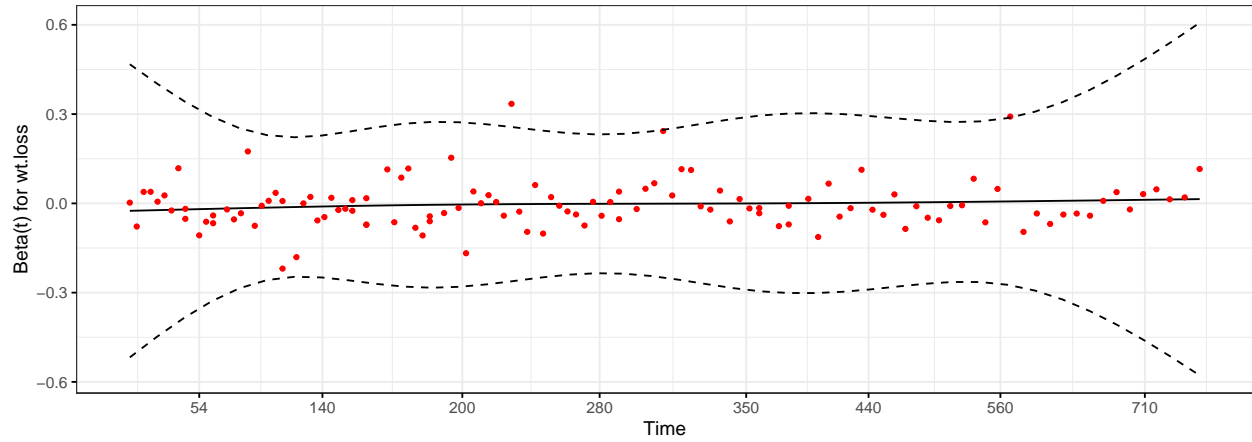
```
ggcoxzph(test_ph, ggtheme = theme_bw())
```

Global Schoenfeld Test p: 0.009723

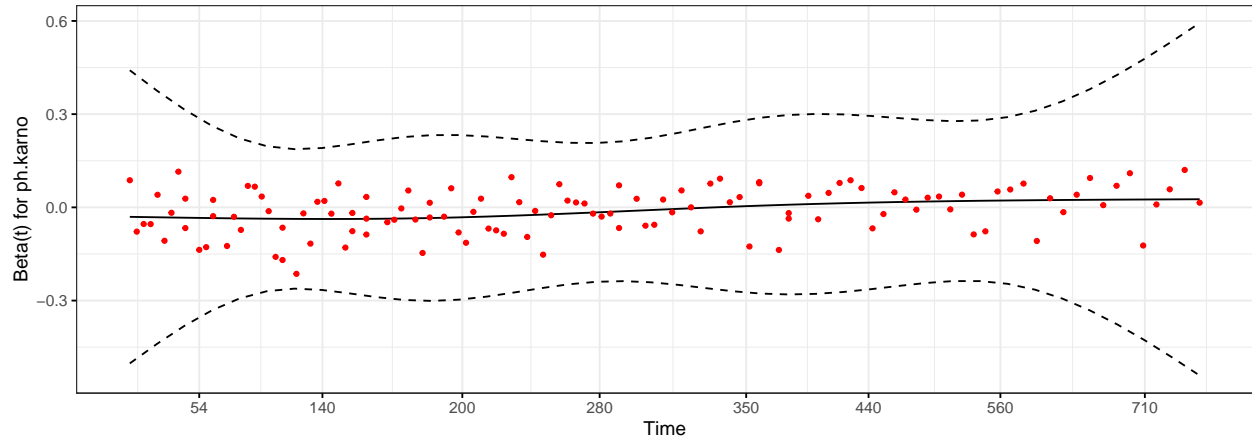
Schoenfeld Individual Test p: 0.2359



Schoenfeld Individual Test p: 0.6559



Schoenfeld Individual Test p: 0.0021



Description and Interpretation:

- Stepwise procedure was performed to select variables from the original dataset and finalize **4** variables including ph.ecog, sex, wt.loss, ph.karno;
- Test interaction for proportionality given **3** selected variables sex, wt.loss, ph.karno, the final result shows that only the interaction of **ph.karno** with **log(time)** is non-significant (i.e., greater than 0.05).

So we conclude that the proportionality assumption for `ph.karno` is violated. The Schoenfeld residual of the fitted model agrees with such conclusion.

Time-varying model

Ways to fit a time-varying model...