PH model checking

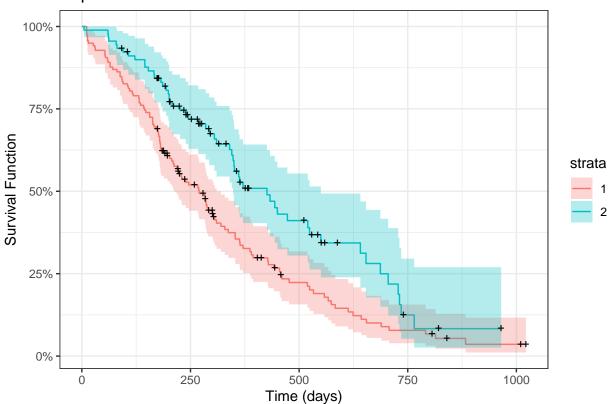
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Import data

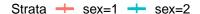
```
dat_lung_raw <- survival::lung
dat_lung <- dat_lung_raw %>%
  mutate_at(c(1, 3, 5, 6), .funs = ~as.factor(.))
```

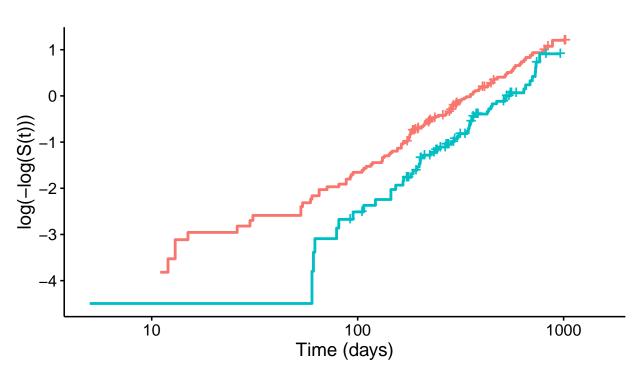
Model checking I

Kaplan-Meier Survival Estimate

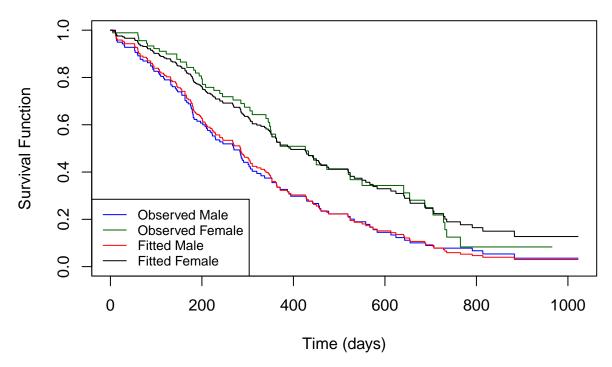


Log of Negative Log of Estimated Survival Functions





Observed vs. Fitted



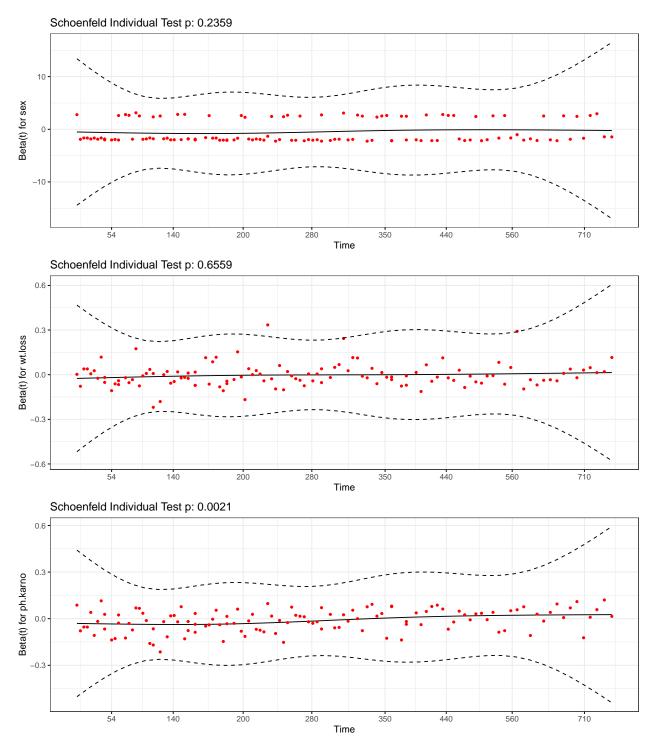
Interpretation: The above two figures demonstrate the proportional hazards assumption is hold given there is only one indicator variable **sex** in the model.

Model checking II

\$Process

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

```
Step EnteredEffect RemovedEffect DF NumberIn
                                                             AIC
## 2
               ph.ecog
                                     3
                                              3 1008.58427913433
       1
## 3
       2
                   sex
                                     1
                                              4 1003.84247093824
## 4
       3
                                     1
                                              5 1003.06920656013
               wt.loss
## 5
       4
              ph.karno
                                     1
                                              6 1002.15358992223
##
## $Variables
## [1] "ph.ecog"
                 "sex"
                            "wt.loss" "ph.karno"
##
## $Coefficients
##
                        exp(coef)
                                    se(coef)
                                                           Pr(>|z|)
                  coef
## 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
           ## 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 ***
                     7.232e-03 1.007e+00 1.390e-02 0.520
## ph.karno
                                                              0.6028
## log(time)
                     -1.786e+02 2.651e-78 2.613e+01 -6.835 8.18e-12 ***
                      2.974e-01 1.346e+00 6.319e-02 4.706 2.52e-06 ***
## sex2:log(time)
## 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
                     1.007e+00 9.928e-01 9.802e-01 1.035e+00
## ph.karno
## 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
```



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 <code>ph.karno</code> is violated. The Schoenfeld residual of the fitted model agrees with such conclusion.

Time-varying model

Ways to fit a time-varying model...