HW 5

February 28, 2025

1 Exercise 1

Use the dataset hodg from the package KMsurv for this exercise. You can load the data via the following sequence of commands: library(KMsurv); data("hodg")

- 1. Fit a Weibull regression using the survreg function in the survival package to the data with main effects for gtype, dtype and an interaction between them. You'll need to change gtype, dtype to factors before you run your analysis because they are coded as integers in the dataset.
- 2. Run a composite score test (see section 4.1.4 in the notes) to determine if the shape parameter (note that this corresponds to 1/scale in survreg's parameterization) is different than 1. Hint: This may involve fitting survreg with a different distribution than Weibull.
- 3. Test the proportional hazards assumption with a likelihood ratio test (see class notes, section 4.3.2). Be explicit about what null hypothesis you'll test, and the asymptotic distribution of the test statistic under the null hypothesis of proportional hazards. Hint: You can use the loglik element returned from surveg as part of this test.
- 4. Examine the fit of the model using the Cox-Snell residual model checking algorithm outlined in class notes section 4.4.2. You will have to define a cumulative hazard function manually; use the pweibull(q, shape, scale, lower.tail=FALSE) function for a shortcut.
- 5. Fit a log-logistic AFT model using the survreg function with the dist = "loglogistic" option specified with main effects for gtype, dtype and an interaction between them.

6. Examine the Cox-Snell residuals for the log-logistic model. R implements the survival function of a logistic distribution as plogis(q, location, scale, lower.tail=FALSE) and the AFT model parameters returned in the surveg object correspond directly to the location and scale parameters in the plogis R function.

2 Exercise 2

Use the dataset kidtran from the package KMsurv for this exercise. You can load the data via the following sequence of commands: library(KMsurv); data("kidtran")

- 1. Create a new categorical predictor variable age_gp which is a transformation of the age variable into age categories: (0,18], (18,40], (40,55], (55,∞]. Use the base R function cut for this task.
- 2. Fit a Weibull regression using the survreg function in the survival package to the data with a main effect for age_gp
- 3. Use the residuals function with the type argument set to "dfbetas". This function returns a matrix where each row corresponds to equation 4.57 in the class notes for the datapoint with index equal to the row index. This score vector is then scaled by the estimated asymptotic standard error of the parameter. Make a plot of the scaled influence statistic for the coefficient corresponding to the (18,40] age group. Do you see any outliers? If so, what is anomalous about this (these) datapoint(s)?
- 4. Refit the model without the outlier datapoint(s). How do the estimated regression coefficients change? If they change, why do the regression coefficients change? See if you can glean any insight from looking at the number of failures within each age group before and after you remove the outliers, and thinking about how you would solve the score equations for the Weibull rate parameter (inverse of the base R's scale parameter) in both scenarios.
- 5. How might you modify your analysis to avoid this issue?