## Problem Set 4

## Applied Stats II

Due: April 16, 2023

## Question 1

We're interested in modeling the historical causes of infant mortality. We have data from 5641 first-born in seven Swedish parishes 1820-1895. Using the "infants" dataset in the eha library, fit a Cox Proportional Hazard model using mother's age and infant's gender as covariates. Present and interpret the output.

```
# Load data
data(child)
head(child)

# Build a survival object out of the 'child' data.frame
child_surv <- with(child, Surv(enter, exit, event))

# Run a Cox Proportional Hazard regression
cox <- coxph(child_surv ~ m.age + sex, data = child)
summary(cox)
drop1(cox, test = "Chisq")
stargazer(cox, type = "text")
```

We can plot this survival function using a Kaplan Meyer plot:

```
# Plot
km <- survfit(child_surv ~ 1, data = child)
summary(km, times = seq(0, 15, 1))
plot(km, main = "Kaplan-Meier Plot", xlab = "Years", ylim = c(0.7, 1))

# Run a Cox Proportional Hazard regression
cox <- coxph(child_surv ~ m.age + sex, data = child)
summary(cox)
drop1(cox, test = "Chisq")
stargazer(cox, type = "text")</pre>
```

Please view output on next page:

## Kaplan-Meier Plot

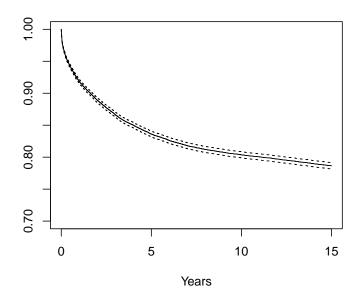


Figure 1: Child Survival Kaplan Meier plot

Table 1:

	Dependent variable:
	$\operatorname{child\_surv}$
m.age	0.008***
	(0.002)
sexfemale	-0.082***
	(0.027)
Observations	26,574
$\mathbb{R}^2$	0.001
Max. Possible $\mathbb{R}^2$	0.986
Log Likelihood	-56,503.480
Wald Test	$22.520^{***} (df = 2)$
LR Test	$22.518^{***} (df = 2)$
Score (Logrank) Test	$22.530^{***} (df = 2)$
Note:	*p<0.1; **p<0.05; ***p<0.0

2

Analysis: The results show that both mother's age and infant's gender are statistically significant predictors of child mortality, with p-values less than 0.01.

For mother's age, the coefficient estimate of 0.008 indicates that for each one-year increase in mother's age, the hazard of child mortality increases by a factor of 1.008 (exp(0.008)).

For infant's gender, the coefficient estimate of -0.082 indicates that the hazard of mortality is lower for female infants than for male infants by a factor of 0.921 (exp(-0.082)).

The drop1() function allows us to asses the model quality by removing each predictor one at a time from the model and comparing the fit of the reduced model to the original model using an LRT.

```
> drop1(cox, test = "Chisq")
Single term deletions

Model:
child_surv ~ m.age + sex
Df AIC LRT Pr(>Chi)
<none> 113011
m.age 1 113022 12.7946 0.0003476 ***
sex 1 113018 9.4646 0.0020947 **
---
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

Here we can see that that removing either mother's age or infant's gender from the model leads to a significant deterioration in the model fit based on the likelihood ratio test. This indicates that both predictors are important in predicting child survival.