Lab 13

Logistic Regression

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Question 1

Data were collected in an effort to relate the safety of certain vehicles to different aspects of those vehicles. This dataset has the following variables:

- Unsafe: binary safety designation (1 = below average (unsafe), 0 = average or above average (safe))
- Type: type of car (Large, Medium, Small, Sport/Utility, Sports)
- Region: manufacturing region (Asia, N America)
- Weight: integer value for car weight ranging from 1 to 6
- Size: size of car corresponding to Type (1 = Small/Sports, 2 = Medium, 3 = Large or Sport/Utility)

Part (a)

Build a logistic regression, predicting Unsafe using the variables Region, Weight, and Size. Treat Weight as a continuous variable. Treat Region and Size as categorical. Make sure to use the factor function for Size.

```
##
## Call:
## glm(formula = Unsafe ~ Region + Weight + Size, family = binomial(link = "logit"),
##
       data = safety)
## Coefficients:
                   Estimate Std. Error z value Pr(>|z|)
##
                               1.3949
## (Intercept)
                    2.7285
                                       1.956 0.05046
## RegionN America -0.3775
                               0.5624 -0.671 0.50203
## Weight
                    -0.6678
                               0.4589 -1.455 0.14559
```

a. Which variables were significant at the 0.05 level?

Solution: The variables that were significant at the 0.05 level were Size2 and Size3.

b. What is the concordance proportion for this model?

Solution: This model has a concordance proportion of 0.8543.

```
concordance(safety.glm)
```

```
## Call:
## concordance.lm(object = safety.glm)
##
## n= 96
## Concordance= 0.8482 se= 0.03897
## concordant discordant tied.x tied.y tied.xy
## 1622 243 115 2273 307
```

Part (b)

Remove variables one at a time that have a p-value above 0.05.

```
##
## glm(formula = Unsafe ~ Size, family = binomial(link = "logit"),
##
      data = safety)
##
## Coefficients:
              Estimate Std. Error z value Pr(>|z|)
## (Intercept) 0.6506
                       0.3561 1.827 0.067708 .
                          0.6070 -3.656 0.000256 ***
## Size2
              -2.2192
## Size3
              -3.3586
                          0.8125 -4.134 3.57e-05 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
```

```
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 119.249 on 95 degrees of freedom
## Residual deviance: 86.629 on 93 degrees of freedom
## AIC: 92.629
##
## Number of Fisher Scoring iterations: 5
```

a. What variables are left in the model?

Solution: Size is the only variable left in the model.

b. What is the proportion of concordance with your final model?

Solution: The concordance proportion is 0.8182.

concordance(safety.glm2)

```
## Call:
## concordance.lm(object = safety.glm2)
##
## n= 96
## Concordance= 0.8182 se= 0.043
## concordant discordant tied.x tied.y tied.xy
## 1392 132 456 1539 1041
```

c. What is the interpretation of the Size variable for comparing categories 1 to 3?

 $\textbf{Solution:} \ \, \text{Cars of Size2 decreases the expected odds of being unsafe by } 89.13\% \ \, \text{compared to cars of Size1}.$

or

Cars of Size2 increases the expected odds of being safe by 89.13% compared to cars of Size1.

Cars of Size3 increases the expected odds of being safe by 96.52% when compared to cars of Size1.

100*(exp(cbind(coef(safety.glm2), confint(safety.glm2)))-1)

```
## 2.5 % 97.5 %
## (Intercept) 91.66667 -2.84649 298.16962
## Size2 -89.13043 -96.98548 -66.42930
## Size3 -96.52174 -99.49411 -85.85478
```