## **Practical 6: Logistic Regression Model Diagnostics (Binary Data)**

The data to be used in this practical are from a survey of female labor force participation. The data is stored in a STATA file called "mroz.dta", which is available on Canvas.

```
library(foreign)
mroz.df<-read.dta("C:\\Temp folder\\mroz.dta")
attach(mroz.df)</pre>
```

View the names of the variables in the data-frame:

```
names (mroz.df)
```

1. Fit a logistic regression model on labor force participation with **age**, number of kids under age 6 (**kidslt6**) and the number of kids aged 6 or over (**kidsge6**).

```
glm1 <- glm(inlf ~ age + kidslt6 + kidsge6,
family = binomial (link = logit))
summary(glm1)</pre>
```

2. Calculate the residuals and other diagnostic statistics for the model as follows:

```
h<-lm.influence(glm1)$hat

rpear <-residuals(glm1, "pearson")/sqrt(1-h)

rdev <-residuals(glm1, "deviance")/sqrt(1-h)

phat<- glm1$fitted.values</pre>
```

```
D < -rpear * rpear * h / (4 * (1-h))
```

3. Plot the Pearson residuals versus the linear predictor. What should you expect?

```
plot(glm1$linear.predictors, rpear,main="Plot of
Pearson Residuals v Linear Predictor")
```

## From notes:

Of the methods studied for Binomial data, the following are applicable to binary data:

- Index Plot: plot of residuals versus case number.
- Leverage Values
- Influence Values

plot(rpear)

4. Obtain an index plot of the residuals. What should you expect? What do you conclude?

identify(rpear, n = 2)

5. Obtain an index plot of the leverage values. What do you conclude?

```
plot(h, main="Index Plot of Leverage Values")
abline(h=0.0106, lty=1)

identify(h, n = 2)

6. Obtain an index plot of the influence (D) values. What do you conclude?

plot(D, main="Index Plot of Cook's Distance Values")

identify(D, n = 2)
```

7. Investigate the cases of concern.

## **Outliers/High influence:**

```
mroz.df[c(74,400), c(1,3,4,5)]
```

## **High leverage:**

```
mroz.df[c(53, 720), c(1,3,4,5)]
```