#### **Generalized Linear Model**

Name: Rutvik Kolhe Unity ID: rkolhe

#### Soultions:

# Question1)

After bulding the logistic regression model for all predictors, we can infer that currency\_GBP is the predictor with the highest estimate for its coefficient. We then build fit.single with using currency\_GBP as the predictor. The equation of relation of Competitive? to currency\_GBP is:

Competitive? = 0.2193 - 1.7233\* categoryEverythingElse

a)

The odds that Y=yes for the given set of X values is given by  $\frac{P(Y = yes)}{1 - P(Y = yes)}$ 

The relationship of log-odds with the predicted variable is:

$$ln\left(\frac{P(Y=yes)}{1-P(Y=yes)}\right) = \beta_o + \beta_1 * X_1 + \beta_2 * X_2$$

ln (odds) = 0.2193 - 1.7233\* category Everything Else

$$odds = e^{0.2193-1.7233* categoryEverythingElse}$$

Let 
$$e^{0.2193-1.7233* categoryEverythingElse} = e^z$$

Hence, 
$$P(Y) = e^z (1 - P(Y))$$

$$P(Y)\left(1-e^z\right) = e^z$$

$$P(Y) = \frac{e^z}{1 - e^z}$$

$$P(Y = yes) = \frac{1}{1 + e^{0.2193 - 1.7233* category Everything Else}}$$

b)

ln(odds) = 0.2193 - 1.7233\* category Everything Else

$$odds = e^{0.2193-1.7233* categoryEverythingElse}$$

c)

$$Logit = ln\left(\frac{P(Y = yes)}{1 - P(Y = yes)}\right) = 0.2193 - 1.7233* category Everything Else$$

## Question 2)

The top four predictors for fit\_all model are: category\_EverythingElse, currency\_GBP, currency\_US, category\_Business/Industrial. Based on these predictors, we can express the different equations as follows:

b)

$$ln\left(odds
ight) = egin{array}{l} -2.5518*categoryEverythingElse + 2.202*currencyGBP \\ +0.8917*categoryBusinessIndustrial + 0.8735*currencyUS \\ \end{array}$$

$$odds = e^{-2.5518*categoryEverythingElse+2.202*currencyGBP+}$$

c)

As shown in solution 1a)

$$ln\left(\frac{P(Y = yes)}{1 - P(Y = yes)}\right) = \beta_o + \beta_1 * X_1 + \beta_2 * X_2 .....$$

$$P(Y) = \frac{e^z}{1 - e^z}$$

$$P(Y) = \frac{1}{-(-2.5518*categoryEverythingElse+2.202*currencyGBP+0.8917*categoryBusinessIndustrial+0.8735*currencyUS)}$$

#### Question 3)

The highest predictor for fit\_all is category\_EverythingElse. The generalized equation for odds can be given as:

$$odds = e^{\beta_o + \beta_1 * X_1 + \beta_2 * X_2}$$

Hence, the value of odds for fit all is:

$$-2.5518* category Everything Else + 2.202* currency GBP + odds = e^{0.8917* category Business Industrial + 0.8735* currency US}$$

Now, if we increse the value of currency\_GBP by 1 and keep the value of coefficients constant, the corresponding equation for odds is:

$$-2.5518* \ ( {\it categoryEverythingElse+1}) + 2.202* currencyGBP + odds' = e^{0.8917* categoryBusinessIndustrial + 0.8735* currencyUS}$$

Hence the odds ratio is:

$$\frac{odds'}{odds} = \frac{e^{-2.5518*(\text{categoryEverythingElse} + 1) + 2.202*currencyGBP} + e^{0.8917*categoryBusinessIndustrial + 0.8735*currencyUS}}{e^{-2.5518*categoryEverythingElse + 2.202*currencyGBP} + e^{0.8917*categoryBusinessIndustrial + 0.8735*currencyUS}}$$

$$= \frac{e^{-2.5518*categoryEverythingElse + 2.202*currencyGBP} + e^{0.8917*categoryBusinessIndustrial + 0.8735*currencyUS} * e^{-2.5518}}{e^{-2.5518*categoryEverythingElse + 2.202*currencyGBP} + e^{0.8917*categoryBusinessIndustrial + 0.8735*currencyUS}}$$

$$= e^{-2.5518}$$

Therefore, if the value of currency\_GBP increases by 1, the value of response changes by a factor of  $e^{-2.5518}$ .

If it was linear regression, then the value of response would change by the factor of 2.5518 (coefficient) times. Since, the value of logistic regression gives us the output of logit function, we have to find the value of  $e^{value}$ . Whereas in linear regression the coefficient output directly affects the response

### Question 4)

We can use anova test to check if the two models- fit\_reduced and fit\_all are equivalent to each other or nont. After running anova test on the two models. The p\_value obtained is 0.7086. Since p-value is greater than 0.05, we can conclude that the two models do not significantly differ from each other.

## Question 5)

Overdispersion occurs when the value of 
$$\phi = \frac{Residual\ deviance}{Residual\ df} \ \gg 1$$
 . In this case,

 $\phi = \text{0.992}$  . Since the value of  $\phi$  is close to 1, there is no overdispersion present in the model.

If there is any overdispersion, then we can use quasi-binomial distribution instead of binomial family distribution.