COSC6323 - Exercise 9

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Task 1:

Work with the project data. Build a regression model based on the article level data which is predicting the cross-disciplinary in CIP (XCIPp variable) Classification of Instructional Programs, using year of publication, log-transformed coauthors count, log-transformed Major MeSH count, regions count and total number of SAps (NSAp) as predictors. Get model summary, pseudo r-squared measures, odds ratio and comment about the results.

Solution:

```
library(tidyverse)
## -- Attaching packages ----- tidyverse
1.3.0 --
## v ggplot2 3.3.3 v purrr 0.3.4
## v tibble 3.0.5 v dplyr 1.0.3
## v tidyr 1.1.2 v stringr 1.4.0
## v readr 1.4.0 v forcats 0.5.1
## -- Conflicts -----
tidyverse conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
library(modelr)
library(broom)
## Warning: package 'broom' was built under R version 4.0.4
##
## Attaching package: 'broom'
## The following object is masked from 'package:modelr':
##
##
       bootstrap
library(ROCR)
## Warning: package 'ROCR' was built under R version 4.0.5
library(questionr)
## Warning: package 'questionr' was built under R version 4.0.5
```

```
setwd("D:/Statistical Methods/Project")
df_article<-read.csv("ArticleLevel-RegData-</pre>
ALLSA Xc 1 NData 655386 LONGXCIP2.csv")
df_article<-as_tibble(df_article)</pre>
model1 <- glm(XCIPp ~ Yp + log(Kp) + log(nMeSHMain) + NRegp + NSAp, data =</pre>
df article, family=binomial(link='logit'))
summary(model1)
##
## Call:
## glm(formula = XCIPp \sim Yp + log(Kp) + log(nMeSHMain) + NRegp +
       NSAp, family = binomial(link = "logit"), data = df_article)
##
## Deviance Residuals:
##
      Min
                 10
                      Median
                                   30
                                           Max
## -3.2664 -0.4450 -0.3693 -0.2874
                                        2.8338
##
## Coefficients:
                    Estimate Std. Error z value Pr(>|z|)
##
                  -2.157e+01 1.093e+00 -19.736 < 2e-16 ***
## (Intercept)
## Yp
                   7.752e-03 5.467e-04 14.178 < 2e-16 ***
                   6.198e-01 7.180e-03 86.322 < 2e-16 ***
## log(Kp)
## log(nMeSHMain) -3.984e-02 1.090e-02 -3.655 0.000257 ***
                   1.992e+00 7.411e-03 268.825 < 2e-16 ***
## NRegp
                   1.938e-01 4.416e-03 43.890 < 2e-16 ***
## NSAp
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
##
       Null deviance: 542453 on 655385 degrees of freedom
## Residual deviance: 418616 on 655380 degrees of freedom
## AIC: 418628
##
## Number of Fisher Scoring iterations: 5
pscl::pR2(model1)["McFadden"]
## fitting null model for pseudo-r2
## McFadden
## 0.2282903
output = odds.ratio(model1) # HEAVY COMPUTATIONAL!
## Waiting for profiling to be done...
output = apply(output, 2, formatC, format="f", digits=4)
output
```

```
## (Intercept) "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.0000" "0.000" "0.000" "0.000" "0.000" "
```

Conclusion:

The summary of the model shows that:

- 1. All of the variables are highly significant in predicting cross-disciplinary in CIP with all p-values are quite less than 0.05. The number of co-authors and region the article was written appear to also have a close standard error.
- 2. As we know that values of pseudo r-squared [ρ 2] between 0.2 to 0.4 represent an excellent fit.Our model pseudo r-squared [ρ 2] is 0.2282903, which represent our model is excellent fit.
- 3. The odds ratio shows that from the data given, the region from where the article research took place was 7.33 times more successful in predicting whether the article had cross-disciplinary co-authors. The log number of co-authors was also successful in predicting whether the article had cross-disciplinary co-authors with a success-to-failure ratio of 1.85. All variables in the odds ratio were shown to be extremely significant.