## Course 2 Section - 3.7 GOODNESS OF FIT

Jiaying Wu

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```
#load library
library(tidyverse)
library(broom)
library(lubridate)
```

```
# Read CO2 data and apply some pre-processing
CO2.spo <- read_csv(
   "https://raw.githubusercontent.com/datascienceprogram/ids_course_data/master/daily_merge_co2_spo.csv"
   col_names = c("date", "time", "day", "decdate", "n", "flg", "co2"), skip = 69) %>%
   filter(flg == 0) %>%
    mutate(date = ymd(date))

# Create variable dayO (a rescaling of day)
CO2.spo <- CO2.spo %>%
   mutate(dayO = day - min(day))
```

## Give it a go

Explore the linear model of CO2 and with and without the quadratic term:

```
co2 = \beta_0 + \beta_1 day_0 + \epsilonco2 = \beta_0 + \beta_1 day_0 + \beta_2 day_0^2 + \epsilon
```

- What is the  $adjusted R^2$  and BIC for both models?
- Which is the preferred model?

## Model 1

```
co2_mod1 <- lm(co2~day0, data=CO2.spo)
tidy(co2_mod1)</pre>
```

```
glance(co2_mod1) %>%
 select(adj.r.squared, AIC)
## # A tibble: 1 x 2
   adj.r.squared AIC
            <dbl> <dbl>
##
## 1
            0.984 6193.
Model 2
co2_mod2 \leftarrow lm(co2\sim day0+I(day0^2), data=CO2.spo)
tidy(co2_mod2)
## # A tibble: 3 x 5
    term estimate std.error statistic p.value
##
##
     <chr> <dbl>
                            <dbl>
                                       <dbl>
## 1 (Intercept) 3.13e+2 1.01e- 1
                                       3096.
                                                   0
## 2 day0
                1.97e-3 1.80e- 5
                                        109.
                                                   0
## 3 I(day0^2)
                 9.42e-8 7.13e-10
                                        132.
                                                   0
glance(co2_mod2) %>%
 select(adj.r.squared, AIC)
## # A tibble: 1 x 2
   adj.r.squared AIC
##
            <dbl> <dbl>
## 1
            0.999 2898.
```

Since model 2 have the higher  $adjusted - R^2$  and lower BIC, the model 2 is preferred.