

Live Session - Week 3: Discrete Response Models

Lecture 3

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#Required Readings:

BL2015: Christopher R. Bilder and Thomas M. Loughin. Analysis of Categorical Data with R. CRC Press. 2015.

- Ch. 2.2.5 – 2.2.8, 2.3

Topics covered in Week 3

- Variable transformation: interactions among explanatory variables
- Variable transformation: quadratic term
- Categorical explanatory variables
- Odds ratio in the context of categorical explanatory variables
- Convergence criteria and complete separation

Familiarity with the concepts and techniques covered in this and last lecture are critical, as they will be used frequently in the next two lectures in situations that are more general (from two categorical to $J > 2$ categories and from unordered categorical variables to ordinal variables). With multinomial logistic regression models, the notation will be heavier.

The key objectives in this live session are to learn how to incorporate various transformation of variables (or, in machine learning terminology, “feature engineering”) and interpret the results when these transformed variables are part of the model specification. Variable transformations (or feature engineering) are useful in real life statistical and machine learning modeling.

In general, the odd ratios answer the question “how much the odds of success have changed by k-unit increase?” The amazing feature of logistic regression model is that the odd ratios (of the odds of success before and after the k-unit increase in a particular explanatory variable) is simplified to the exponential of the product between k and the coefficient estimate associated with that variable. That is, “the odds of a success change by $\exp(k\beta_j)$ times for every k-unit increase in x_j ”

Review some concepts from w203

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + (\beta_3 x_1 \times x_2) + \epsilon$$

$$\frac{\partial y}{\partial x_1} = \beta_1 + \beta_3 x_2$$

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_1^2 + \epsilon$$

$$\frac{\partial y}{\partial x_1} = \beta_1 + 2\beta_2 x_1$$

Start-up code

```
library(knitr)
opts_chunk$set(tidy.opts=list(width.cutoff=60),tidy=TRUE)

# Set working directory
#setwd("~/Documents/Teach/Cal/w271/LiveSessions/week03/")
#wd <- getwd()
#wd

# Start with a clean R environment
rm(list = ls())

# Load Libraries
library(car)
library(dplyr)
library(Hmisc)
library(skimr)
library(ggplot2)
library(stargazer)
library(mcprofile)
```

In this live session, we will practice binary logistic regression modeling, with a focus on the materials covered in week 3, using an autism screening dataset (for toddlers) obtained from kaggle

The dataset comes with a very detailed description in the included Word document. Below are some brief description provided on the aforementioned kaggle webpage:

Context: The dataset was developed by Dr Fadi Fayez Thabtah (fadifayez.com) using a mobile app called ASDTests (ASDtests.com) to screen autism in toddlers. See the description file attached with the CSV data to know more about the variables and the class. This data can be used for descriptive and predictive analyses such as classification, clustering, regression, etc. You may use it to estimate the predictive power of machine learning techniques in detecting autistic traits.

Brief Description of the Variables This data page on kaggle also provide some very basic descriptive graphs on the variables in this dataset.

A1 - A10: Items within Q-Chat-10 in which questions possible answers : “Always, Usually, Sometimes, Rarely & Never” items’ values are mapped to “1” or “0” in the dataset.

There are two variables in the data that will not be used in our analysis: 1. `Case_No`: the individual case number; this is an identifier variable 2. `Qchat.10.Score`: the dataset document suggests that this variable not be used in a classification problem, as the score variable is used to defined the `Class.ASD.Traits`.

```
# Load data
autism <- read.csv("autism.csv", header = TRUE, sep = ",")

# Attach the dataframe autism
attach(autism)

# View(autism)

# Examine the structure of the data
str(autism)
```

```
## 'data.frame':    1054 obs. of  19 variables:
## $ Case_No      : int  1 2 3 4 5 6 7 8 9 10 ...
## $ A1           : int  0 1 1 1 1 1 1 0 0 1 ...
## $ A2           : int  0 1 0 1 1 1 0 1 0 1 ...
## $ A3           : int  0 0 0 1 0 0 0 0 0 1 ...
## $ A4           : int  0 0 0 1 1 0 1 0 0 0 ...
## $ A5           : int  0 0 0 1 1 1 1 1 0 1 ...
## $ A6           : int  0 1 0 1 1 1 1 0 0 1 ...
## $ A7           : int  1 1 1 1 1 1 0 1 1 0 ...
## $ A8           : int  1 0 1 1 1 1 0 1 0 1 ...
## $ A9           : int  0 0 0 1 1 1 1 1 0 1 ...
## $ A10          : int  1 0 1 1 1 1 0 1 1 1 ...
## $ Age_Mons     : int  28 36 36 24 20 21 33 33 36 22 ...
## $ Qchat.10.Score : int  3 4 4 10 9 8 5 6 2 8 ...
## $ Sex          : Factor w/ 2 levels "f","m": 1 2 2 2 1 2 2 2 2 2 ...
## $ Ethnicity    : Factor w/ 11 levels "asian","black",...: 5 11 5 3 11 2 1 1 1 10 ...
## $ Jaundice     : Factor w/ 2 levels "no","yes": 2 2 2 1 1 1 2 2 1 1 ...
## $ Family_mem_with_ASD : Factor w/ 2 levels "no","yes": 1 1 1 1 2 1 1 1 1 1 ...
## $ Who.completed.the.test: Factor w/ 5 levels "family member",...: 1 1 1 1 1 1 1 1 1 3 ...
## $ Class.ASD.Traits. : Factor w/ 2 levels "No","Yes": 1 2 2 2 2 2 2 2 1 2 ...
```

```
# Conduct some very basic EDA
describe(autism)
```

```
## autism
##
## 19 Variables      1054 Observations
## -----
```

Case_No	n	missing	distinct	Info	Mean	Gmd	.05	.10
	1054	0	1054	1	527.5	351.7	53.65	106.30
	.25	.50	.75	.90	.95			
	264.25	527.50	790.75	948.70	1001.35			

```
##
## lowest :      1      2      3      4      5, highest: 1050 1051 1052 1053 1054
## -----
```

A1	n	missing	distinct	Info	Sum	Mean	Gmd
	1054	0	2	0.738	594	0.5636	0.4924

```
##
## -----
```

```

## A2
##      n missing distinct      Info      Sum      Mean      Gmd
##    1054         0         2    0.742     473    0.4488    0.4952
##
## -----
## A3
##      n missing distinct      Info      Sum      Mean      Gmd
##    1054         0         2    0.721     423    0.4013    0.481
##
## -----
## A4
##      n missing distinct      Info      Sum      Mean      Gmd
##    1054         0         2    0.75      540    0.5123    0.5002
##
## -----
## A5
##      n missing distinct      Info      Sum      Mean      Gmd
##    1054         0         2    0.748     553    0.5247    0.4993
##
## -----
## A6
##      n missing distinct      Info      Sum      Mean      Gmd
##    1054         0         2    0.732     608    0.5769    0.4887
##
## -----
## A7
##      n missing distinct      Info      Sum      Mean      Gmd
##    1054         0         2    0.683     685    0.6499    0.4555
##
## -----
## A8
##      n missing distinct      Info      Sum      Mean      Gmd
##    1054         0         2    0.745     484    0.4592    0.4971
##
## -----
## A9
##      n missing distinct      Info      Sum      Mean      Gmd
##    1054         0         2    0.75      516    0.4896    0.5003
##
## -----
## A10
##      n missing distinct      Info      Sum      Mean      Gmd
##    1054         0         2    0.728     618    0.5863    0.4856
##
## -----
## Age_Mons
##      n missing distinct      Info      Mean      Gmd      .05      .10
##    1054         0         25    0.971    27.87    8.859      12      15
##      .25      .50      .75      .90      .95
##      23      30      36      36      36
##
## lowest : 12 13 14 15 16, highest: 32 33 34 35 36
## -----
## Qchat.10.Score

```

```

##          n missing distinct      Info      Mean      Gmd      .05      .10
##      1054         0         11    0.991    5.213    3.338         0         1
##      .25      .50      .75      .90      .95
##         3         5         8         9         10
##
## Value          0         1         2         3         4         5         6         7         8         9
## Frequency      54      88      88      96     110     120      96     135      97      95
## Proportion 0.051 0.083 0.083 0.091 0.104 0.114 0.091 0.128 0.092 0.090
##
## Value          10
## Frequency       75
## Proportion 0.071
## -----
## Sex
##          n missing distinct
##      1054         0         2
##
## Value          f         m
## Frequency      319      735
## Proportion 0.303 0.697
## -----
## Ethnicity
##          n missing distinct
##      1054         0         11
##
## asian (299, 0.284), black (53, 0.050), Hispanic (40, 0.038), Latino (26,
## 0.025), middle eastern (188, 0.178), mixed (8, 0.008), Native Indian (3,
## 0.003), Others (35, 0.033), Pacifica (8, 0.008), south asian (60, 0.057),
## White European (334, 0.317)
## -----
## Jaundice
##          n missing distinct
##      1054         0         2
##
## Value          no      yes
## Frequency      766      288
## Proportion 0.727 0.273
## -----
## Family_mem_with_ASD
##          n missing distinct
##      1054         0         2
##
## Value          no      yes
## Frequency      884      170
## Proportion 0.839 0.161
## -----
## Who.completed.the.test
##          n missing distinct
##      1054         0         5
##
## family member (1018, 0.966), Health care professional (5, 0.005), Health
## Care Professional (24, 0.023), Others (3, 0.003), Self (4, 0.004)
## -----
## Class.ASD.Traits.

```

```
##          n missing distinct
##      1054         0         2
##
## Value          No    Yes
## Frequency      326    728
## Proportion 0.309 0.691
## -----
```

```
summary(autism)
```

```
##      Case_No          A1          A2          A3
##  Min.   : 1.0    Min.   :0.0000    Min.   :0.0000    Min.   :0.0000
## 1st Qu.: 264.2    1st Qu.:0.0000    1st Qu.:0.0000    1st Qu.:0.0000
## Median : 527.5    Median :1.0000    Median :0.0000    Median :0.0000
## Mean   : 527.5    Mean   :0.5636    Mean   :0.4488    Mean   :0.4013
## 3rd Qu.: 790.8    3rd Qu.:1.0000    3rd Qu.:1.0000    3rd Qu.:1.0000
## Max.   :1054.0    Max.   :1.0000    Max.   :1.0000    Max.   :1.0000
##
##      A4          A5          A6          A7
##  Min.   :0.0000    Min.   :0.0000    Min.   :0.0000    Min.   :0.0000
## 1st Qu.:0.0000    1st Qu.:0.0000    1st Qu.:0.0000    1st Qu.:0.0000
## Median :1.0000    Median :1.0000    Median :1.0000    Median :1.0000
## Mean   :0.5123    Mean   :0.5247    Mean   :0.5769    Mean   :0.6499
## 3rd Qu.:1.0000    3rd Qu.:1.0000    3rd Qu.:1.0000    3rd Qu.:1.0000
## Max.   :1.0000    Max.   :1.0000    Max.   :1.0000    Max.   :1.0000
##
##      A8          A9          A10         Age_Mons
##  Min.   :0.0000    Min.   :0.0000    Min.   :0.0000    Min.   :12.00
## 1st Qu.:0.0000    1st Qu.:0.0000    1st Qu.:0.0000    1st Qu.:23.00
## Median :0.0000    Median :0.0000    Median :1.0000    Median :30.00
## Mean   :0.4592    Mean   :0.4896    Mean   :0.5863    Mean   :27.87
## 3rd Qu.:1.0000    3rd Qu.:1.0000    3rd Qu.:1.0000    3rd Qu.:36.00
## Max.   :1.0000    Max.   :1.0000    Max.   :1.0000    Max.   :36.00
##
## Qchat.10.Score Sex          Ethnicity Jaundice
##  Min.   : 0.000 f:319    White European:334 no :766
## 1st Qu.: 3.000 m:735    asian          :299 yes:288
## Median : 5.000    middle eastern:188
## Mean   : 5.213    south asian   : 60
## 3rd Qu.: 8.000    black         : 53
## Max.   :10.000    Hispanic      : 40
##              (Other) : 80
## Family_mem_with_ASD Who.completed.the.test Class.ASD.Traits.
## no :884             family member :1018    No :326
## yes:170             Health care professional: 5    Yes:728
##                   Health Care Professional: 24
##                   Others : 3
##                   Self : 4
##
##
```

```
skim(autism)
```

```
## Skim summary statistics
## n obs: 1054
```

```
## n variables: 19
##
## -- Variable type:factor -----
##      variable missing complete    n n_unique
##      Class.ASD.Traits.      0    1054 1054      2
##      Ethnicity              0    1054 1054     11
##      Family_mem_with_ASD     0    1054 1054      2
##      Jaundice                0    1054 1054      2
##      Sex                    0    1054 1054      2
##      Who.completed.the.test  0    1054 1054      5
##
##      top_counts ordered
##      Yes: 728, No: 326, NA: 0 FALSE
##      Whi: 334, asi: 299, mid: 188, sou: 60 FALSE
##      no: 884, yes: 170, NA: 0 FALSE
##      no: 766, yes: 288, NA: 0 FALSE
##      m: 735, f: 319, NA: 0 FALSE
##      fam: 1018, Hea: 24, Hea: 5, Sel: 4 FALSE
##
## -- Variable type:integer -----
##      variable missing complete    n  mean    sd p0    p25    p50    p75
##      A1      0    1054 1054    0.56    0.5  0    0      1      1
##      A10     0    1054 1054    0.59    0.49 0    0      1      1
##      A2      0    1054 1054    0.45    0.5  0    0      0      1
##      A3      0    1054 1054    0.4     0.49 0    0      0      1
##      A4      0    1054 1054    0.51    0.5  0    0      1      1
##      A5      0    1054 1054    0.52    0.5  0    0      1      1
##      A6      0    1054 1054    0.58    0.49 0    0      1      1
##      A7      0    1054 1054    0.65    0.48 0    0      1      1
##      A8      0    1054 1054    0.46    0.5  0    0      0      1
##      A9      0    1054 1054    0.49    0.5  0    0      0      1
##      Age_Mons 0    1054 1054   27.87    7.98 12   23     30     36
##      Case_No  0    1054 1054  527.5   304.41 1  264.25 527.5 790.75
##      Qchat.10.Score 0    1054 1054   5.21    2.91 0    3      5      8
##      p100     hist
##      1
##      1
##      1
##      1
##      1
##      1
##      1
##      1
##      1
##      1
##      1
##      36
##      1054
##      10
```

```
# skim(Age_Mons)
```

```
# Define a function to examine factor variables:
```

```
exam_cat_var = function(var.names) {
  table(var.names)
  round(prop.table(table(var.names)), 2)
```

```
}
apply(autism[, 14:19], 2, table)
```

```
## $Sex
##
##   f   m
## 319 735
##
## $Ethnicity
##
##      asian      black      Hispanic      Latino middle eastern
##      299      53      40      26      188
##      mixed Native Indian      Others      Pacifica      south asian
##      8      3      35      8      60
## White European
##      334
##
## $Jaundice
##
##   no yes
## 766 288
##
## $Family_mem_with_ASD
##
##   no yes
## 884 170
##
## $Who.completed.the.test
##
##      family member Health care professional Health Care Professional
##      1018      5      24
##      Others      Self
##      3      4
##
## $Class.ASD.Traits.
##
##   No Yes
## 326 728
```

```
apply(autism[, 14:19], 2, exam_cat_var)
```

```
## $Sex
## var.names
##   f   m
## 0.3 0.7
##
## $Ethnicity
## var.names
##      asian      black      Hispanic      Latino middle eastern
##      0.28      0.05      0.04      0.02      0.18
##      mixed Native Indian      Others      Pacifica      south asian
##      0.01      0.00      0.03      0.01      0.06
## White European
##      0.32
```



```
##
## $Jaundice
## var.names
##   no  yes
## 0.73 0.27
##
## $Family_mem_with_ASD
## var.names
##   no  yes
## 0.84 0.16
##
## $Who.completed.the.test
## var.names
##           family member Health care professional Health Care Professional
##                0.97                                0.00                    0.02
##                Others                                Self
##                0.00                                0.00
##
## $Class.ASD.Traits.
## var.names
##   No  Yes
## 0.31 0.69

# Age describe(Age_Mons) summary(Age_Mons)
skim(Age_Mons)

##
## Skim summary statistics
##
## -- Variable type:integer -----
## variable missing complete    n mean  sd p0 p25 p50 p75 p100    hist
## Age_Mons          0      1054 1054 27.87 7.98 12  23  30  36   36

# Crosstab
xtabs(~Sex + Class.ASD.Traits.)

##      Class.ASD.Traits.
## Sex   No  Yes
## f 125 194
## m 201 534

round(prop.table(xtabs(~Sex + Class.ASD.Traits.), 1), 2)

##      Class.ASD.Traits.
## Sex   No  Yes
## f 0.39 0.61
## m 0.27 0.73

xtabs(~Ethnicity + Class.ASD.Traits.)

##           Class.ASD.Traits.
## Ethnicity   No  Yes
##   asian      87 212
##   black      14  39
##   Hispanic    10  30
##   Latino       6  20
##   middle eastern 92  96
```

```
##      mixed          3    5
##      Native Indian    0    3
##      Others           6   29
##      Pacifica         1    7
##      south asian     23   37
##      White European  84  250
```

```
round(prop.table(xtabs(~Ethnicity + Class.ASD.Traits.), 1), 2)
```

```
##              Class.ASD.Traits.
## Ethnicity      No  Yes
##   asian      0.29 0.71
##   black      0.26 0.74
##   Hispanic   0.25 0.75
##   Latino     0.23 0.77
##   middle eastern 0.49 0.51
##   mixed      0.38 0.62
##   Native Indian 0.00 1.00
##   Others      0.17 0.83
##   Pacifica    0.12 0.88
##   south asian  0.38 0.62
##   White European 0.25 0.75
```

```
xtabs(~Jaundice + Class.ASD.Traits.)
```

```
##              Class.ASD.Traits.
## Jaundice      No  Yes
##      no      253 513
##      yes      73 215
```

```
round(prop.table(xtabs(~Jaundice + Class.ASD.Traits.), 1), 2)
```

```
##              Class.ASD.Traits.
## Jaundice      No  Yes
##      no      0.33 0.67
##      yes     0.25 0.75
```

```
xtabs(~Family_mem_with_ASD + Class.ASD.Traits.)
```

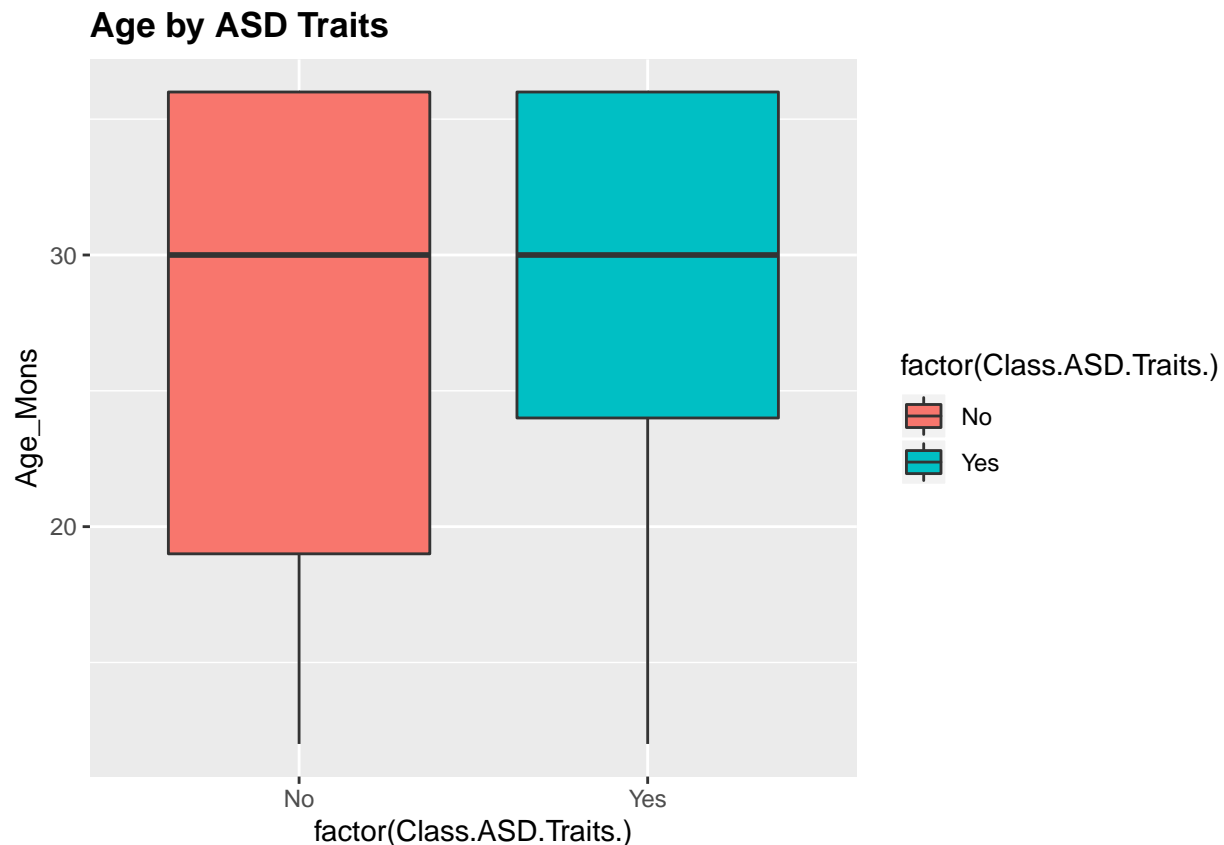
```
##              Class.ASD.Traits.
## Family_mem_with_ASD No  Yes
##      no      271 613
##      yes      55 115
```

```
round(prop.table(xtabs(~Family_mem_with_ASD + Class.ASD.Traits.),
1), 2)
```

```
##              Class.ASD.Traits.
## Family_mem_with_ASD No  Yes
##      no      0.31 0.69
##      yes     0.32 0.68
```

```
# Distribution of the Toddlers' Age by ASD Traits
```

```
ggplot(autism, aes(factor(Class.ASD.Traits.), Age_Mons)) + geom_boxplot(aes(fill = factor(Class.ASD.Traits.))) +
  ggtitle("Age by ASD Traits") + theme(plot.title = element_text(lineheight = 1,
face = "bold"))
```



Interactions between explanatory variables are needed when the effect of one explanatory variable on the probability of success depends on the value for another explanatory variable. From these graphs, interactions between some of the explanatory variables seem to be warranted.

In R, there are several ways to implement interaction terms in a logistic regression model:

- `formula = y ~ x1 + x2 + x1:x2`
- `formula = y ~ x1*x2`
- `formula = y ~ (x1 + x2)^2`

To include a quadratic term in a logistic regression model, use the following:

- `formula = y ~ x1 + I(x1^2)`

For factor variables, either turn them into factor variables and enter them into a logistic regression model, which is my preferred method, or use the `factor()` function inside a logistic regression: `formula = y ~ x1 + factor(x2)`, if `x2` needs to enter the regression as a factor variable.

Binary Logistic Regression Modeling

```
# Model 1 (Base Model)
mod.glm1 <- glm(Class.ASD.Traits. ~ Age_Mons + Sex + Ethnicity +
  Jaundice + Family_mem_with_ASD, family = "binomial", data = autism)
summary(mod.glm1)
```

```
##
## Call:
```

```
## glm(formula = Class.ASD.Traits. ~ Age_Mons + Sex + Ethnicity +
##      Jaundice + Family_mem_with_ASD, family = "binomial", data = autism)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.0941  -1.2292   0.7224   0.8317   1.4565
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      0.281885   0.304062   0.927  0.35389
## Age_Mons          0.008482   0.008859   0.957  0.33834
## Sexm              0.448503   0.147869   3.033  0.00242 **
## Ethnicityblack     0.177787   0.342323   0.519  0.60351
## EthnicityHispanic  0.247918   0.392078   0.632  0.52718
## EthnicityLatino    0.423043   0.488840   0.865  0.38682
## Ethnicitymiddle eastern -0.768519  0.202754  -3.790  0.00015 ***
## Ethnicitymixed     -0.330292   0.746778  -0.442  0.65828
## EthnicityNative Indian 13.396350 508.130189  0.026  0.97897
## EthnicityOthers     0.696988   0.471729   1.478  0.13954
## EthnicityPacifica   1.080882   1.079130   1.002  0.31653
## Ethnicitysouth asian -0.425479   0.299045  -1.423  0.15480
## EthnicityWhite European 0.207318   0.186398   1.112  0.26604
## Jaundiceyes         0.353792   0.163423   2.165  0.03040 *
## Family_mem_with_ASDyes -0.276408   0.187558  -1.474  0.14056
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 1303.9  on 1053  degrees of freedom
## Residual deviance: 1243.4  on 1039  degrees of freedom
## AIC: 1273.4
##
## Number of Fisher Scoring iterations: 13

# Model 2 (Model with both interaction and Non-linear effect)
# interaction of age and sex a quadratic term on age

# mod.glm2 <- glm(YOUR FORMULA HERE, family = 'binomial',
# data = autism) summary(mod.glm2)

# Display the models together stargazer(mod.glm1, mod.glm2,
# type = 'text')
```

Testing Model Differences

```
# CODE HERE (1 line)
```

Based on the test result, we will use `mod.glm2`.

Our model:

$$\log\left(\frac{\pi}{1-\pi}\right) = \beta_0 + \beta_1 \text{Age_Mons} + \beta_2 \text{Age_Mons}^2 + \beta_3 \text{Sexm} + \beta_4 \text{Ethnicityblack} + \beta_5 \text{EthnicityHispanic} + \beta_6 \text{EthnicityLatino} +$$

The odds ratio for an increase in age by c months is expressed in the following formula:

$$OR = \exp(c\beta_1 + c\beta_2(2 \times age + c))$$

which depends on the level of age.

Model Interpretation

We need some questions, such as

- What is the effect of being a 30-month old boy on the odds of having ADS traits?

```
c = 1
Age_Mons = 30
# YOUR CODE HERE (1 line)
```

- What is the effect of an one month increase in age (measured in months) of a 24 months old **female** toddler on the odds of having ADS traits? (Hint: use the formula above.)

```
c = 1
Age_Mons = 24
Sexm = 0
# YOUR CODE HERE (1 line)
```