

# MATH 4322 - Introduction to Data Science & Machine Learning

Exam 1 - A Solution  
Fall 2019

September 29, 2019

## Part 1(40 points)

**Problem 1.** (4 points) The response variable is *Accept* which is a **quantitative** variable. Thus this is a **regression** task with **linear** regression.

**Problem 2.** (4 points) The predictor that is qualitative is *Private*, we would use a **dummy variable** to represent it.

**Problem 3a.** (6 points) The  $p$ -value  $< 2.2e^{-16} \approx 0$ .

The percentage of variation is the *Multiple R – squared* 74.5%.

$H_0 : \beta_{PrivateYes} = \beta_{F.Undergrad} = \beta_{Grad.Rate} = \beta_{Top25perc} = 0$ . (Can use numbers instead).

**Problem 3b.** (4 points)

$$\hat{Accept} = \begin{cases} -2233 + 1269 + 0.6875 \times F.Undergrad + 14.78 \times Grad.Rate + 4.416 \times Top25perc, & \text{if Private University} \\ -2233 + 0.6875 \times F.Undergrad + 14.78 \times Grad.Rate + 4.416 \times Top25perc, & \text{if Public University} \end{cases}$$

**Problem 3c.** (4 points) Significant predictors are *Private* and *F.Undergrad*.

**Problem 3d.** (4 points)  $H_0 : \beta_{F.Undergrad} = 0$ , given  $\beta_{F.Undergrad}$  and  $\beta_{Private}$  and  $\beta_{Top25perc}$  are in the model.

**Problem 3e.** (8 points)  $\hat{\beta}_{Private} = 1269$  that implies that the number of accepted applications will increase on average by 1,269 if the university is Private.

$\hat{\beta}_{F.Undergrad} = 0.6875$  For each additional full time undergrad, the number of accepted applications will increase on average by 0.6875.

**Problem 3f.** (6 points) **Confidence Interval:** The universities that have a 65% graduation rate and 15,000 full time undergrads will accept on average between 8,433 and 10,343 applications.

**Prediction Interval:** For one university that has a 65% graduation rate and 15,000 full time undergrads is predicted to accept between 5,634 and 13,142 applications.

The **prediction interval** is wider.

## Part 2 (30 Points)

**Problem 1.** (9 points) The response is *student(yes/no)* thus this is a **classification** task. We will use **logistic regression**.

**Problem 2.** (8 points)

$$\left( \log \frac{p(\hat{x})}{1 - p(\hat{x})} \right) = -11.65 - 0.6095 \times student + 0.00525 \times balance + 0.0000535 \times income$$

or

$$p(\hat{x}) = \frac{\exp(-11.65 - 0.6095 \times student + 0.00525 \times balance + 0.0000535 \times income)}{1 + \exp(-11.65 - 0.6095 \times student + 0.00525 \times balance + 0.0000535 \times income)}$$

Where

$$student = \begin{cases} 0, & \text{if no} \\ 1, & \text{if yes} \end{cases}$$

**Problem 3.** (4 points) *balance* is significant, (if they also say *income* that is fine).

**Problem 4.** (5 points) As the *balance* increases, the chance of a default increases.

As the *income* increases, the chance of a default increases. (Again this is only needed if they say that income is significant).

**Problem 5.** (4 points) The **chance** or **probability** of a person defaulting on the credit card is at 1.2% for a customer that has a balance of \$1,000.

**Part 3 (30 Points + 10 Bonus)**

**Problem 1.** (20 points)

- Linear relationship
- The expected values of the errors (residuals) are zero.
- Constant variance of the errors (residuals).
- Error have a Normal distribution for each value of  $x$ .

**Problem 2.** (Only 2 are necessary for points)(2 bonus points)

- Scatterplot or "pairs" to see linearity
- Residual plot to see constant variance, linearity and outliers
- Q-Q plot to see Normality

**Problem 3a.** (2 points) This is not a linear relationship.

**Problem 3b.** (8 points) This would be a polynomial regression. **Cubic**

$$y = \beta_0 + \beta_1x + \beta_2x^2 + \beta_3x^3 + \epsilon, \quad \epsilon \sim N(0, \sigma^2)$$

**Problem 3c.** (4 bonus points) If we have too many degrees for a polynomial regression, we will have a problem of **overfitting** the training data. Then the results will not work best for any other test data that we would use from the same population.

**Problem 4.** (Only need 2 for points)(4 bonus points)

- $C_p$  - small
- AIC - small
- BIC - small
- adjusted  $R^2$  - large (it has to say adjusted).