**MSDS 6306: Doing Data Science**

**Case Study 1**

**Due: Sunday, October 27th 11:59pm CST**

**Description**: DDSAnalytics is an analytics company that specializes in talent management solutions for Fortune 100 companies. Talent management is defined as the iterative process of developing and retaining employees. It may include workforce planning, employee training programs, identifying high-potential employees and reducing/preventing voluntary employee turnover (attrition). To gain a competitive edge over its competition, DDSAnalytics is planning to leverage data science for talent management. The executive leadership has identified predicting employee turnover as its first application of data science for talent management. Before the business green lights the project, they have tasked your data science team to conduct an analysis of existing employee data.

You have been given a dataset (**CaseStudy1-data.csv**)to do a data analysis to identify factors that lead to attrition. You should identify the top three factors that contribute to turnover (backed up by evidence provided by analysis). There may or may not be a need to create derived attributes/variables/features. The business is also interested in learning about any job role specific trends that may exist in the data set (e.g., “Data Scientists have the highest job satisfaction”). You can also provide any other interesting trends and observations from your analysis. The analysis should be backed up by robust experimentation and appropriate visualization. Experiments and analysis must be conducted in R. You will also be asked to build a model to predict attrition.

This is an individual project.

**Deliverables:**

**UNIT 8 and 9 Live Sessions:**

The due date for videoed submission is Sunday October 27th 11:59pm CST (Week 9). We will meet for Live Session 8 at the beginning for a DSNOW and a discussion of methods of handling imbalanced classification data.

We will also, meet for Live Session 9. I will answer any questions about the project that develop by that time and we will have a DSNOW!

**Further Details:**

One of the final deliverables is a YouTube video in which you present your findings. You will need to record and upload to YouTube a **7-minute** presentation or provide the link to your Zoom recording. You can assume that your audience is the CEO and CFO of Frito Lay (your client). It is a diverse audience; the CEO is a statistician and the CFO has had only one class in statistics. They have indicated that you cannot take more than 7 minutes. 20% of your grade will be based on the presentation. The goal is to communicate the findings of the project in a clear, concise and scientific manner. Finally, include the link in your RMarkdown file. Finally, finally make sure to put the link to the YouTube / Zoom video in the Google Doc. The links will be available for a week at which time you may take your video off of YouTube / Zoom if you wish. Please make sure and check out at least 3 of your peer’s presentations! Make this your MASTERPIECE! ☺

GOOGLE DOC:

<https://docs.google.com/document/d/16McabKpyhpx0U-bUsEmJqWny-SQ0Bc3kgqG_sC5j3_E/edit?usp=sharing>

I provided an additional data set of 300 observations that do not have the labels (attrition or not attrition). We will refer to this data set as the “Competition Set” and is in the file “**CaseStudy1CompSet No Attrition.csv**”. I have the real labels and will thus assess the accuracy rate of your best classification model. 10% of your grade will depend on the sensitivity and specificity rate of your “best” classification model for identifying attrition. You must provide a model that will attain at least 60% sensitivity and specificity (60 each = 120 total) for the training and the validation set. Therefore, you must provide the labels (ordered by ID) in a csv file. Please include this in your GitHub repository and call the file **“Case1PredictionsXXXX Attrition.csv”.** XXXX is your last name. (Example: Case1PredictionsSadler Attrition.csv” would be mine.) An example submission file can be found on Github in the Unit 8 and 9 folder: **Case1PredictionsClassifyEXAMPLE.csv**.

Create a GitHub repository named **CaseStudy1DDS** with a RMarkdown file containing an executive summary (in the Readme.md), introduction to the project, all supporting code and analysis, and the slides for the presentation. The repository should also include your prediction csv file and don’t forget to put the link to the YouTube / Zoom video in the RMarkdown file. Submit a link to the GitHub repository via the space provided for the Case Study 01 page in Canvas. Finally, make sure and put the link to the YouTube / Zoom video on the Google Doc.

**Due Dates:**

Sunday, October 27th 11:59pm CST: Rmd, Powerpoint, and Final videoed submission due.

**BONUS:**

The individual with the highest sensitivity + specificity (both at least 60%) on the classification validation set will win the Bonus: 3 extra points and bragging rights!

**Rubric:**

10% RMarkdown File

30% Final Video Presentation and analysis(15% slide content, 15% presentation)

**Minimal Stumbles / mis statements / etc.** if you trip up more than a couple of times, reshoot the video.It

will be much better with the practice!

**Labeled Plots with readable labels**

**7-minute time limit**

**Voice inflection**

**Creativity**

**Robust analysis** – this means supporting your findings with plots, charts, confusion matrices, sensitivity and specificity, F1, and sound logic that ties your points together. An example of this “logic” is to evaluate any differences between your model making a false positive and a false negative.

40% Analysis

Correct interpretation

Appropriate analysis (tests, methods, descriptions)

20% Validation Requirement for Attrition(Sensitivity > 60% and Specificity > 60%)

**FAQ and Comments:**

**1. Question: In the dataset, what does Relationship Satisfaction mean...(relationship to manager, to peers)**

Relationship satisfaction with manager.

2. Advice: Don't eliminate variables simply because they have a high correlation with one another.  This is an indication that they do share some information although the information they don't share may be correlated with the response individually.

3. Advice: When plotting and exploring attrition, the percentage of those who left is probably more useful than the count.

4. Question: Is the dataset, is the distance from home in miles or kilos?

We don't have that information (however we do know whether its high or low)

5.  **Question: In the dataset: what is the definition of pay rates: Hourly, Daily & Monthly.  These values to not seem to relate to each other or the Monthly Salary (which is different than Monthly Rate).**

We don't have that information (however we do know whether they are high or low). They may or may not relate to each other or the monthly salary (this is for the student to infer and decide whether theres any correlation or whether this is a useful feature for attrition)

6**. Question: In the dataset: we do see that Job Levels go from 1-5 and assume that 1 may symbolize a lower level employee, but this is not defined.  Though this level does have evidence of a positive linear relationship with Monthly Income, it does not seem to correlate well with the Job Titles. in other words someone with a Director can be a 2-5, and manager a 3-5.**

Yes we can assume 1 is a lower job level than 5.

**7. Question: In the dataset, does overtime mean Hourly vs. Salaried worker?**

We can assume that people with overtime are non-exempt / hourly employees.

**8 Question: In the dataset, Performance Ratings are only 3 & 4, is there a mistake?  Unless a corrupted system, hard to imagine ratings consistently high, even as 2 still means "good".**

**It is self-reported data, think about why the employees may only answer 3 and 4**

No this is the only data we have, there is no mistake.

**9 Question: In the dataset, does Training times mean: hours, weeks, or instances and over what period?**

Training times last year means number of training sessions attended by the employee.

10. Question: Do we have any information on the other columns?

Yes…. Here is the break down for a few columns. Other columns don’t have any additional information but may be useful in predicting attrition. This is often the case and can be a feature rather than a bug in that it may prevent biased decisions. You may plausibly speculate on their meaning (just make sure it is clear that it is speculation) or you can simply report that no information was given and report what you are able to about the relationship and what it might mean. Here is some additional information on a few of the columns:

Education  
1 'Below College'  
2 'College'  
3 'Bachelor'  
4 'Master'  
5 'Doctor'

EnvironmentSatisfaction  
1 'Low'  
2 'Medium'  
3 'High'  
4 'Very High'

JobInvolvement  
1 'Low'  
2 'Medium'  
3 'High'  
4 'Very High'

JobSatisfaction  
1 'Low'  
2 'Medium'  
3 'High'  
4 'Very High'

PerformanceRating  
1 'Low'  
2 'Good'  
3 'Excellent'  
4 'Outstanding'

RelationshipSatisfaction  
1 'Low'  
2 'Medium'  
3 'High'  
4 'Very High'

WorkLifeBalance  
1 'Bad'  
2 'Good'  
3 'Better'  
4 'Best'