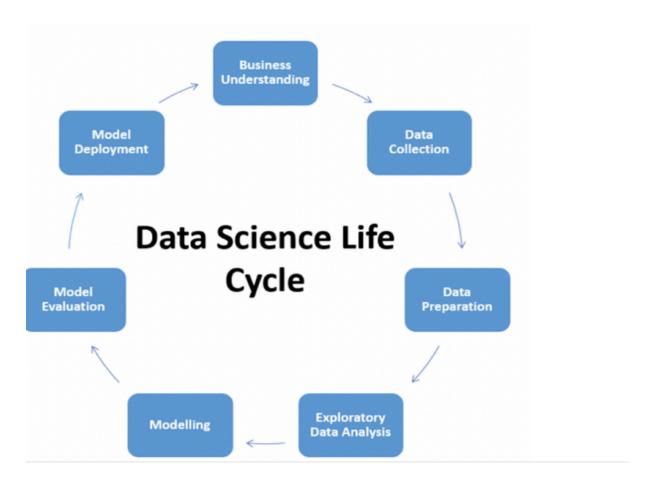
Today was my first day as an intern, and I learned many things from my colleague, Rachit Rijal. First, we discussed which project would be feasible for us. We decided to focus on data wrangling. Together, Rachit and I started searching for datasets on Kaggle and found a graduation-related dataset.

I have also gained a clear understanding of the data lifecycle.



I start to do data wrangling work.

First of all i start to find the input variable of my data: input Variables

- 1. Serial No. Unique row ID
- 2. GRE Scores Out of 340
- 3. TOEFL Scores Out of 120
- 4. University Rating Out of 5
- 5. Statement of Purpose Strength Out of 5
- 6. Letter of Recommendation Strength Out of 5
- 7. Undergraduate GPA Out of 10

Data collection and import data in google colab:

```
df = pd.read_csv('/content/filename.csv')
```

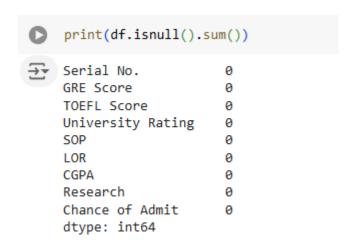
8. Resea rch

Experience - Either 0 or 1

We remove the shape and comma error from the column

```
[ ] df.columns = df.columns.str.strip()
```

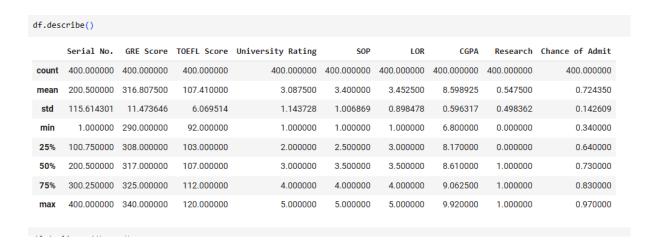
We try to find out the missing value in our data set



Descriptive Statistics

Descriptive statistics involve a set of summary measures that provide a snapshot of the dataset's characteristics. These measures help us understand the distribution, central tendency, and variability within the data.

- Mean: The average value of the data.
- Median: The middle value when the data is sorted.
- Mode: The most frequently occurring value.
- Range: The difference between the maximum and minimum values.
- Standard Deviation: A more interpretable measure of data spread. These statistics provide a preliminary understanding of the dataset, which is valuable for subsequent analysis and decision-making.



Remove any null value in data set



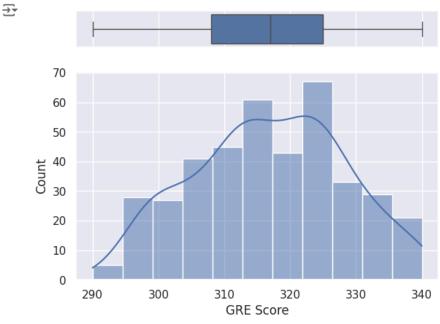
Data Wrangling

- Data Inspection
 - Checking Duplicate Entries
 - Checking Missing Values
 - Checking standard format
 - Checking data entry typos and errors
- Data Cleaning
 - Removing Duplicates
 - Handling Missing Values
 - Standardising Formats
 - Correcting Errors
- Data Transformation
 - Feature Engineering
 - Normalisation/Scaling
 - One-Hot Encoding
- Data Integration
- Data Reduction
- Data Formatting
- Data Enrichment
- Data Validation
- Documentation
- Exploratory Data Analysis (EDA)

I have learn this through this session in data wrangling process:

After that I have learn the univarient analysis in data

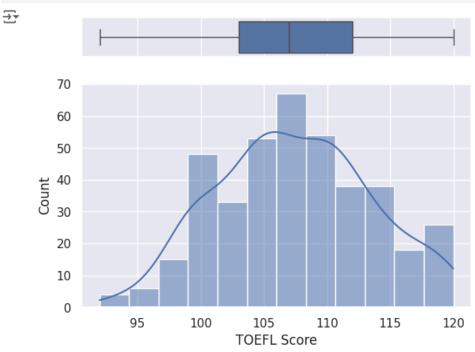
```
[ ] sns.set(style="darkgrid")
   f, (ax_box, ax_hist) = plt.subplots(2, sharex=True, gridspec_kw={"height_ratios": (.15, .85)})
   sns.boxplot(data=df, x='GRE Score', ax=ax_box,)
   sns.histplot(data=df, x="GRE Score", ax=ax_hist,kde=True)
   ax_box.set(xlabel='')
   plt.show()
```



From the above graph, we can say -

- Average GRE score is around 318.
- 25% of the score is less than 310.
- 75% of the score is less than 325.

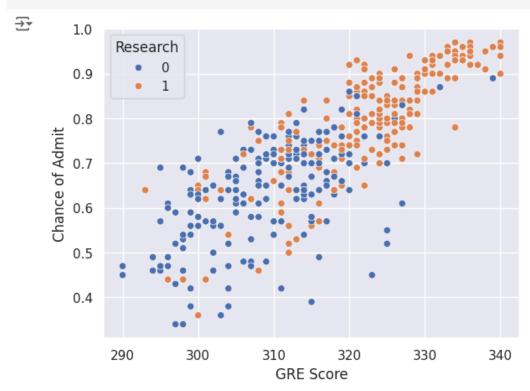
```
sns.set(style="darkgrid")
f, (ax_box, ax_hist) = plt.subplots(2, sharex=True, gridspec_kw={"height_ratios": (.15, .85)})
sns.boxplot(data=df,x='TOEFL Score', ax=ax_box)
sns.histplot(data=df, x="TOEFL Score", ax=ax_hist,kde=True)
ax_box.set(xlabel='')
plt.show()
```



From the above graph, we can say

- Average TOEFL score is around 106.
- 25% of the score is less than 104.
- 75% of the score is less than 112

```
[ ] sns.scatterplot(x='GRE Score',y='Chance of Admit',data=df,hue='Research')
plt.show()
```



From this scatter plot we can analysis:

From the above graph we can say that GRE score and Chance of admit has a linear relationship.

Remaining other work is done by my friend Richit Rijal.

Now we have to do many more things in this project. We will continue our project from here.