# 1.02. univariate\_analaysis

October 14, 2024

# 1 EDA - Univariate Analysis

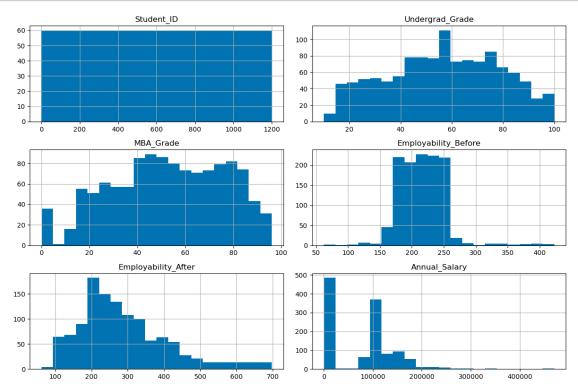
```
[1]: import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     import seaborn as sns
     import httpimport
     from pathlib import Path
[2]: # Import personal library
     with httpimport.github_repo("junclemente", "jcds", ref="master"):
         import jcds.eda as jq
[3]: # Import clean dataset
     datasets = Path("../datasets")
     df = pd.read_csv(datasets / "school_clean.csv")
     df.head()
[3]:
        Student_ID Undergrad_Degree Undergrad_Grade MBA_Grade Work_Experience
                            Business
                                                  68.4
                                                             90.2
                 2
     1
                            Business
                                                  62.1
                                                             92.8
                                                                               No
     2
                 3 Computer Science
                                                  70.2
                                                             68.7
                                                                              Yes
                 4
     3
                         Engineering
                                                  75.1
                                                             80.7
                                                                               No
     4
                 5
                                                             74.9
                             Finance
                                                  60.9
                                                                                No
        Employability_Before Employability_After
                                                        Status
                                                                Annual_Salary
     0
                       252.0
                                             276.0
                                                        Placed
                                                                     111000.0
                       423.0
                                             410.0 Not Placed
     1
                                                                          0.0
     2
                       101.0
                                             119.0
                                                        Placed
                                                                     107000.0
     3
                       288.0
                                             334.0 Not Placed
                                                                          0.0
                                             252.0 Not Placed
     4
                       248.0
                                                                          0.0
```

### 1.1 Distributions

# 1.1.1 Continuous Features

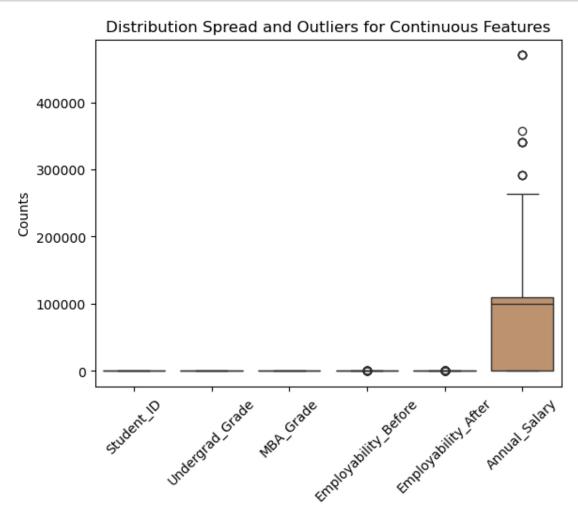
```
[4]: # Set global color palette
global_color = "colorblind"
# Set global color for Seaborn
sns.set_palette(global_color)
# Set global color palette for Matplotlib
colors = sns.color_palette(global_color)
plt.rcParams["axes.prop_cycle"] = plt.cycler(color=colors)
```

```
[5]: # Show histogram of each continuous variable
    df.hist(bins=20, figsize=(12, 8))
    plt.tight_layout()
    plt.show()
```

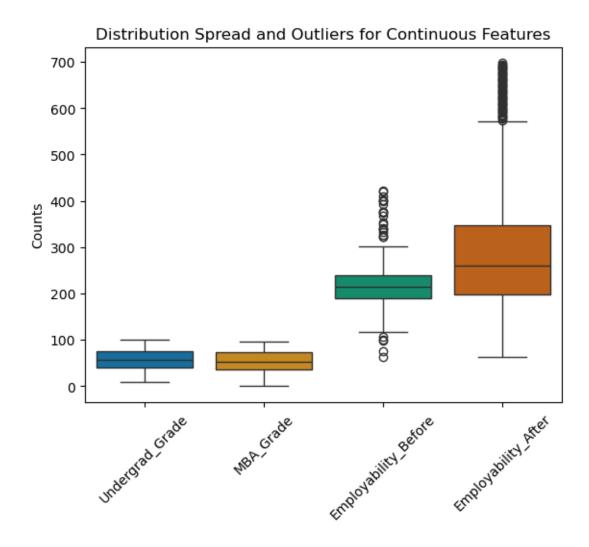


- Undergrad Grade shows a normal distribution.
- MBA Grade has a somewhat normal distribution but values at 0 may be outliers. This could also be related to students who dropped out or did not place after graduation.
- Employability variables are both skewed to the left but the range in "After" has increased.
- Annual Salary shows a left skew with most students that placed at around \$100K. The 0s are all graduates who did not place.

```
[6]: # Visualize outliers
sns.boxplot(data=df)
plt.title("Distribution Spread and Outliers for Continuous Features")
plt.ylabel("Counts")
plt.xticks(rotation=45)
plt.show()
```



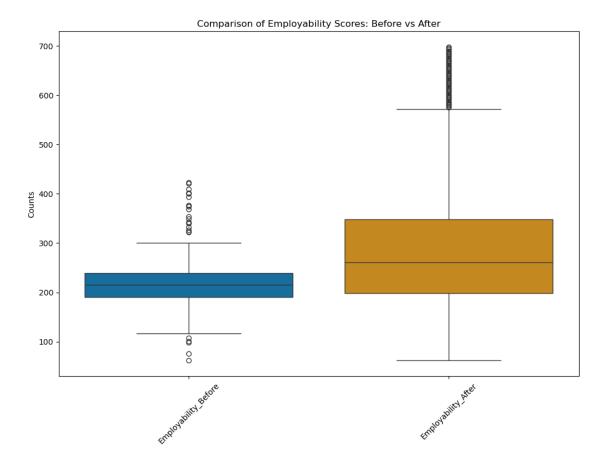
```
[7]: # Remove student ID and annual salary to show detail of other columns sns.boxplot(data=df.drop(columns=["Student_ID", "Annual_Salary"])) plt.title("Distribution Spread and Outliers for Continuous Features") plt.ylabel("Counts") plt.xticks(rotation=45) plt.show()
```



The range increase in Employability After compared to Before is evident showing that the program has a positive impact on students' career.

There does not appear to be any significant differences between the Undegrad and MBA grade.

```
[12]: # Comparison of Employability before and after
plt.figure(figsize=(12, 8))
sns.boxplot(data=df[["Employability_Before", "Employability_After"]])
plt.title("Comparison of Employability Scores: Before vs After")
plt.ylabel("Counts")
plt.xticks(rotation=45)
plt.show()
```

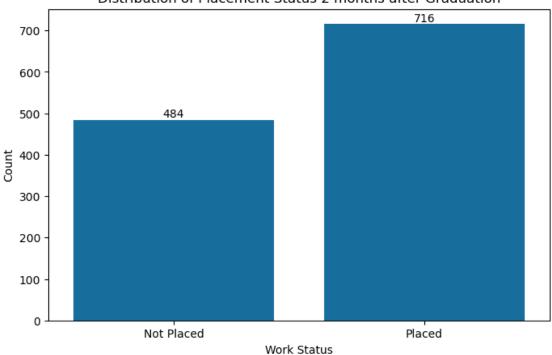


A wider range of scores post-graduation shows an improvement in employability. The program has a positive impact on students' career.

# 1.1.2 Categorical Features

```
Job Placement Status
[13]: | # categorical_vars = ["Undergrad_Degree", "Work_Experience", "Status"]
      # Status
      plt.figure(figsize=(8, 5))
      ax = sns.countplot(x="Status", data=df, order=["Not Placed", "Placed"])
      for container in ax.containers:
          ax.bar_label(container)
      plt.title("Distribution of Placement Status 2 months after Graduation")
      plt.xlabel("Work Status")
      plt.ylabel("Count")
      plt.show()
```





```
[17]: status_counts = df["Status"].value_counts()
    percentages = (status_counts / len(df)) * 100

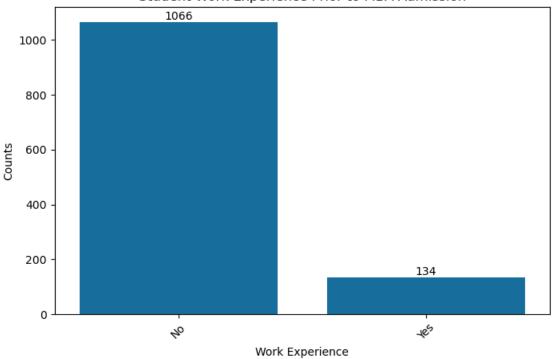
    print(f"Total Placed: {status_counts['Placed']} ({percentages['Placed']:.1f}%)")
    print(
        f"Total Not Placed: {status_counts['Not Placed']} ({percentages['Not_\sum_\supersetting \text{Placed'}]:.1f}%)"
        )
}
```

Total Placed: 716 (59.7%)
Total Not Placed: 484 (40.3%)

# Work Experience

```
[22]: # Work Experience
plt.figure(figsize=(8, 5))
ax = sns.countplot(x="Work_Experience", data=df)
for container in ax.containers:
        ax.bar_label(container)
plt.title(f"Student Work Experience Prior to MBA Admission")
plt.xlabel("Work Experience")
plt.ylabel("Counts")
plt.ylabel("Counts")
plt.xticks(rotation=45)
plt.show()
```





```
[19]: status_counts = df["Work_Experience"].value_counts()
    percentages = (status_counts / len(df)) * 100

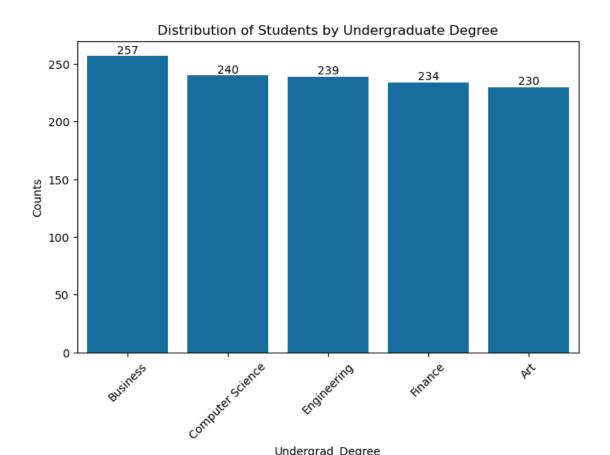
print(f"Total Placed: {status_counts['Yes']} ({percentages['Yes']:.1f}%)")
    print(f"Total Not Placed: {status_counts['No']} ({percentages['No']:.1f}%)")
```

Total Placed: 134 (11.2%)
Total Not Placed: 1066 (88.8%)

Most MBA students did not have prior work experience before entering the program.

# Undergraduate Degree

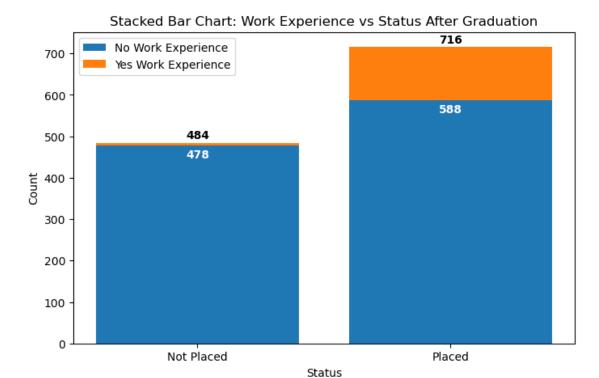
```
[23]: # Undergrad Degree
plt.figure(figsize=(8, 5))
ax = sns.countplot(x="Undergrad_Degree", data=df)
for container in ax.containers:
        ax.bar_label(container)
plt.title(f"Distribution of Students by Undergraduate Degree")
# plt.xlabel("Undergraduate Degree")
plt.ylabel("Counts")
plt.xticks(rotation=45)
plt.show()
```



Undergrad\_Degree

```
[24]: crosstab = pd.crosstab(df["Status"], df["Work_Experience"])
      display(crosstab)
     Work_Experience
                       No
                           Yes
     Status
     Not Placed
                      478
                              6
     Placed
                      588
                           128
[25]: plt.figure(figsize=(8, 5))
      # Plot the bars manually for 'No' and 'Yes' work experience
      bars1 = plt.bar(
          crosstab.index, crosstab["No"], label="No Work Experience", color="#1f77b4"
      bars2 = plt.bar(
          crosstab.index,
          crosstab["Yes"],
          bottom=crosstab["No"],
          label="Yes Work Experience",
```

```
color="#ff7f0e",
)
# Add number counts on top of each bar for 'No Work Experience'
for bar in bars1:
    height = bar.get_height()
    plt.text(
        bar.get_x() + bar.get_width() / 2.0,
        height - 30,
        "%d" % int(height),
        ha="center",
        color="white",
        fontweight="bold",
    )
# Add number counts on top of each bar for 'Yes Work Experience'
for bar in bars2:
    height = bar.get_height() + bar.get_y()
    plt.text(
        bar.get_x() + bar.get_width() / 2.0,
        height + 10,
        "%d" % int(height),
        ha="center",
        color="black",
        fontweight="bold",
    )
# Add labels and title
plt.xlabel("Status")
plt.ylabel("Count")
plt.title("Stacked Bar Chart: Work Experience vs Status After Graduation")
plt.legend()
# Show the plot
plt.show()
```



Of all 716 students who found a job within 2 months of graduation, 588 students did not have work experience. Of the 484 students who did not find a job within 2 months of graduation, 478 did not have work experience. Six students who did not place had prior work experience.