```
#Importing the Libraries
In [67]:
          import numpy as np
          import pandas as pd
          import os
          import plotly.express as px
          import plotly.graph_objects as go
          from plotly.subplots import make_subplots
          import plotly.express as px
          from sklearn import preprocessing
          import seaborn as sns
          import matplotlib.pyplot as plt
          import math
          from sklearn.ensemble import RandomForestClassifier
          from sklearn.metrics import accuracy_score
          from sklearn.neighbors import KNeighborsClassifier
          from sklearn import metrics
          from sklearn.model_selection import train_test_split
          from sklearn.model_selection import cross_val_score
          from sklearn import preprocessing
          from sklearn import svm
          from sklearn.svm import SVC
 In [ ]: #Data Collection and Preparation
          #Read The Data Set
         df=pd.read csv("E:\\NMDS\collegePlace.csv")
         df.head()
 In [3]: df['Stream'].value_counts()
         Computer Science
                                           776
 Out[3]:
         Information Technology
                                           691
         Electronics And Communication
                                           424
         Mechanical
                                           424
         Electrical
                                           334
         Civil
                                           317
         Name: Stream, dtype: int64
         #Handling Mising Values
In [10]:
         df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 2966 entries, 0 to 2965
         Data columns (total 8 columns):
              Column
                                 Non-Null Count Dtype
          ---
              ----
                                  -----
          0
              Age
                                 2966 non-null
                                                 int64
          1
              Gender
                                 2966 non-null object
              Stream
                                 2966 non-null object
          2
          3
              Internships
                                 2966 non-null
                                                 int64
          4
              CGPA
                                 2966 non-null
                                                 int64
          5
              Hostel
                                 2966 non-null
                                                 int64
          6
              HistoryOfBacklogs 2966 non-null
                                                 int64
              PlacedOrNot
                                 2966 non-null
                                                 int64
         dtypes: int64(6), object(2)
         memory usage: 162.3+ KB
         #Handling Mising Values
 In [2]:
```

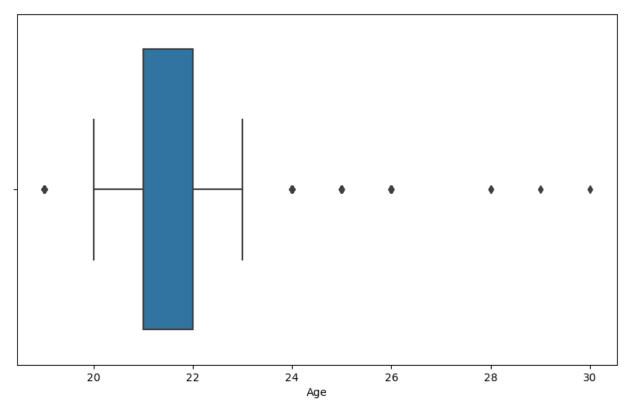
4/21/23, 5:58 AM

```
CollegePlace
          df.isnull().sum()
                                  0
          Age
 Out[2]:
          Gender
                                  0
          Stream
                                  0
          Internships
                                  0
          CGPA
                                  0
          Hostel
                                  0
          HistoryOfBacklogs
                                  0
          PlacedOrNot
                                  0
          dtype: int64
          #Handling Categorical Values
In [13]:
          df.describe()
                                                CGPA
                                                           Hostel HistoryOfBacklogs
                                                                                     PlacedOrNot
Out[13]:
                        Age
                              Internships
          count 2966.000000
                              2966.000000
                                          2966.000000
                                                      2966.000000
                                                                         2966.000000
                                                                                      2966.000000
           mean
                   21.485840
                                 0.703641
                                             7.073837
                                                          0.269049
                                                                            0.192178
                                                                                         0.552596
             std
                    1.324933
                                 0.740197
                                             0.967748
                                                          0.443540
                                                                            0.394079
                                                                                         0.497310
                                 0.000000
                                             5.000000
                                                          0.000000
                                                                            0.000000
                                                                                         0.000000
            min
                   19.000000
            25%
                   21.000000
                                 0.000000
                                             6.000000
                                                          0.000000
                                                                            0.000000
                                                                                         0.000000
            50%
                   21.000000
                                 1.000000
                                             7.000000
                                                          0.000000
                                                                            0.000000
                                                                                          1.000000
            75%
                   22.000000
                                 1.000000
                                             8.000000
                                                          1.000000
                                                                            0.000000
                                                                                          1.000000
                   30.000000
                                 3.000000
                                             9.000000
                                                          1.000000
                                                                            1.000000
                                                                                          1.000000
            max
          #Handling Outliers
In [12]:
          def transformationplot(feature):
               plt.figure(figsize=(12,5))
               plt.subplot(1,2,1)
               sns.displot(feature)
               transformationplot(np.log(df['Age']))
               plt.show()
          # I tried all the columns and find out that only age column has some outliers.
In [14]:
          plt.figure(figsize = (10, 6), dpi = 100)
```

sns.boxplot(x = "Age", data = df)

<AxesSubplot:xlabel='Age'>

Out[14]:



```
print(max_thresold)

min_thresold = df['Age'].quantile(0.01)
print(min_thresold)

df = df[(df['Age']<max_thresold) & (df['Age']>min_thresold)]

24.0
19.0

In [16]: #HandLing Categorical values
#Lable Encoding
df=df.replace(['Male'],[0])
df=df.replace(['Female'],[1])
df=df.replace(['Computer Science','Information Technology','Electronics And Communicat
df.drop(['Hostel'],axis=1)
```

In [15]:

#Removing Outlier

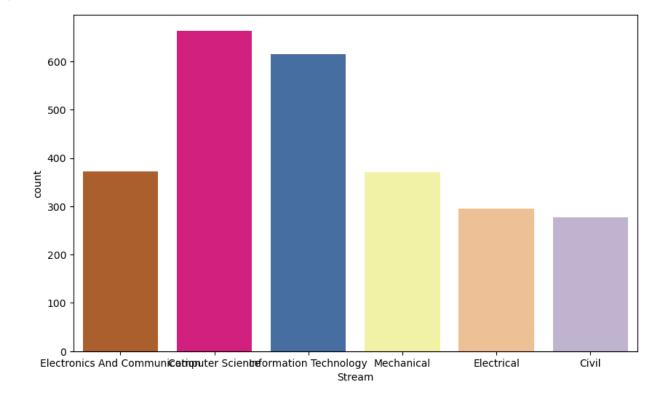
max\_thresold = df['Age'].quantile(0.95)

Out[16]:		Age	Gender	Stream	Internships	CGPA	HistoryOfBacklogs	PlacedOrNot
	0	22	0	2	1	8	1	1
	1	21	1	0	0	7	1	1
	2	22	1	1	1	6	0	1
	3	21	0	1	0	8	1	1
	4	22	0	4	0	8	0	1
	•••							
	2961	23	0	1	0	7	0	0
	2962	23	0	4	1	7	0	0
	2963	22	0	1	1	7	0	0
	2964	22	0	0	1	7	0	0
	2965	23	0	5	0	8	0	1

2595 rows × 7 columns

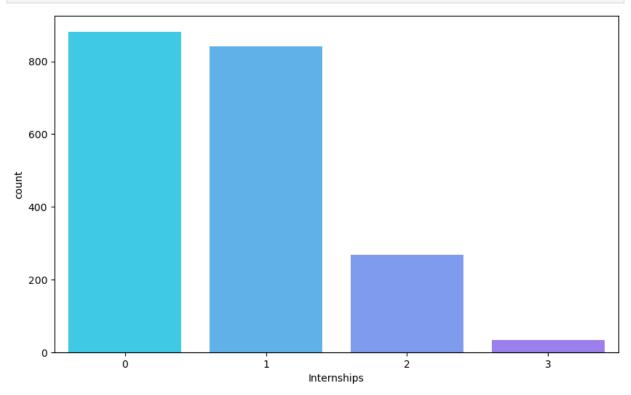
```
In [25]: #Exploratory Data Abnliysis
    #Data Visualization count of stream
    plt.figure(figsize = (10, 6), dpi = 100)
    color_palette = sns.color_palette("Accent_r")
    sns.set_palette(color_palette)
    sns.countplot(x = "Stream", data = df)
```

Out[25]: <Axes: xlabel='Stream', ylabel='count'>



```
In [27]: #Data Visualization count of Internships
plt.figure(figsize = (10, 6), dpi = 100)
```

```
color_palette = sns.color_palette("cool")
sns.set_palette(color_palette)
sns.countplot(x = "Internships", data = df)
plt.show()
```



```
In [5]: #Data Visualization Distribution of CGPA
plt.figure(figsize = (10, 6), dpi = 100)
grp = dict(df.groupby('CGPA').groups)

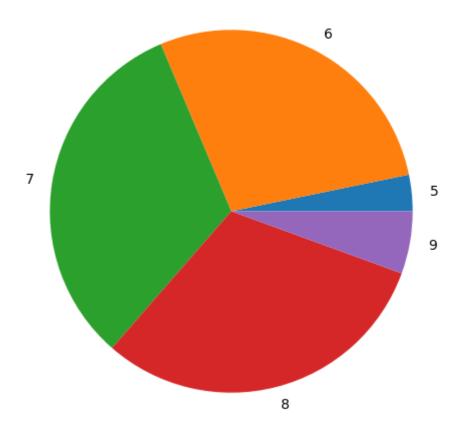
m = {}

for key, val in grp.items():
    if key in m:
        m[key] += len(val)

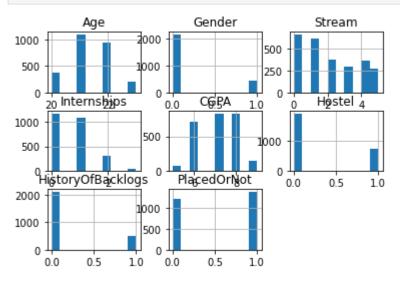
    else:
        m[key] = len(val)

plt.title("Distribution of CGPA")
plt.pie(m.values(), labels = m.keys())
plt.show()
```

## Distribution of CGPA

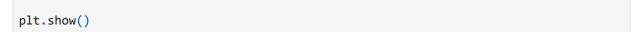


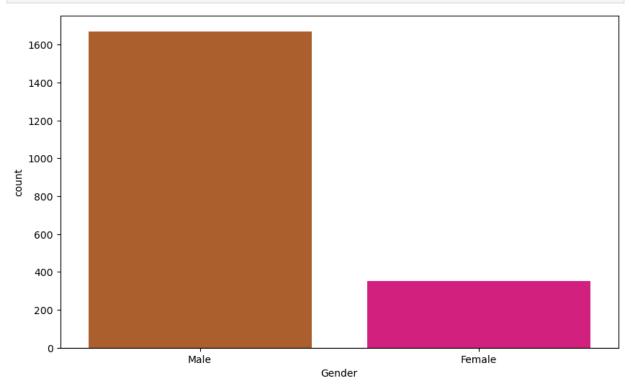
```
In [24]: #Univariate Analysis
    df.hist()
    plt.show()
```



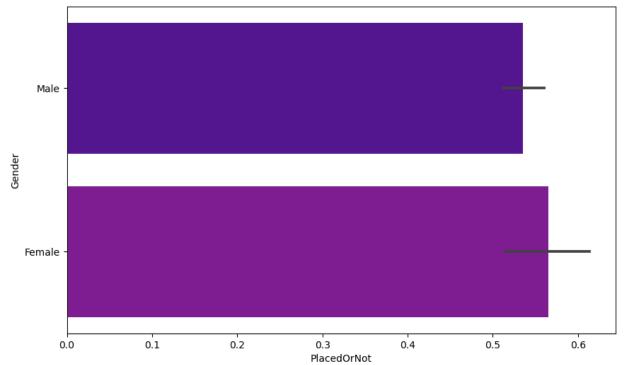
```
In [31]: plt.figure(figsize = (10, 6), dpi = 100)
# setting the different color palette
color_palette = sns.color_palette("Accent_r")
sns.set_palette(color_palette)

sns.countplot(x = "Gender", data = df)
```

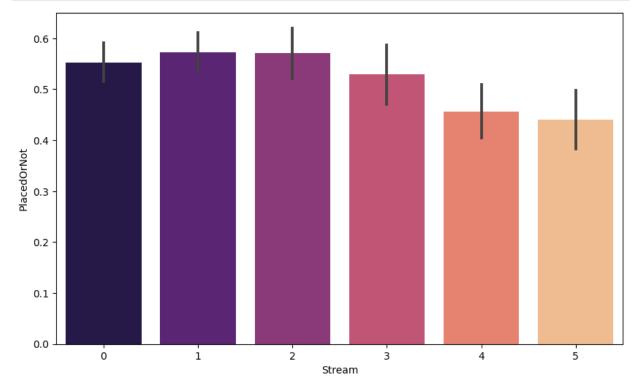








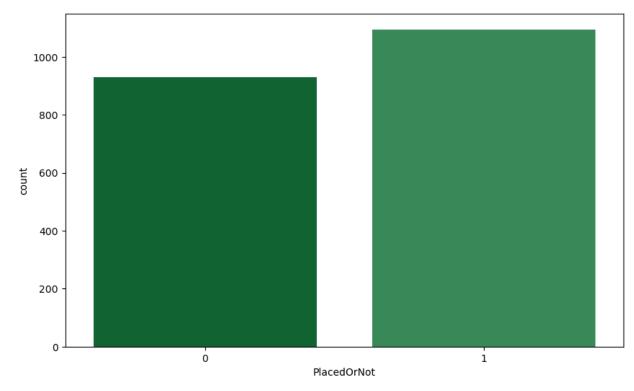
```
In [33]: #Multivariate Analysis
  plt.figure(figsize = (10, 6), dpi = 100)
  # setting the different color palette
  color_palette = sns.color_palette("magma")
  sns.set_palette(color_palette)
  sns.barplot(x = "Stream", y = "PlacedOrNot", data = df)
  plt.show()
```



```
In [35]: # How many placed
plt.figure(figsize = (10, 6), dpi = 100)

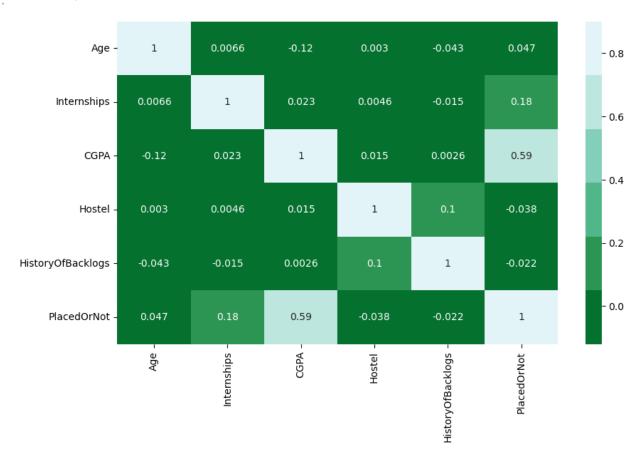
# setting the different color palette
color_palette = sns.color_palette("BuGn_r")
sns.set_palette(color_palette)

sns.countplot(x = "PlacedOrNot", data = df)
plt.show()
```



```
In [6]: #Scaling the Data
#Correllation
plt.figure(figsize = (10, 6), dpi = 100)
color = sns.color_palette("BuGn_r")
sns.heatmap(df.corr(), vmax=0.9, annot=True,cmap = color)
```

Out[6]: <AxesSubplot:>

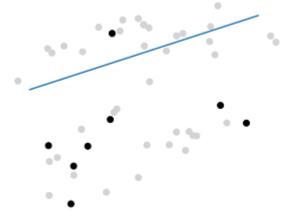


```
#Model Building
 In [8]:
         #mode1:SVM
          classifier=svm.SVC(kernel='linear')
In [18]: #mode1:SVM
         import numpy as np
         from sklearn.datasets import make classification
         from sklearn import svm
         from sklearn.model selection import train test split
          classes = 4
         X,t= make_classification(100, 5, n_classes = classes, random_state= 40, n_informative
         X_train, X_test, y_train, y_test= train_test_split(X, t , test_size=0.50)
         model = svm.SVC(kernel = 'linear', random state = 0, C=1.0)
         model.fit(X_train, y_train)
         #%%
         y=model.predict(X test)
         y2=model.predict(X train)
         from sklearn.metrics import accuracy score
          score =accuracy score(y, y test)
          print(score)
          score2 =accuracy_score(y2, y_train)
          print(score2)
         #%%
          import matplotlib.pyplot as plt
         color = ['black' if c == 0 else 'lightgrey' for c in y]
         plt.scatter(X_train[:,0], X_train[:,1], c=color)
         # Create the hyperplane
         w = model.coef [0]
         a = -w[0] / w[1]
         xx = np.linspace(-2.5, 2.5)
         yy = a * xx - (model.intercept [0]) / w[1]
         # Plot the hyperplane
```

0.880.96

plt.plot(xx, yy)

plt.axis("off"), plt.show();



```
# Spot-Check Algorithms
In [43]:
         models = []
          #models.append(('LR', LogisticRegression()))
          #models.append(('LDA', LinearDiscriminantAnalysis()))
          models.append(('KNN', KNeighborsClassifier()))
          #models.append(('CART', DecisionTreeClassifier()))
          #models.append(('NB', GaussianNB()))
          models.append(('SVM', SVC()))
          # evaluate each model in turn
          results = []
          names = []
          for name, model in models:
            # kfold = KFold(n splits=10, random state=seed)
              cv_results = cross_val_score(model, X_train, y_train, scoring='accuracy')
          results.append(cv results)
          names.append(name)
          msg = "%s: %f (%f)" % (name, cv results.mean(), cv results.std())
          print(msg)
         SVM: 0.920000 (0.040000)
In [63]:
         # KNN Classification
          #Handling Categorical values
          #Lable Encoding
          df=pd.read csv("E:\\NMDS\collegePlace.csv")
          df=df.replace(['Male'],[0])
          df=df.replace(['Female'],[1])
          df=df.replace(['Computer Science','Information Technology','Electronics And Communicat
          df.drop(['Hostel'],axis=1)
          #from pandas import read csv
          #from sklearn.model selection import KFold
          #from sklearn.model selection import cross val score
          from sklearn.neighbors import KNeighborsClassifier
          #df = read csv(filename, names=names)
          array = df.values
          X = array[:,0:7]
          Y = array[:,7]
          #num_folds = 5
          #kfold = KFold(n splits=5, random state=3)
          model = KNeighborsClassifier()
          results = cross val score(model,X,Y)
          print(results.mean())
         0.8334491627914898
         import pikle
In [66]:
          import joblib
          pickle.dump((knn,open("placement.pkl","wb"))
          model=pickle.load(open("placement.pkl","rb")))
           Input In [66]
             model=pickle.load(open("placement.pkl","rb")))
         SyntaxError: invalid syntax
         from flask
 In [ ]:
 In [ ]:
```