```
In [2]: # This Python 3 environment comes with many helpful analytics libraries instal
        # It is defined by the kaggle/python Docker image: https://github.com/kaggle/d
        ocker-python
        # For example, here's several helpful packages to load
        import warnings
        warnings.filterwarnings("ignore")
        import numpy as np # linear algebra
        import pandas as pd # data processing, CSV file I/O (e.g. pd.read csv)
        import matplotlib.pyplot as plt
        # Input data files are available in the read-only "../input/" directory
        # For example, running this (by clicking run or pressing Shift+Enter) will lis
        t all files under the input directory
        import os
        for dirname, _, filenames in os.walk('/kaggle/input'):
            for filename in filenames:
                print(os.path.join(dirname, filename))
        # You can write up to 20GB to the current directory (/kagqle/working/) that ge
        ts preserved as output when you create a version using "Save & Run All"
        # You can also write temporary files to /kaggle/temp/, but they won't be saved
        outside of the current session
```

/kaggle/input/students-datta/students\_data.csv

```
In [3]: df = pd.read_csv("/kaggle/input/students-datta/students_data.csv")
    df.head(3)
```

## Out[3]:

Oliver Thompson		Harvard University		Unnamed: 3	
0	Emma Johnson	Stanford University	3.82	NaN	
1	Liam Smith	Massachusetts Institute of Technology (MIT)	3.6	NaN	
2	Olivia Brown	University of Cambridge	2.93	NaN	

```
In [4]: columns_n=['Name', 'University', 'CGPA','none']
    df.columns = columns_n
```

In [5]: df

Out[5]:

	Name	University	CGPA	none
0	Emma Johnson	Stanford University	3.82	NaN
1	Liam Smith	Massachusetts Institute of Technology (MIT)	3.6	NaN
2	Olivia Brown	University of Cambridge	2.93	NaN
3	Noah Davis	University of Oxford	3.3	NaN
4	Ava Wilson	California Institute of Technology (Caltech)	2.95	NaN
594	NaN	NaN	NaN	NaN
595	NaN	NaN	NaN	NaN
596	NaN	NaN	NaN	NaN
597	NaN	NaN	NaN	NaN
598	NaN	NaN	NaN	NaN

599 rows × 4 columns

In [6]: df = df.drop('none', axis=1)

In [7]: df

Out[/]:		
	Name	

	Name	University	CGPA		
0	Emma Johnson	Stanford University	3.82		
1	Liam Smith	Massachusetts Institute of Technology (MIT)	3.6		
2	Olivia Brown	a Brown University of Cambridge			
3	Noah Davis	Noah Davis University of Oxford			
4	Ava Wilson	California Institute of Technology (Caltech)	2.95		
594	NaN	NaN	NaN		
595	NaN	NaN	NaN		
596	NaN	NaN	NaN		
597	NaN	NaN	NaN		
598	NaN	NaN	NaN		

599 rows × 3 columns

```
In [8]: df['CGPA'].unique()
 Out[8]: array(['3.82', '3.6', '2.93', '3.3', '2.95', '2.5', '2.8', '2.45', '2.9',
                  '3.01', '3.68', 'Annual ', '3.33', '2.33', '3.08(updated after 4 semester).', '2.76', nan, '2', '3.37',
                  '2.52', '3.1', '3.4', '3.53', '3.52', '3.2', '2.7', '2.83',
                  '3.08', '3.47', '3', '3.7', '2.58', '3.51', '3.25', '3.81', '2.97',
                        , '2.94', '3.23', '2.92', '7.5', '2.51', '2.61', '3.28',
                  '3.61'
                  '2.4', '3.26', '3.13', '2.6', '2.2', '3.18', '3.5', '2.96', '3.06',
                  '3.58', '3.21 (6th semester completed)', '2.63', '3.12', '3.318',
                  '3.45', '2.64', '2.3', '3.54', '2.84', '2.71', '3.78', '2.6/4',
                  '3.07', '3.55', '2.48', '3.94', '3.04', '2.73', '3.05', '3.66',
                  '3.74', '3.17', '2.78', '2.75', '3.95', '3.64', '3.88', '3.41'
                          '3.36', '2.89', '3.65', '2.74', '3.09', '3.24', '3.79',
                  '4', '3.34', '3.72', '3.49', '2.57', '3.85', '2.87', '3.48',
                  '3.43', '3.93/4', '3.44', '3.39', '3.59', '3.69', '3.96', '3.86',
                  '3.57', '2.34', '2.41', '2.91', '2.69', '2.55', '3.76', '3.22', '2.31', '1.8', '3.84', '3.985', '3.32', '3.15', '2.24', '3.31',
                  '3.92', '2.43'], dtype=object)
 In [9]: | df['CGPA'] = df['CGPA'].astype(str) # Convert all values to strings
          df['CGPA'] = df['CGPA'].str.extract(r'(\d+\.\d+)').astype(float) # Extract an
          d convert to float
In [10]: | df=df.dropna(subset=['CGPA'])
          df
```

## Out[10]:

	Name	University	CGPA
0	Emma Johnson	Stanford University	3.82
1	Liam Smith	Massachusetts Institute of Technology (MIT)	3.60
2	Olivia Brown	University of Cambridge	2.93
3	Noah Davis	University of Oxford	3.30
4	Ava Wilson	California Institute of Technology (Caltech)	2.95
294	Max Turner	University of Amsterdam	3.09
295	Brooklyn Wilson	University of Copenhagen	3.23
296	Colton Carter	University of California, Santa Barbara (UCSB)	2.20
297	Gabriella Lee	King's College London	2.50
298	Jack Green	University of California, Irvine	2.43

274 rows × 3 columns

```
In [11]: | df = df.loc[df['CGPA'] <= 4]</pre>
         top_uni = df.groupby('University')['CGPA'].mean().sort_values(ascending=Fals
         e).head(5)
         top uni
Out[11]: University
         University of Toronto
                                                                        3.525
         Massachusetts Institute of Technology (MIT)
                                                                        3.474
         University of Southern California (USC)
                                                                        3.455
         University of Bristol
                                                                        3.440
         Swiss Federal Institute of Technology Zurich (ETH Zurich)
                                                                        3.410
         Name: CGPA, dtype: float64
In [12]: | df['Name Length'] = df['Name'].str.len()
         # Calculate the correlation between 'CGPA' and 'Name Length'
         correlation = df['CGPA'].corr(df['Name_Length'])
         # Print the correlation value
         print("Correlation between CGPA and Name Length:", correlation)
         Correlation between CGPA and Name Length: 0.021044223047317186
In [13]: | c_uni = df.groupby('University')['CGPA'].describe()
         c_uni
Out[13]:
```

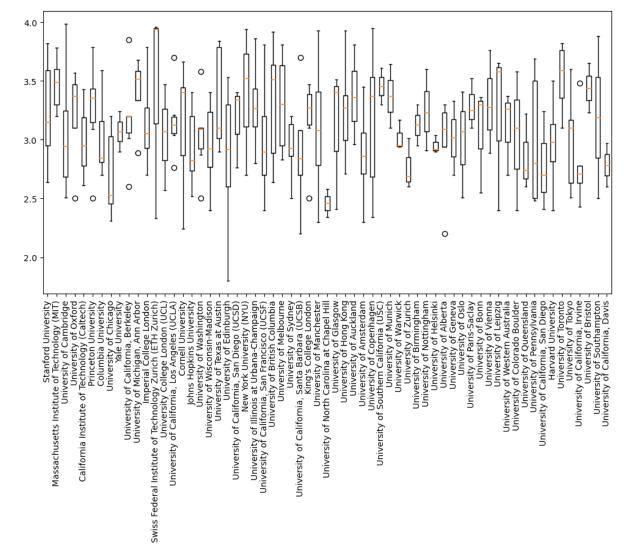
	count	mean	std	min	25%	50%	75%	max
University								
California Institute of Technology (Caltech)	3.0	2.996667	0.411987	2.61	2.7800	2.950	3.1900	3.43
Columbia University	6.0	3.006667	0.344654	2.70	2.8075	2.845	3.1600	3.59
Cornell University	6.0	3.141667	0.549014	2.24	2.8675	3.400	3.4450	3.66
Harvard University	4.0	2.965000	0.450074	2.40	2.8125	2.980	3.1325	3.50
Imperial College London	6.0	3.138333	0.381650	2.70	2.9275	3.055	3.2725	3.79
<b>University of Washington</b>	6.0	3.038333	0.352387	2.50	2.9225	3.090	3.1000	3.58
University of Western Australia	3.0	3.110000	0.359305	2.70	2.9800	3.260	3.3150	3.37
University of Wisconsin-Madison	6.0	2.951667	0.373117	2.40	2.7700	2.925	3.2375	3.40
University of Zurich	3.0	2.766667	0.215484	2.60	2.6450	2.690	2.8500	3.01
Yale University	3.0	3.070000	0.170000	2.90	2.9850	3.070	3.1550	3.24

63 rows × 8 columns

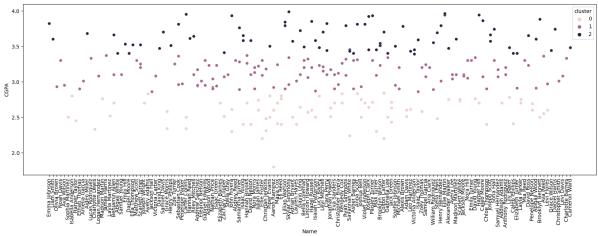
```
In [15]: from sklearn.model selection import train test split
         # Assuming you have already loaded the data into the 'df' DataFrame
         # Filter out non-string values from the 'Name' column
         df = df[df['Name'].apply(lambda x: isinstance(x, str))]
         x = df['Name'].apply(len).values.reshape(-1, 1)
         y = df['CGPA'].values
         x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.35, rand
         om state=42)
In [18]: from sklearn.linear model import LinearRegression
         mod=LinearRegression()
         mod.fit(x train,y train)
Out[18]:
          ▼ LinearRegression
          LinearRegression()
In [25]:
         import statsmodels.api as sm
         from sklearn.metrics import mean squared error
         y pred = mod.predict(x test)
In [27]: | mse=mean_squared_error(y_test,y_pred)
         mse
Out[27]: 0.18228806353414384
In [ ]:
In [41]:
         t = 3.5
         uni=df[df['CGPA']>t]['University'].value counts().idxmax()
```

```
outliers=df.groupby('University')['CGPA'].apply(lambda x: x[x.between(*x.quant
          ile([0.25,0.75]).values)])
         outliers
Out[42]: University
         California Institute of Technology (Caltech)
                                                                2.95
         Columbia University
                                                        106
                                                                2.86
                                                         185
                                                                2.83
         Cornell University
                                                        18
                                                                3.37
                                                         187
                                                                3.43
                                                                . . .
         University of Western Australia
                                                         59
                                                                3.26
         University of Wisconsin-Madison
                                                         121
                                                                3.05
                                                         268
                                                                2.80
         University of Zurich
                                                         245
                                                                2.69
         Yale University
                                                         108
                                                                3.07
         Name: CGPA, Length: 105, dtype: float64
```

```
In [43]: # box plot for cgpa stats
    data_plot = [df[df.University == uni]['CGPA'] for uni in df.University.unique
    ()]
    plt.figure(figsize=(12, 6))
    plt.boxplot(data_plot, labels=df.University.unique())
    plt.xticks(rotation=90)
    plt.show()
```



```
In [44]: from sklearn.cluster import KMeans
    import matplotlib.pyplot as plt
    import seaborn as sns
    k=df[['CGPA']]
    kmean=KMeans(n_clusters=3,random_state=42)
    df['cluster']=kmean.fit_predict(k)
    plt.figure(figsize=(20,6))
    sns.scatterplot(x='Name',y='CGPA',hue='cluster',data=df)
    plt.xticks(rotation=90)
    plt.show()
```



```
In [45]: avg=df.groupby('cluster')['CGPA'].mean()
avg
```

Out[45]: cluster

0 2.5750621 3.1192932 3.632434

Name: CGPA, dtype: float64

```
In [47]: from sklearn.tree import DecisionTreeClassifier
         from sklearn.model selection import train test split
         from sklearn.metrics import accuracy score
         import pandas as pd
         # Assuming you have already loaded the data into the 'df' DataFrame
         # Filter out non-string values from the 'Name' column
         df = df[df['Name'].apply(lambda x: isinstance(x, str))]
         # Select the features (independent variables)
         features = df[['CGPA']]
         # Select the target variable (dependent variable)
         target = df['University'] # Replace 'University' with the actual target colum
         n name
         # Split the data into training and testing sets for both features and target
         x_train, x_test, y_train, y_test = train_test_split(features, target, test_siz
         e=0.35, random state=42)
         # Create a DecisionTreeClassifier model
         model = DecisionTreeClassifier(random state=42)
         # Fit the model to the training data
         model.fit(x train, y train)
         # Predict the target values using the model
         y pred = model.predict(x test)
         # Calculate the accuracy score
         accuracy = accuracy score(y test, y pred)
         print("Accuracy:", accuracy)
```

Accuracy: 0.010416666666666666

```
In [48]: avg2=df['CGPA'].mean()
avg2
```

Out[48]: 3.1006703296703293

```
In [50]: from sklearn.preprocessing import LabelEncoder
         # Encode the 'University' column using label encoding
         encoder = LabelEncoder()
         df['University encoded'] = encoder.fit transform(df['University'])
         # Task 1: Build a Regression Model for CGPA Prediction
         x regression = df[['University encoded']]
         y regression = df['CGPA']
         x_train_reg, x_test_reg, y_train_reg, y_test_reg = train_test_split(x_regressi
         on, y regression, test size=0.2, random state=42)
         regression model = LinearRegression()
         regression_model.fit(x_train_reg, y_train_reg)
         y pred reg = regression model.predict(x test reg)
         mse = mean squared error(y test reg, y pred reg)
         print("Regression Model MSE:", mse)
         Regression Model MSE: 0.18061559865339447
In [51]: # Task 3: Check for Missing or Erroneous CGPA Values
         missing cgpa = df['CGPA'].isna().sum()
         print("Missing CGPA Values:", missing cgpa)
         Missing CGPA Values: 0
In [54]: from sklearn.cluster import AgglomerativeClustering
         x clustering = df[['CGPA']]
         clustering model = AgglomerativeClustering(n clusters=3)
         df['cluster'] = clustering model.fit predict(x clustering)
In [55]: # Task 7: Build a Classification Model for University Prediction
         x_classification = df[['CGPA', 'Name_Length']]
         y_classification = df['University']
         x_train_class, x_test_class, y_train_class, y_test_class = train_test_split(x_
         classification, y classification, test size=0.2, random state=42)
         classification model = DecisionTreeClassifier(random state=42)
         classification model.fit(x train class, y train class)
```

Classification Model Accuracy: 0.03636363636363636

y\_pred\_class = classification\_model.predict(x\_test\_class)
accuracy = accuracy\_score(y\_test\_class, y\_pred\_class)
print("Classification Model Accuracy:", accuracy)

```
correlations = df.groupby('University')[['Name Length', 'CGPA']].corr().loc[:,
In [56]:
          'Name Length']
          print("Correlation between Name Length and CGPA within Each University:\n", co
          rrelations)
          Correlation between Name Length and CGPA within Each University:
          California Institute of Technology (Caltech)
                                                             Name_Length
                                                                             1.000000
                                                             CGPA
                                                                             0.098097
          Columbia University
                                                             Name_Length
                                                                             1.000000
                                                             CGPA
                                                                             -0.199380
          Cornell University
                                                             Name_Length
                                                                             1.000000
                                                                                . . .
          University of Wisconsin-Madison
                                                             CGPA
                                                                             0.742487
          University of Zurich
                                                             Name Length
                                                                             1.000000
                                                             CGPA
                                                                             -0.977951
          Yale University
                                                             Name_Length
                                                                             1.000000
                                                             CGPA
                                                                             0.866025
          Name: Name Length, Length: 126, dtype: float64
In [57]: # Visualizations
          plt.figure(figsize=(10, 6))
          sns.scatterplot(x='CGPA', y='University', hue='cluster', data=df)
          plt.show()
                   Massachusetts Institute
                   California Institute of
                 University of California
                   University of North
                                               2.0
                                                                                             4.0
                                                                                 3.5
 In [ ]:
```