```
In [ ]: import pandas as pd
        import matplotlib.pyplot as plt
        from sklearn.linear model import LogisticRegression
        from sklearn.model selection import train test split
        from sklearn.metrics import confusion matrix, accuracy score, classification
        # Disable scientific notation for large numbers
        pd.options.display.float format = '{:.0f}'.format
        # Setting display options for Pandas to show three decimal places for floati
        pd.set option('display.float format', lambda x: '%.2f' % x)
```

Data Loading

```
In [ ]: |# import data
        student performance df = pd.read csv('/content/drive/MyDrive/student perform
```

Data Exploration

```
In [ ]: student performance df.info() # Display information about the DataFrame, ind
       <class 'pandas.core.frame.DataFrame'>
      RangeIndex: 40000 entries, 0 to 39999
      Data columns (total 7 columns):
           Column
                                                        Non-Null Count Dtype
       --- -----
           Student ID
                                                        40000 non-null object
       1 Study Hours per Week
                                                        38005 non-null float64
                                                        38008 non-null float64
          Attendance Rate
       3
          Previous Grades
                                                        38006 non-null float64
          Participation in Extracurricular Activities 38000 non-null object
       5
           Parent Education Level
                                                        38000 non-null object
           Passed
                                                        38000 non-null object
       dtypes: float64(3), object(4)
       memory usage: 2.1+ MB
In [ ]: student performance df.head() # Display top 5 records
```

Out[]:		Student ID	Study Hours per Week	Attendance Rate	Previous Grades	Participation in Extracurricular Activities	Parent Education Level	Passed
	0	S00001	12.50	NaN	75.00	Yes	Master	Yes
	1	S00002	9.30	95.30	60.60	No	High School	No
	2	S00003	13.20	NaN	64.00	No	Associate	No
	3	S00004	17.60	76.80	62.40	Yes	Bachelor	No
	4	S00005	8.80	89.30	72.70	No	Master	No

In []: student_performance_df.describe() # Statistical info about dataset

Out[]:	Study Hours per Week	Attendance Rate	Previous Grades
out[]:	Study Hours per Week	Attenuance Rate	Frevious Grade

•	•		
count	38005.00	38008.00	38006.00
mean	9.96	75.28	65.44
std	5.03	20.39	16.50
min	-12.30	-14.30	8.30
25%	6.60	61.60	55.10
50%	10.00	75.30	65.20
75 %	13.40	88.80	75.20
max	32.40	150.20	200.00

Data Preprocessing

```
In [ ]: # Map 'yes' to 1 and 'no' to 0
    student_performance_df['Passed'] = student_performance_df['Passed'].map({'Yestudent_performance_df['Participation in Extracurricular Activities'] = student_performance_df['Participation in Extracurr
```

Out[]:		Student ID	Study Hours per Week	Attendance Rate	Previous Grades	Participation in Extracurricular Activities	Parent Education Level	Passed
	0	S00001	12.50	NaN	75.00	1.00	2.00	1.00
	1	S00002	9.30	95.30	60.60	0.00	0.00	0.00
	2	S00003	13.20	NaN	64.00	0.00	3.00	0.00
	3	S00004	17.60	76.80	62.40	1.00	1.00	0.00
	4	S00005	8.80	89.30	72.70	0.00	2.00	0.00

In []: student_performance_df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 40000 entries, 0 to 39999
Data columns (total 7 columns):

#	Column	Non-Null Count	Dtype
0	Student ID	40000 non-null	object
1	Study Hours per Week	38005 non-null	float64
2	Attendance Rate	38008 non-null	float64
3	Previous Grades	38006 non-null	float64
4	Participation in Extracurricular Activities	38000 non-null	float64
5	Parent Education Level	38000 non-null	float64
6	Passed	38000 non-null	float64

dtypes: float64(6), object(1)

memory usage: 2.1+ MB

Data Cleaning

```
In []: # Replace null values in 'Passed', 'Participation in Extracurricular Activiti
    student_performance_df['Passed'].fillna(0, inplace=True)
    student_performance_df['Participation in Extracurricular Activities'].fillna
    student_performance_df['Parent Education Level'].fillna(0, inplace=True)
```

/tmp/ipython-input-12-1938445274.py:2: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behave s as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'd f.method({col: value}, inplace=True)' or df[col] = df[col].method(value) ins tead, to perform the operation inplace on the original object.

student performance df['Passed'].fillna(0, inplace=True)

/tmp/ipython-input-12-1938445274.py:3: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

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student_performance_df['Participation in Extracurricular Activities'].fill
na(0, inplace=True)

/tmp/ipython-input-12-1938445274.py:4: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behave s as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'd f.method({col: value}, inplace=True)' or df[col] = df[col].method(value) ins tead, to perform the operation inplace on the original object.

student performance df['Parent Education Level'].fillna(0, inplace=True)

In []: student_performance_df.isna().sum()

	0
Student ID	0
Study Hours per Week	1995
Attendance Rate	1992
Previous Grades	1994
rticipation in Extracurricular Activities	0
Parent Education Level	0
Passed	0
	Study Hours per Week Attendance Rate Previous Grades rticipation in Extracurricular Activities Parent Education Level

dtype: int64

```
In []: # Convert 'Passed', 'Participation in Extracurricular Activities' and 'Parent
student_performance_df['Passed'] = student_performance_df['Passed'].astype(i
student_performance_df['Parent Education Level'] = student_performance_df['F
student_performance_df['Participation in Extracurricular Activities'] = student_performance_df['Participation in Extracurricular Activities']
```

In []: student_performance_df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 40000 entries, 0 to 39999
Data columns (total 7 columns):

#	Column	Non-Null Count	Dtype
0	Student ID	40000 non-null	object
1	Study Hours per Week	38005 non-null	float64
2	Attendance Rate	38008 non-null	float64
3	Previous Grades	38006 non-null	float64
4	Participation in Extracurricular Activities	40000 non-null	int64
5	Parent Education Level	40000 non-null	int64
6	Passed	40000 non-null	int64

dtypes: float64(3), int64(3), object(1)

memory usage: 2.1+ MB

In []: student performance df.head()

Out[]:		Student ID	Study Hours per Week	Attendance Rate	Previous Grades	Participation in Extracurricular Activities	Parent Education Level	Passed
	0	S00001	12.50	NaN	75.00	1	2	1
	1	S00002	9.30	95.30	60.60	0	0	0
	2	S00003	13.20	NaN	64.00	0	3	0
	3	S00004	17.60	76.80	62.40	1	1	0
	4	S00005	8.80	89.30	72.70	0	2	0

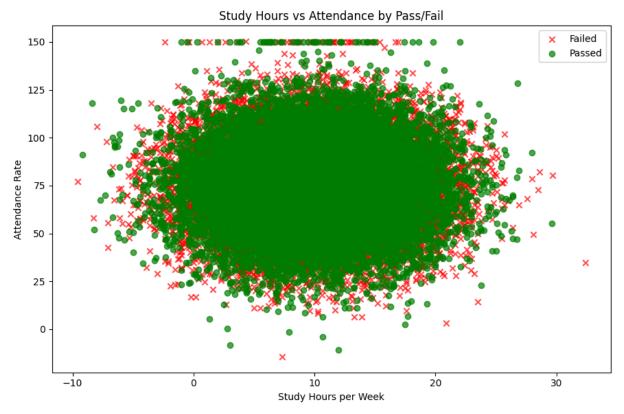
dtype: int64

```
In [ ]: student_performance_df.fillna(0, inplace=True) # Replace missing value with
In [ ]: print(student_performance_df.isna().sum().sum()) # Find sum of missing value
0
In [ ]: print(student_performance_df.duplicated().sum()) # Find sum of duplicated value of the student of the stude
```

Since, sum of duplicated values is zero, so there is no need to drop duplicates.

Data Exploration

```
plt.title('Study Hours vs Attendance by Pass/Fail')
plt.legend()
plt.tight_layout()
plt.show()
```



Model training and evaluation

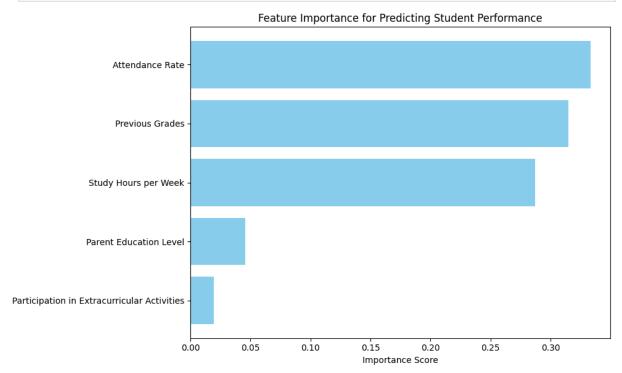
```
In [ ]: # Select features and target, handle missing values and target conversions
        X = student performance df[['Study Hours per Week', 'Attendance Rate']]
        y = student performance df['Passed']
        # Split into training and test datasets
        X train, X test, y train, y test = train test split(X, y, test size=0.2, rar
        # Instantiate and train logistic regression model
        model = LogisticRegression()
        model.fit(X train, y train)
        # Model predictions
        y pred = model.predict(X test)
        # Print Evaluation metrics
        print("Confusion Matrix:")
        print(confusion_matrix(y_test, y_pred))
        print("\nAccuracy Score:", accuracy score(y test, y pred))
        print("\nClassification Report:")
        print(classification report(y test, y pred))
```

```
Confusion Matrix:
[[3803 1]
[3419
       011
Accuracy Score: 0.5265125294199087
Classification Report:
           precision recall f1-score support
             0.53 1.00
        0
                              0.69
                                      3804
              0.00
                      0.00
                              0.00
                                      3419
                              0.53
                                      7223
   accuracy
            0.26 0.50
                             0.34
                                      7223
  macro avq
weighted avg 0.28
                             0.36
                     0.53
                                      7223
```

Key predictors of student performance

```
In [ ]: from sklearn.ensemble import RandomForestClassifier
        # Define features and target
        exclude cols = ['Student ID', 'Passed'] # Exclude ID and target from featur
        features = [col for col in student performance df.columns if col not in excl
        X = student performance df[features]
        y = student performance df['Passed']
        # Train a Random Forest Classifier
        model = RandomForestClassifier(random state=42)
        model.fit(X, y)
        # Feature importance
        importances = model.feature importances
        feature ranking = sorted(zip(features, importances), key=lambda x: x[1], rev
        print("Key Predictors of Student Performance:")
        for feature, importance in feature ranking:
            print(f"{feature}: {importance:.3f}")
       Kev Predictors of Student Performance:
       Attendance Rate: 0.333
       Previous Grades: 0.315
       Study Hours per Week: 0.287
       Parent Education Level: 0.046
       Participation in Extracurricular Activities: 0.020
In [ ]: # Create plot for key predictors of student performance
        feature names = X.columns
        # Sort features by importance
        indices = importances.argsort()[::-1]
        sorted features = feature names[indices]
```

```
# Plot
plt.figure(figsize=(10,6))
plt.barh(sorted_features, sorted_importances, color='skyblue')
plt.xlabel("Importance Score")
plt.title("Feature Importance for Predicting Student Performance")
plt.gca().invert_yaxis() # Highest at top
plt.tight_layout()
plt.show()
```



This notebook was converted with convert.ploomber.io