

## ✓ School Result Data Analyzer .

```
1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4 import seaborn as sns
5 from sklearn.model_selection import train_test_split
6 from sklearn.linear_model import LogisticRegression
7 from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score, confusion_ma
```

### ✓ 1. Load Dataset

```
1 df = pd.read_csv("/content/drive/MyDrive/school_results_cleaned.csv")
```

### ✓ 2. Data Cleaning

```
1 df.drop_duplicates(inplace=True)
2
3 subjects = ["Hindi", "English", "Science", "Maths", "History", "Geography"]
4 df[subjects] = df[subjects].astype(int)
5
6 df["Total"] = df[subjects].sum(axis=1)
7 df["Pass/Fail"] = df["Total"].apply(lambda x: "Pass" if (x/len(subjects)) >= 40 else "Fail")
```

### ✓ 3. Subject-wise Analysis

```
1 subject_avg = df[subjects].mean().round(2)
2 subject_topper = {sub: df.loc[df[sub].idxmax(), "Student"] for sub in subjects}
3
4 print("\n--- Subject Averages ---")
5 print(subject_avg)
6
7 print("\n--- Subject Toppers ---")
8 for sub, stu in subject_topper.items():
9     marks = df.loc[df[sub].idxmax(), sub]
10    print(f"{sub} → {stu} ({marks} marks)")
```



```
--- Subject Averages ---
Hindi      51.64
English    50.11
Science     49.44
Maths       49.55
History     49.03
Geography   50.03
dtype: float64

--- Subject Toppers ---
Hindi → S246 (99 marks)
English → S252 (99 marks)
Science → S103 (99 marks)
Maths → S3 (99 marks)
History → S196 (99 marks)
Geography → S235 (99 marks)
```

### ✓ 4. Class-wise Toppers

```
1 class_toppers = df.groupby("Class").apply(lambda x: x.loc[x["Total"].idxmax()][["Student", "Total"]])
2
3 print("\n--- Class-wise Toppers ---")
4 print(class_toppers)
```

```

5
6 # 5. Pass/Fail Rate
7 pass_rate = round((df['Pass/Fail'].value_counts(normalize=True)['Pass']*100),2)
8 print(f"\nPass Rate: {pass_rate}%")

```

↻

```

--- Class-wise Toppers ---
      Student  Total
Class
9          S533   505
10         S574   467
11         S760   447
12         S613   474

```

Pass Rate: 80.6%

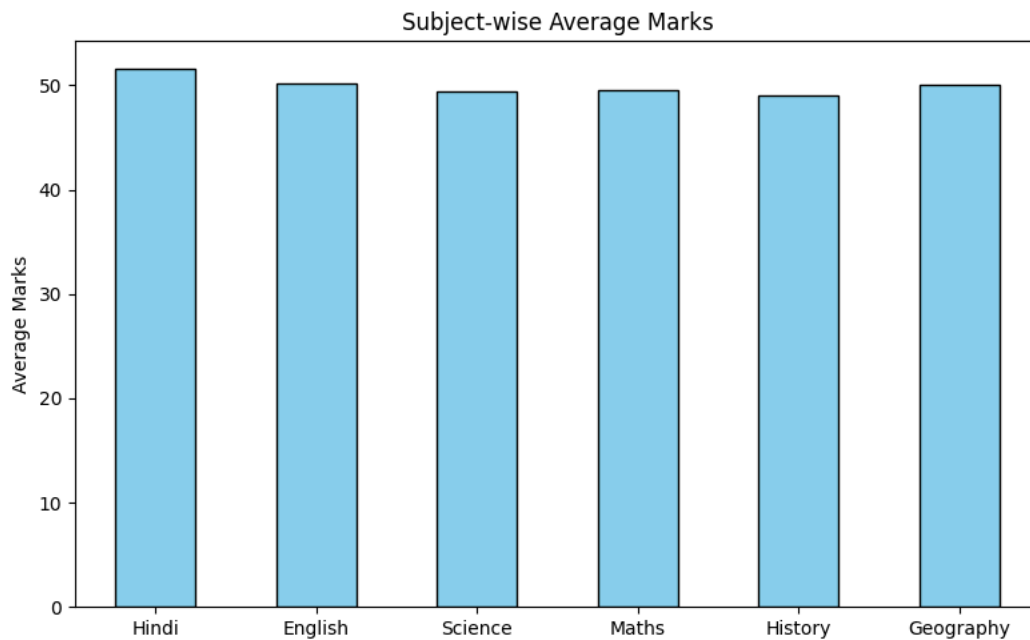
/tmp/ipython-input-745328902.py:1: DeprecationWarning: DataFrameGroupBy.apply operated on the grouping columns. This behavior is deprecated  
class\_toppers = df.groupby("Class").apply(lambda x: x.loc[x["Total"].idxmax()][["Student", "Total"]])

## 6. Visualizations

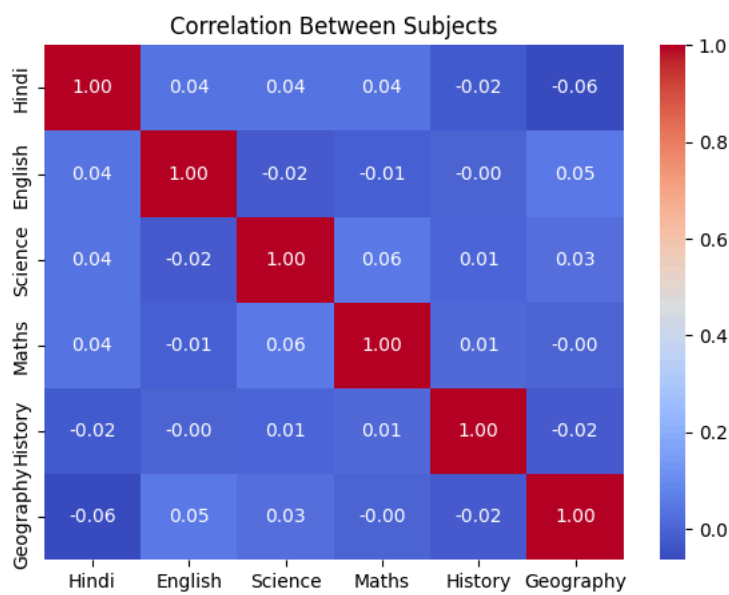
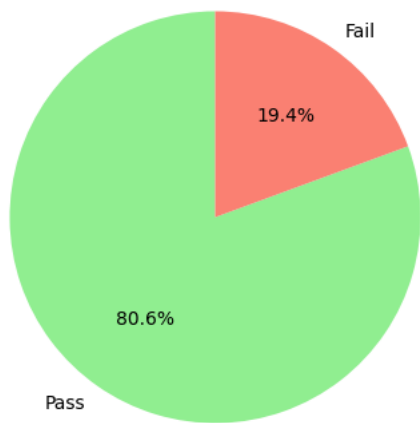
```

1 plt.figure(figsize=(8,5))
2 subject_avg.plot(kind="bar", color="skyblue", edgecolor="black")
3 plt.title("Subject-wise Average Marks")
4 plt.ylabel("Average Marks")
5 plt.xticks(rotation=0)
6 plt.tight_layout()
7 plt.show()
8
9 plt.figure(figsize=(5,5))
10 df["Pass/Fail"].value_counts().plot(kind="pie", autopct='%1.1f%%', startangle=90, colors=["light
11 plt.title("Pass vs Fail Distribution")
12 plt.ylabel("")
13 plt.show()
14
15 plt.figure(figsize=(7,5))
16 sns.heatmap(df[subjects].corr(), annot=True, cmap="coolwarm", fmt=".2f")
17 plt.title("Correlation Between Subjects")
18 plt.show()

```



Pass vs Fail Distribution



## 7. ML Model: Predict Pass/Fail

```

1 X = df[subjects]
2 y = df["Pass/Fail"].map({"Pass":1, "Fail":0})
3

```

```

4 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
5
6 model = LogisticRegression(max_iter=1000)
7 model.fit(X_train, y_train)
8
9 y_pred = model.predict(X_test)
10
11 # Evaluation Metrics
12 acc = accuracy_score(y_test, y_pred)
13 prec = precision_score(y_test, y_pred)
14 rec = recall_score(y_test, y_pred)
15 f1 = f1_score(y_test, y_pred)
16 cm = confusion_matrix(y_test, y_pred)
17
18 print("\n--- Model Evaluation ---")
19 print(f"Accuracy: {acc:.2f}")
20 print(f"Precision: {prec:.2f}")
21 print(f"Recall: {rec:.2f}")
22 print(f"F1 Score: {f1:.2f}")
23 print("\nConfusion Matrix:\n", cm)
24 print("\nClassification Report:\n", classification_report(y_test, y_pred))

```



--- Model Evaluation ---

Accuracy: 0.99

Precision: 0.99

Recall: 0.99

F1 Score: 0.99

Confusion Matrix:

```
[[ 43  1]
 [  1 155]]
```

Classification Report:

	precision	recall	f1-score	support
0	0.98	0.98	0.98	44
1	0.99	0.99	0.99	156
accuracy			0.99	200
macro avg	0.99	0.99	0.99	200
weighted avg	0.99	0.99	0.99	200

## 8. Export Summary Report

```

1 report = {
2     "Subject Averages": subject_avg.to_dict(),
3     "Subject Toppers": subject_topper,
4     "Class Toppers": class_toppers.to_dict(),
5     "Pass Rate (%)": pass_rate,
6     "Model Accuracy": round(acc,2),
7     "Model Precision": round(prec,2),
8     "Model Recall": round(rec,2),
9     "Model F1 Score": round(f1,2)

```