

```

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import (
    accuracy_score,
    classification_report,
    confusion_matrix,
    roc_curve,
    auc
)

```

```

url = "https://raw.githubusercontent.com/plotly/datasets/master/diabetes.csv"
df = pd.read_csv(url)

```

```
df.head()
```

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age	Outcome	grid icon
0	6	148	72	35	0	33.6		0.627	50	1
1	1	85	66	29	0	26.6		0.351	31	0
2	8	183	64	0	0	23.3		0.672	32	1
3	1	89	66	23	94	28.1		0.167	21	0
4	0	137	40	35	168	43.1		2.288	33	1

Next steps: [Generate code with df](#) [New interactive sheet](#)

```

df.info()
df.describe()

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 9 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Pregnancies      768 non-null    int64  
 1   Glucose          768 non-null    int64  
 2   BloodPressure    768 non-null    int64  
 3   SkinThickness    768 non-null    int64  
 4   Insulin          768 non-null    int64  
 5   BMI              768 non-null    float64 
 6   DiabetesPedigreeFunction 768 non-null    float64 
 7   Age              768 non-null    int64  
 8   Outcome          768 non-null    int64  
dtypes: float64(2), int64(7)
memory usage: 54.1 KB

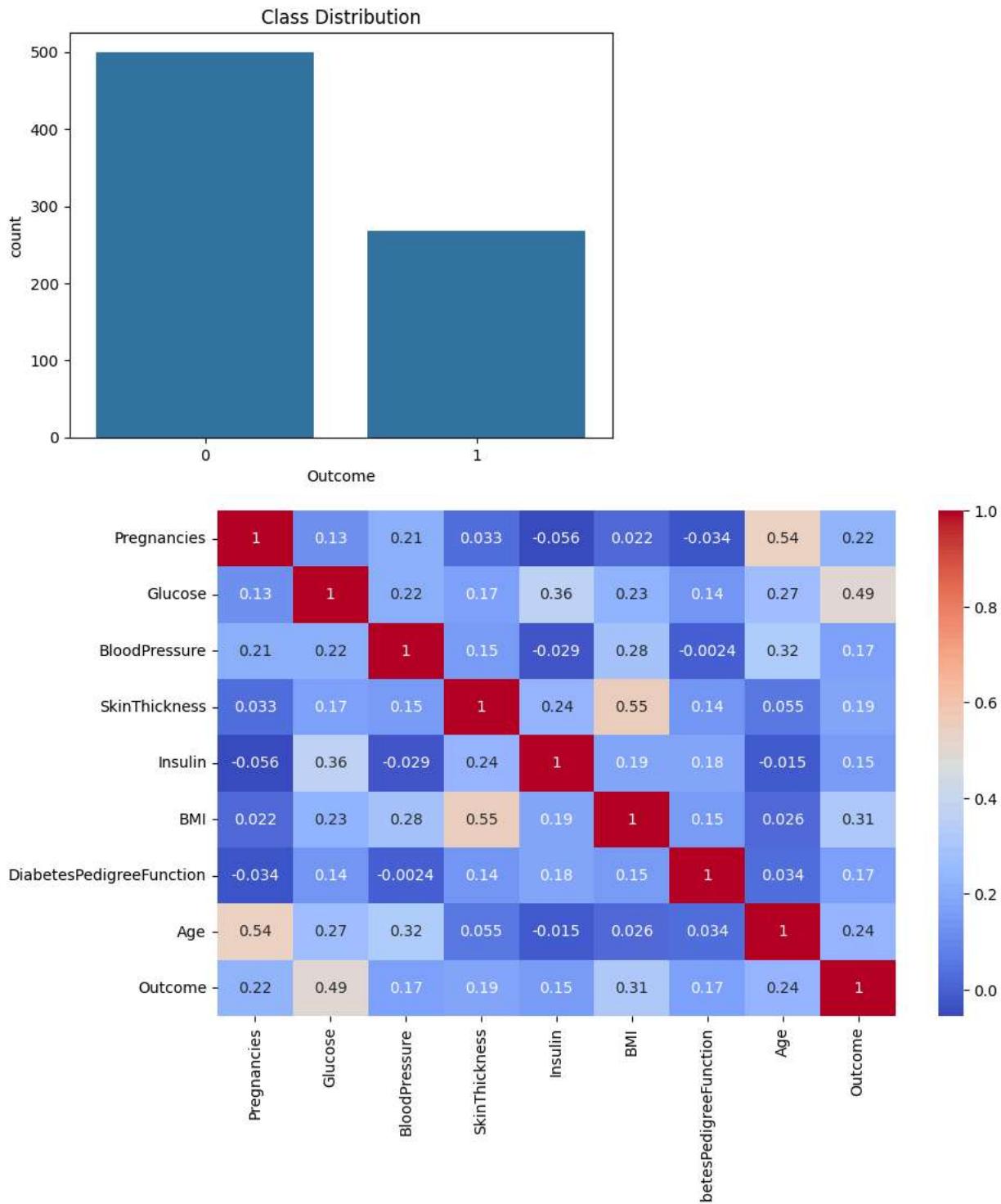
```

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age	Outcome
count	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000		768.000000	768.000000
mean	3.845052	120.894531	69.105469	20.536458	79.799479	31.992578		0.471876	33.240885
std	3.369578	31.972618	19.355807	15.952218	115.244002	7.884160		0.331329	11.760232
min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000		0.078000	21.000000
25%	1.000000	99.000000	62.000000	0.000000	0.000000	27.300000		0.243750	24.000000
50%	3.000000	117.000000	72.000000	23.000000	30.500000	32.000000		0.372500	29.000000
75%	6.000000	140.250000	80.000000	32.000000	127.250000	36.600000		0.626250	41.000000
max	17.000000	199.000000	122.000000	99.000000	846.000000	67.100000		2.420000	81.000000

```
cols_with_zero = ["Glucose", "BloodPressure", "SkinThickness", "Insulin", "BMI"]
```

```
for col in cols_with_zero:
    df[col] = df[col].replace(0, df[col].median())
```

```
sns.countplot(x="Outcome", data=df)
plt.title("Class Distribution")
plt.show()
plt.figure(figsize=(10,6))
sns.heatmap(df.corr(), annot=True, cmap="coolwarm")
plt.show()
```



```
x = df.drop("Outcome", axis=1)
y = df["Outcome"]
```

```
X_train, X_test, y_train, y_test = train_test_split(  
    X, y,  
    test_size=0.2,  
    random_state=42,  
    stratify=y  
)
```

```
scaler = StandardScaler()  
X_train = scaler.fit_transform(X_train)  
X_test = scaler.transform(X_test)
```

```
model = LogisticRegression()  
model.fit(X_train, y_train)
```

* LogisticRegression ⓘ ⓘ

```
LogisticRegression()
```

```
print("Accuracy:", accuracy_score(y_test, model.predict(X_test)))  
print(classification_report(y_test, model.predict(X_test)))  
cm = confusion_matrix(y_test, model.predict(X_test))  
sns.heatmap(cm, annot=True, fmt="d", cmap="Blues")  
plt.xlabel("Predicted")  
plt.ylabel("Actual")  
plt.show()  
y_prob = model.predict_proba(X_test)[:,1]  
fpr, tpr, _ = roc_curve(y_test, y_prob)  
roc_auc = auc(fpr, tpr)  
  
plt.plot(fpr, tpr, label="AUC = %.2f" % roc_auc)  
plt.plot([0,1], [0,1], linestyle="--")  
plt.legend()  
plt.show()
```

Accuracy: 0.7077922077922078

	precision	recall	f1-score	support
0	0.75	0.82	0.78	100
1	0.60	0.50	0.55	54
accuracy			0.71	154
macro avg	0.68	0.66	0.67	154
weighted avg	0.70	0.71	0.70	154

