

```
import numpy as np
import pandas as pd
```

```
df=pd.read_csv('pima-indians-diabetes.data.csv')
df.head()
```

6	148	72	35	0	33.6	0.627	50	1	
0	1	85	66	29	0	26.6	0.351	31	0
1	8	183	64	0	0	23.3	0.672	32	1
2	1	89	66	23	94	28.1	0.167	21	0
3	0	137	40	35	168	43.1	2.288	33	1
4	5	116	74	0	0	25.6	0.201	30	0

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

```
df.columns = ['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin', 'BMI', 'DiabetesPedigree', 'Age', 'Outcome']
X = df.drop('Outcome', axis=1)
y = df['Outcome']
```

```
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25, random_state=42, stratify=y)
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
```

```
from sklearn.svm import SVC
from sklearn.naive_bayes import GaussianNB
from sklearn.tree import DecisionTreeClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score, recall_score, precision_score, f1_score, confusion_matrix, roc_curve, auc

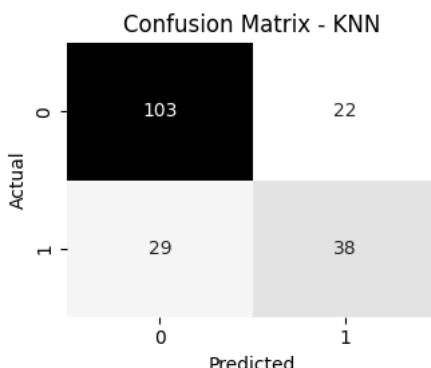
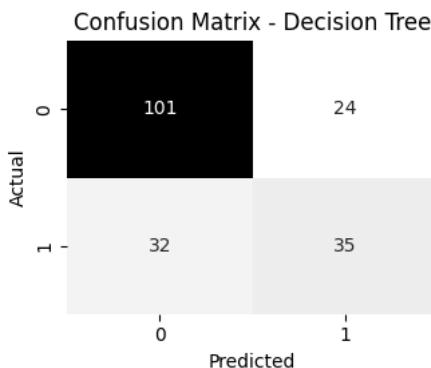
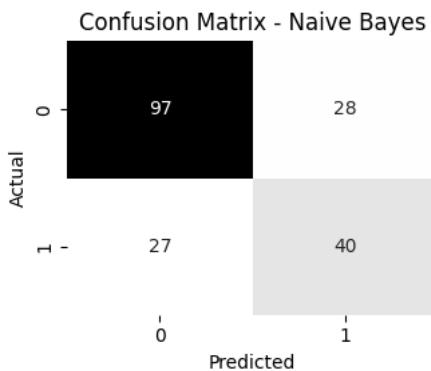
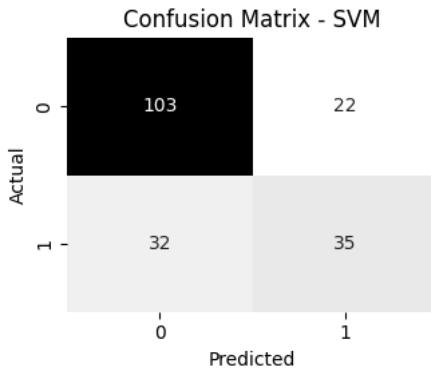
models = {
    'SVM': SVC(kernel='rbf', probability=True, random_state=42),
    'Naive Bayes': GaussianNB(),
    'Decision Tree': DecisionTreeClassifier(random_state=42),
    'KNN': KNeighborsClassifier(n_neighbors=5)
}

results = {}
for name, model in models.items():
    model.fit(X_train, y_train)
    y_pred = model.predict(X_test)
    y_score = model.predict_proba(X_test)[:,1]
    acc = accuracy_score(y_test, y_pred)
    rec = recall_score(y_test, y_pred)
    prec = precision_score(y_test, y_pred)
    f1 = f1_score(y_test, y_pred)
    cm = confusion_matrix(y_test, y_pred)
    fpr, tpr, _ = roc_curve(y_test, y_score)
    roc_auc = auc(fpr, tpr)
    results[name] = {
        'accuracy': acc,
        'recall': rec,
        'precision': prec,
        'f1': f1,
        'confusion_matrix': cm,
        'fpr': fpr,
        'tpr': tpr,
        'auc': roc_auc
    }
```

```
import seaborn as sns
import matplotlib.pyplot as plt

for name, r in results.items():
    plt.figure(figsize=(3.5,3))
    sns.heatmap(r['confusion_matrix'], annot=True, fmt='d', cmap='Greys', cbar=False)
    plt.title(f'Confusion Matrix - {name}')
    plt.xlabel('Predicted')
```

```
plt.ylabel('Actual')
plt.tight_layout()
plt.show()
```



```
comparison = pd.DataFrame({
    'Model': list(results.keys()),
    'Accuracy': [results[m]['accuracy'] for m in results],
    'Recall': [results[m]['recall'] for m in results],
    'Precision': [results[m]['precision'] for m in results],
    'F1-Score': [results[m]['f1'] for m in results]
})
print(comparison)
```

	Model	Accuracy	Recall	Precision	F1-Score
0	SVM	0.718750	0.522388	0.614035	0.564516
1	Naive Bayes	0.713542	0.597015	0.588235	0.592593
2	Decision Tree	0.708333	0.522388	0.593220	0.555556
3	KNN	0.734375	0.567164	0.633333	0.598425

```
plt.figure(figsize=(6,6))
linestyles = ['-','--',':', '-.']
for (name, r), ls in zip(results.items(), linestyles):
    plt.plot(r['fpr'], r['tpr'], linestyle=ls, label=f"{name} (AUC={r['auc']:.3f})")
plt.plot([0,1],[0,1], linestyle=':', linewidth=0.8)
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('ROC Curve Comparison')
plt.legend(edgecolor='black', fontsize=8)
plt.grid(True, linestyle=':', linewidth=0.5)
plt.tight_layout()
plt.show()
```

