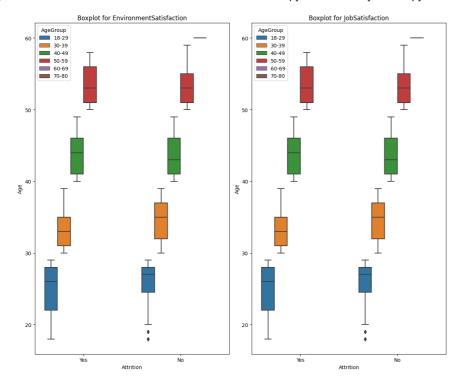
```
import pandas as pd
df = pd.read_csv("/content/HR-Employee-Attrition.csv")
# Remove null values
df_cleaned = df.dropna()
# Remove duplicate rows
df_cleaned = df_cleaned.drop_duplicates()
# Display the cleaned DataFrame
print("\nDataFrame after removing null values and duplicates:")
print(df_cleaned)
     DataFrame after removing null values and duplicates:
           Age Attrition
                              BusinessTravel DailyRate
                                                                       Department
     0
            41
                      Yes
                               Travel_Rarely
                                                    1102
                                                                            Sales
     1
            49
                      No
                           Travel_Frequently
                                                     279 Research & Development
                               Travel_Rarely
                                                    1373 Research & Development
     2
            37
                      Yes
     3
                           Travel Frequently
                                                    1392 Research & Development
            33
                      No
     4
            27
                      No
                               Travel_Rarely
                                                     591 Research & Development
                           Travel_Frequently
                                                     884 Research & Development
     1465
            36
                      Nο
     1466
            39
                       No
                               Travel_Rarely
                                                     613
                                                           Research & Development
                               Travel_Rarely
     1467
            27
                       No
                                                     155
                                                           Research & Development
     1468
            49
                       No
                           Travel_Frequently
                                                     1023
     1469
            34
                               Travel_Rarely
                                                     628 Research & Development
                      No
           DistanceFromHome
                              Education EducationField EmployeeCount \
     0
                                       2 Life Sciences
                           1
                           8
                                       1 Life Sciences
     1
                                                                      1
     2
                           2
                                       2
                                                  0ther
                                                                      1
     3
                           3
                                       4
                                          Life Sciences
                                                                      1
     4
                           2
                                       1
                                                Medical
                                                                      1
     1465
                          23
                                       2
                                                Medical
     1466
                           6
                                       1
                                                Medical
                                                                      1
     1467
                                       3
                                          Life Sciences
                                                                      1
     1468
                           2
                                       3
                                                Medical
                                                                      1
     1469
                           8
                                                Medical
                                       3
           EmployeeNumber
                                 RelationshipSatisfaction StandardHours
                           . . .
     0
                         1
                            ...
                                                         1
                                                                        80
     1
                         2
                            . . .
                                                          4
                                                                        80
     2
                         4
                                                         2
                                                                        80
     3
                         5
                                                         3
                                                                        80
                            . . .
     4
                         7
                                                          4
                                                                        80
                            . . .
                            . . .
     1465
                      2061
                                                                        80
     1466
                      2062
                                                         1
                                                                        80
                            . . .
                                                         2
     1467
                      2064
                                                                        80
                            . . .
     1468
                      2065
                                                         4
                                                                        80
     1469
                      2068
                                                         1
                                                                        80
           StockOptionLevel TotalWorkingYears TrainingTimesLastYear
     0
                           0
                                               8
                                                                        0
     1
                           1
                                              10
                                                                        3
     2
                                                                        3
     3
                           0
                                               8
                                                                        3
     4
                           1
                                               6
                                                                        3
     1465
                                              17
                                                                        3
                                               9
                                                                        5
     1466
                           1
     1467
                                                                       a
                           1
                                               6
     1468
                           0
                                              17
                                                                        3
     1469
                           0
                                               6
                                                                        3
          {\tt WorkLifeBalance} \quad {\tt YearsAtCompany} \ {\tt YearsInCurrentRole}
     0
                         1
                                                              4
                         3
                                         10
                                                              7
     1
     2
```

```
import pandas as pd
df = pd.read_csv('/content/HR-Employee-Attrition.csv')
df['Attrition'] = df['Attrition'].map({'Yes': 1, 'No': 0})
print(df['Attrition'])
result_department = df.groupby('Department')['Attrition'].mean()
result_gender = df.groupby('Gender')['Attrition'].mean()
print("Mean Attrition by Department:")
print(result_department)
print("\nMean Attrition by Gender:")
print(result_gender)
percentage\_attrition = len(df[df['Attrition'] == 1]) / len(df) * 100
print(f"\nPercentage of Attrition: {percentage_attrition:.2f}%")
     0
     1
             0
             1
     3
             0
     1465
             0
     1466
             a
     1467
             a
     1468
             0
     1469
             0
     Name: Attrition, Length: 1470, dtype: int64
     Mean Attrition by Department:
     Department
     Human Resources
                                 0.190476
     Research & Development
                                 0.138398
     Sales
                                 0.206278
     Name: Attrition, dtype: float64
     Mean Attrition by Gender:
     Gender
     Female
                0.147959
     Male
                0.170068
     Name: Attrition, dtype: float64
     Percentage of Attrition: 16.12%
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
df = pd.read_csv("/content/HR-Employee-Attrition.csv")
age_bins = [18, 30, 40, 50, 60, 70, 80]
age_labels = ['18-29', '30-39', '40-49', '50-59', '60-69', '70-80']
df['AgeGroup'] = pd.cut(df['Age'], bins=age_bins, labels=age_labels, right=False)
categorical_values = ['EnvironmentSatisfaction', 'JobSatisfaction']
fig, axes = plt.subplots(nrows=1, ncols=2, figsize=(12, 10))
for i, var in enumerate(categorical_values):
    sns.boxplot(x='Attrition', y='Age', hue='AgeGroup', data=df, ax=axes[i])
    axes[i].set title(f'Boxplot for {var}')
plt.tight_layout()
plt.show()
```

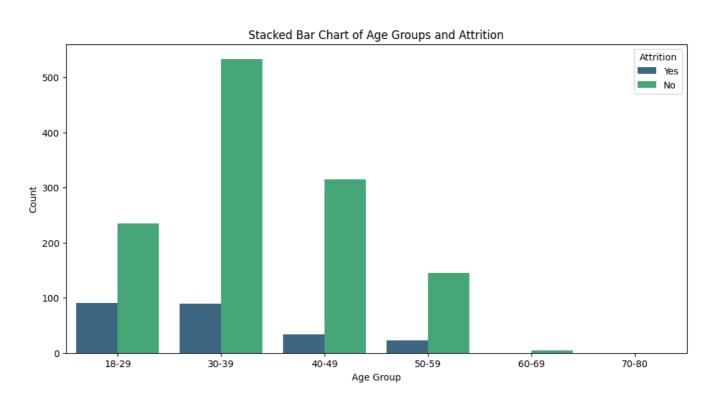


```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

df = pd.read_csv("/content/HR-Employee-Attrition.csv")

age_bins = [18, 30, 40, 50, 60, 70, 80]
age_labels = ['18-29', '30-39', '40-49', '50-59', '60-69', '70-80']

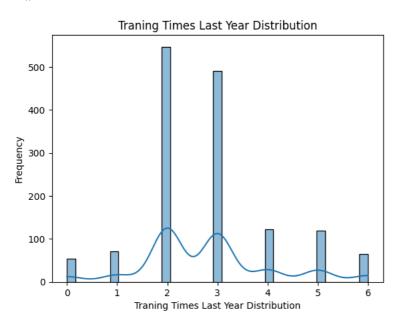
df['AgeGroup'] = pd.cut(df['Age'], bins=age_bins, labels=age_labels, right=False)
plt.figure(figsize=(12, 6))
sns.countplot(x='AgeGroup', hue='Attrition', data=df, palette='viridis')
plt.title('Stacked Bar Chart of Age Groups and Attrition')
plt.xlabel('Age Group')
plt.ylabel('Count')
plt.legend(title='Attrition', loc='upper right')
plt.show()
```



import pandas as pd

```
import seaborn as sns
import matplotlib.pyplot as plt
df=pd.read_csv('/content/HR-Employee-Attrition.csv')
sns.histplot(df['TrainingTimesLastYear'],kde=True)
plt.xlabel('Training Times Last Year Distribution')
plt.ylabel('Frequency')
plt.title('Training Times Last Year Distribution')
plt.show()
```





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

df = pd.read_csv('/content/HR-Employee-Attrition.csv')

df['Attrition_binary'] = df['Attrition'].map({'Yes': 1, 'No': 0})

attrition_data = df.groupby('YearsAtCompany')['Attrition_binary'].mean().reset_index()

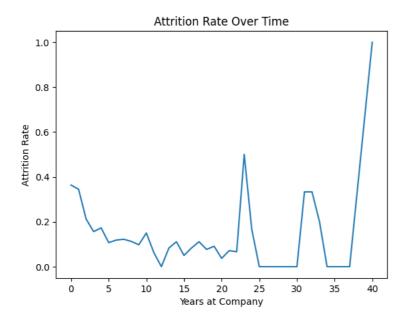
sns.lineplot(x='YearsAtCompany', y='Attrition_binary', data=attrition_data)

plt.xlabel('Years at Company')

plt.ylabel('Attrition Rate')

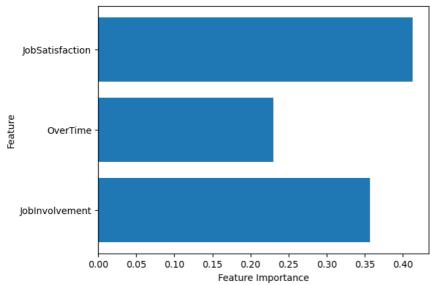
plt.title('Attrition Rate Over Time')

plt.show()
```



```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import mean_squared_error
import matplotlib.pyplot as plt
data = pd.read_csv('/content/HR_Analytics.csv')
data['OverTime'] = data['OverTime'].map({'Yes': 1, 'No': 0})
data = data[['JobInvolvement', 'JobSatisfaction','OverTime', 'PerformanceRating']]
categorical_cols = data.select_dtypes(include=['object']).columns.tolist()
data = pd.get_dummies(data, columns=categorical_cols)
X = data[['JobInvolvement','OverTime', 'JobSatisfaction']]
y = data['PerformanceRating']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
model = RandomForestRegressor(n_estimators=100, random_state=42)
model.fit(X_train, y_train)
y_pred = model.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
print(f"Mean Squared Error: {mse}")
feature_importances = model.feature_importances_
plt.barh(X.columns, feature_importances)
plt.xlabel("Feature Importance")
plt.ylabel("Feature")
plt.show()
```

Mean Squared Error: 0.13239068774206364



```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, confusion_matrix
import matplotlib.pyplot as plt
data = pd.read_csv('/content/HR_Analytics.csv')
data['Attrition'] = data['Attrition'].map({'Yes': 1, 'No': 0})
data['OverTime'] = data['OverTime'].map({'Yes': 1, 'No': 0})
selected_columns = ['JobInvolvement', 'JobSatisfaction', 'OverTime', 'PerformanceRating', 'Attrition']
data = data[selected_columns]
categorical_cols = ['JobInvolvement', 'JobSatisfaction','OverTime', 'PerformanceRating']
data = pd.get_dummies(data, columns=categorical_cols)
X = data.drop('Attrition', axis=1)
y = data['Attrition']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
model = RandomForestClassifier(n_estimators=100, random_state=42)
model.fit(X_train, y_train)
y pred = model.predict(X test)
accuracy = accuracy_score(y_test, y_pred)
conf_matrix = confusion_matrix(y_test, y_pred)
print(f"Accuracy: {accuracy}")
print("Confusion Matrix:")
print(conf_matrix)
feature_importances = model.feature_importances_
feature_importance_df = pd.DataFrame({'Feature': X.columns, 'Importance': feature_importances})
print(feature importance df.sort values('Importance', ascending=False)
feature_importance_df = feature_importance_df.sort_values(by='Importance', ascending=False)
import seaborn as sns
plt.figure(figsize=(10, 6))
sns.barplot(x='Importance', y='Feature', data=feature_importance_df, palette='viridis')
plt.xlabel("Feature Importance")
plt.ylabel("Feature")
plt.show()
     Accuracy: 0.847972972972973
     Confusion Matrix:
     [[247 2]
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, confusion_matrix
import matplotlib.pyplot as pt
data = pd.read_csv('/content/HR_Analytics.csv')
selected_columns = ['Age', 'Department', 'Education', 'EnvironmentSatisfaction', 'JobInvolvement', 'OverTime']
categorical_cols = ['Department', 'Age']
X = data[selected_columns[:-1]]
y = data['Attrition']
X = pd.get_dummies(X, columns=categorical_cols)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)
model = RandomForestClassifier(random_state=42)
model.fit(X_train, y_train)
def predict_attrition(employee_id, model, data, X_columns):
    if employee_id not in data['EmpID'].values:
        return "Employee ID not found in the dataset."
```