```
from google.colab import files
import io
# Step 1: Upload the dataset
uploaded = files.upload()
# Get the uploaded file name
file name = list(uploaded.keys())[0]
# Step 2: Load the dataset
df = pd.read_excel(io.BytesIO(uploaded[file_name]), sheet_name='autism')
# Step 3: Display missing values before processing
print("Missing values before processing:\n", df.isnull().sum())
# Step 4: Count "?" values before removal
print("\nCount of '?' in ethnicity:", (df['ethnicity'] == '?').sum())
print("Count of '?' in relation:", (df['relation'] == '?').sum())
# Step 5: Remove rows where 'ethnicity' or 'relation' contains "?"
df = df[(df['ethnicity'] != '?') & (df['relation'] != '?')]
# Step 6: Remove rows where 'age' is NaN
df = df.dropna(subset=['age'])
# Step 7: Remove rows where 'age' is greater than 80
df = df[df['age'] <= 80]</pre>
# Step 8: Display missing values after processing
print("\nMissing values after processing:\n", df.isnull().sum())
# Step 9: Save the cleaned dataset
cleaned file name = "cleaned autism.xlsx"
df.to excel(cleaned file name, index=False)
# Step 10: Download the cleaned dataset
files.download(cleaned_file_name)
print("\nDownload started: 'cleaned_autism.xlsx'")
Choose Files No file chosen
                                       Upload widget is only available when the cell has been executed in
    the current browser session. Please rerun this cell to enable.
import pandas as pd
import numpy as np
from sklearn.model selection import train test split
from sklearn.preprocessing import LabelEncoder, StandardScaler
from sklearn.svm import SVC
from sklearn.linear model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy score, classification report, confusion matrix
from imblearn.over_sampling import SMOTE
# Load the cleaned dataset
df = pd.read excel("cleaned autism.xlsx")
```

import pandas as pd

```
# Encode categorical variables
categorical_columns = ['gender', 'ethnicity', 'jundice', 'austim', 'contry_of_res', 'used_a
label encoders = {}
for col in categorical columns:
    le = LabelEncoder()
    df[col] = le.fit transform(df[col])
    label encoders[col] = le
# Define features (X) and target variable (y)
X = df.drop(columns=['Class/ASD'])
y = df['Class/ASD']
# Standardize numerical features
scaler = StandardScaler()
X = scaler.fit_transform(X)
# Split data into training (70%) and testing (30%)
X_train, X_temp, y_train, y_temp = train_test_split(X, y, test_size=0.3, random_state=42)
X_val, X_test, y_val, y_test = train_test_split(X_temp, y_temp, test_size=0.5, random_state
# Apply SMOTE
smote = SMOTE(random state=42)
X_train, y_train = smote.fit_resample(X_train, y_train)
svm_model = SVC(kernel='rbf', C=0.5, gamma=0.01, class_weight='balanced')
svm_model.fit(X_train, y_train)
y test pred svm = svm model.predict(X test)
print("SVM Testing Accuracy:", accuracy_score(y_test, y_test_pred_svm))
print("\nSVM Classification Report:\n", classification report(y test, y test pred svm))
print("\nSVM Confusion Matrix:\n", confusion_matrix(y_test, y_test_pred_svm))
→ SVM Testing Accuracy: 0.9782608695652174
    SVM Classification Report:
                                  recall f1-score
                     precision
                                                       support
               NO
                         0.98
                                    0.98
                                              0.98
                                                           66
              YES
                         0.96
                                    0.96
                                              0.96
                                                           26
                                              0.98
                                                           92
         accuracy
                                              0.97
                                                           92
        macro avg
                         0.97
                                    0.97
```

SVM Confusion Matrix:

0.98

0.98

0.98

92

[[65 1] [ 1 25]]

weighted avg