```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.metrics import classification_report, accuracy_score
from sklearn.preprocessing import StandardScaler
```

# Load the dataset
dfmpd.read\_csv('alzheimers\_disease\_data.csv')
df.head()



	PatientiD	Age	Gender	Ethnicity	EducationLevel	BMI	Smoking	AlcoholConsumption	PhysicalActivity	DietQuality	***	MemoryComplaints	BehavioralProbl
0	4751	73	0	0	2	22.927749	0	13,297218	6,327112	1.347214	***	0	
1	4752	89	0	0	0	26.827681	0	4.542524	7.619885	0.518767	100	0	
2	4753	73	0	3	1	17.795882	0	19.555085	7.844988	1.826335	-	0	
3	4754	74	1	0	1	33.800817	1	12.209266	8.428001	7,435604	***	0	
4	4755	89	0	0	0	20.716974	0	18.454356	6.310461	0.795498		0	

## df.ismull().sum()

PatientID 0 Age Gender 8 8 Ethnicity EducationLevel BMI Smoking AlcoholConsumption PhysicalActivity DietQuality SleepQuality FamilyMistoryAlzheimers CardiovascularDisease Diabetes Depression HeadInjury Hypertension SystolicBP DiastolicBP CholesterolTotal CholesterolLDL ChalesterolHDL CholesterolTriglycerides NISE FunctionalAssessment HemoryComplaints BehavioralProblems Confusion Disorientation PersonalityChanges DifficultyCompletingTasks Forgetfulness Diagnosis DoctorInCharge dtype: int64

```
: df=df.drop('DoctorInCharge', axis=1)
  df.head()
     PatientID Age Gender Ethnicity EducationLevel
                                                          BMI Smoking AlcoholConsumption PhysicalActivity DietQuality ... FunctionalAssessment MemoryComp
         4751
                 73
                                   0
                                                  2 22.927749
                                                                                   13.297218
                                                                                                    6.327112
                                                                                                                1.347214 ...
                                                                                                                                        6.518877
                          O
                                                                      0
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         4752
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                          Ö
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         4754
                       1
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                                                                      1
                                                                                                                7.435604 ...
                                                                                                                                        8.965106
               74
                                                  1 33,800817
                                                                                   12.209266
                                                                                                    8.428001
                                                                      0
  4
         4755
                89
                          0
                                   0
                                                  0 20.716974
                                                                                    18.454356
                                                                                                    6.310461
                                                                                                               0.795498 ...
                                                                                                                                        6.045039
 5 rows × 34 columns
 4
X = df.drop(columns=['Diagnosis'])
  y = df['Diagnosis']
: # Scale the features
  scaler * StandardScaler()
  X_scaled = scaler.fit_transform(X)
# Split the dataset
  \label{eq:continuous} $$X_{\text{train}}$, $X_{\text{test}}$, $y_{\text{train}}$, $y_{\text{test}}=8.3$, $random_{\text{state}}=42$)$
: clf = SVC(kernel='linear')
  clf.fit(X_train, y_train)
          SVC
  SVC(kernel='linear')
```

```
y_pred = clf.predict(X_test)
accuracy accuracy_score(y_test, y_pred)
    print(f'Accuracy: (accuracy:.2f)')
    print(classification_report(y_test,y_pred))
    Accuracy: 0.81
                    precision
                                   recall f1-score support
                          0.82
                                      0.89
                          0.79
                                      0.68
                                                   0.73
                                                                244
                                                   0.81
                                                                645
         accuracy
                         0.81
                                      0.79
                                                   0.79
                                                                545
    weighted avg
                       0.81
                                      0.81
                                                   0.81
                                                                645
]: new_value = [[72, 0, 0, 1, 21.8112234230586, 0, 3.70465344544926, 3.36545201815567, 4.51379891620157, 7.29482034608182, 0, 1, 0, 0, 0, 0, 137, 75, 231.54541225488, 63.4110356567039, 30.4597138210004, 281.094893298805, 8.65799972593658, 6.87802469048415, 1, 0, 0.666181587511921,
                       0, 0, 0, 0, 1,1]]
): # Scale the new sample using the same scaler
    new_value_scaled = scaler.transform(new_value)
    # Predict the diagnosis for the new sample
    new_prediction = clf.predict(new_value_scaled)
    print("Prediction for new sample:", new_prediction)
    Prediction for new sample: [1]
```