

```
In [308... import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.feature_selection import SelectKBest, SelectPercentile, mutual_info_cl
from sklearn.ensemble import AdaBoostClassifier
from sklearn.svm import SVC
from sklearn.neighbors import KNeighborsClassifier
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import GridSearchCV
from sklearn.metrics import confusion_matrix, classification_report
from sklearn.metrics import ConfusionMatrixDisplay
from sklearn.metrics import accuracy_score
```

```
In [309... df=pd.read_csv('SaYoPillow.csv')
```

```
In [310... df.head()
```

```
Out[310...
   snoring  respiration  body  limb  blood  eye  sleeping  heart  str
   rate      rate      temperature  movement  oxygen  movement  hours  rate  le
0    93.80      25.680      91.840      16.600      89.840      99.60      1.840  74.20
1    91.64      25.104      91.552      15.880      89.552      98.88      1.552  72.76
2    60.00      20.000      96.000      10.000      95.000      85.00      7.000  60.00
3    85.76      23.536      90.768      13.920      88.768      96.92      0.768  68.84
4    48.12      17.248      97.872       6.496      96.248      72.48      8.248  53.12
```

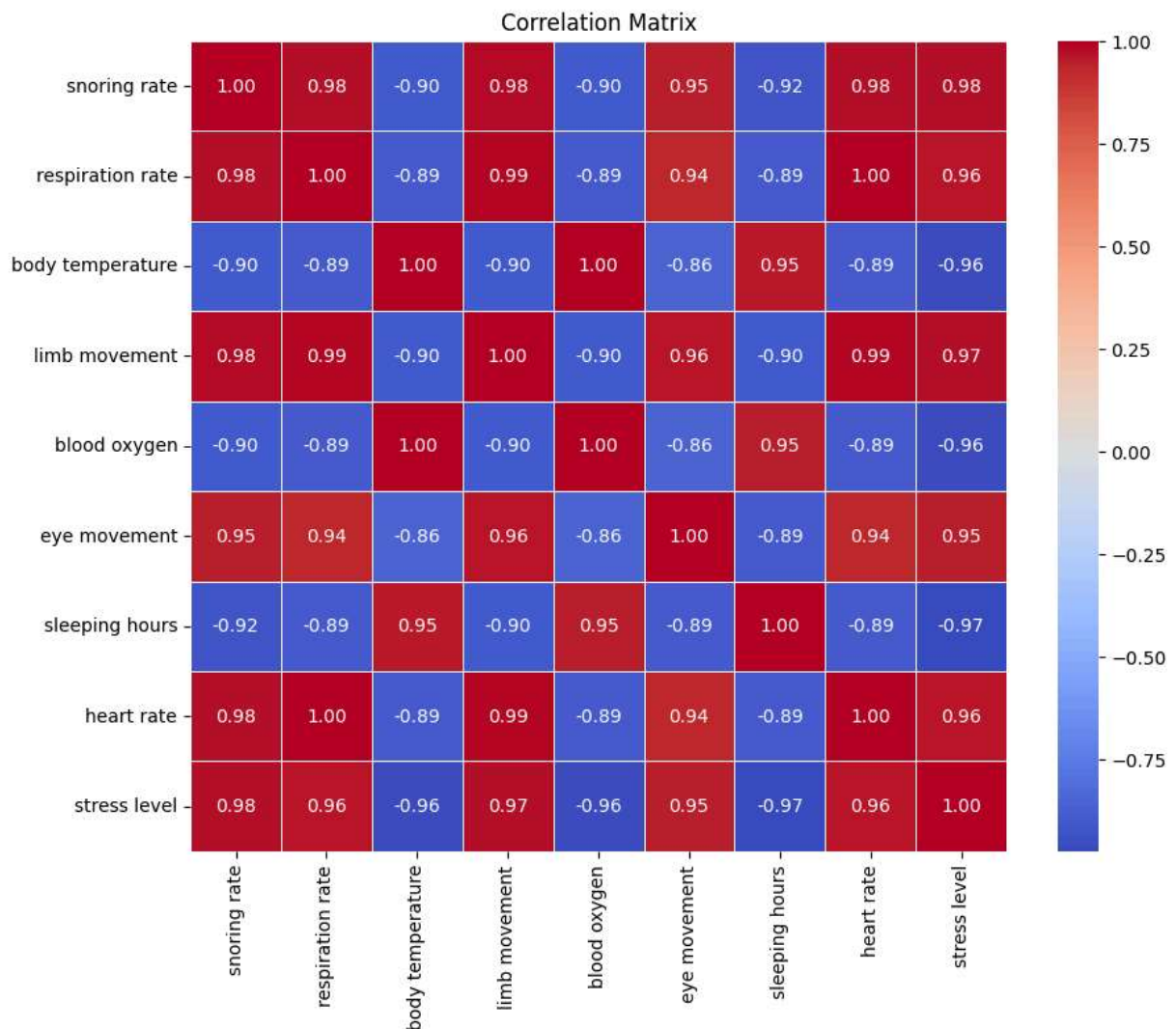
Missing values

```
In [311... df.isnull().sum()
```

```
Out[311... snoring rate      0
respiration rate  0
body temperature  0
limb movement     0
blood oxygen      0
eye movement      0
sleeping hours    0
heart rate        0
stress level      0
dtype: int64
```

Correlation Matrix

```
In [312... corr = df.corr()
plt.figure(figsize=(10, 8))
sns.heatmap(corr, annot=True, cmap='coolwarm', fmt=".2f", linewidths=.5)
plt.title('Correlation Matrix')
plt.show()
```

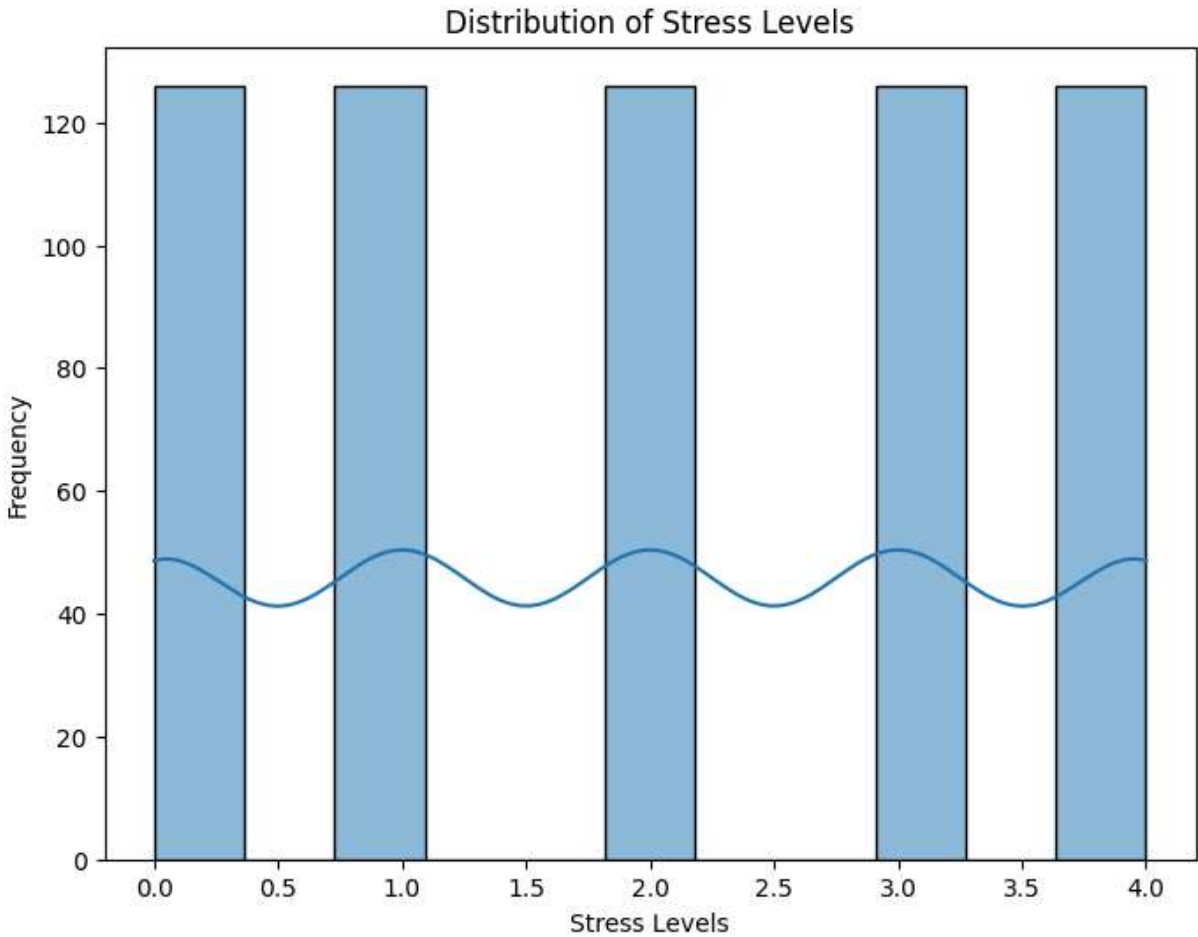


Stress Level Counts

```
In [313... # Stress Level Counts
stress_level_counts = df['stress level'].value_counts()
print("Stress Level Counts:")
print(stress_level_counts)

# Visualize the distribution of Stress Levels
plt.figure(figsize=(8, 6))
sns.histplot(df['stress level'], kde=True)
plt.title('Distribution of Stress Levels')
plt.xlabel('Stress Levels')
plt.ylabel('Frequency')
plt.show()
```

Stress Level Counts:
stress level
3 126
1 126
0 126
2 126
4 126
Name: count, dtype: int64



```
In [314... from sklearn.model_selection import train_test_split
x=df.iloc[:, :-1]
y=df.iloc[:, -1]
```

```
In [315... x.head()
```

Out[315...

	snoring rate	respiration rate	body temperature	limb movement	blood oxygen	eye movement	sleeping hours	heart rate
0	93.80	25.680	91.840	16.600	89.840	99.60	1.840	74.20
1	91.64	25.104	91.552	15.880	89.552	98.88	1.552	72.76
2	60.00	20.000	96.000	10.000	95.000	85.00	7.000	60.00
3	85.76	23.536	90.768	13.920	88.768	96.92	0.768	68.84
4	48.12	17.248	97.872	6.496	96.248	72.48	8.248	53.12

```
In [316... xtrain,xtest,ytrain,ytest= train_test_split(x,y,test_size=0.20)
```

Feature Selection 1

```
In [317... from sklearn.feature_selection import SelectKBest
from sklearn.feature_selection import mutual_info_classif
```

```
In [318... kbest = SelectKBest(mutual_info_classif, k=5)
select_feature = kbest.fit(xtrain ,ytrain)
```

```
In [319... selected_features = xtrain.columns[select_feature.get_support()]
```

```
In [320... xtrain.head()
```

```
Out[320...
      snoring  respiration      body      limb  blood      eye  sleeping  heart
      rate    rate  temperature movement oxygen movement hours    rate
561  49.480    17.792    98.688    7.584  96.792    77.92    8.792    54.48
236  88.640    24.304    91.152   14.880  89.152    97.88    1.152    70.76
426  58.720    19.744    95.744    9.744  94.616    84.36    6.744    59.36
148  45.120    16.048    96.072    4.096  95.048    60.48    7.048    50.12
364  99.072    29.072    88.840   18.536  86.608   103.84    0.000    82.68
```

```
In [321... x1=df[selected_features]
```

```
In [322... x1train,x1test,y1train,y1test= train_test_split(x1,y,test_size=.20)
```

```
In [323... model1=AdaBoostClassifier()
model1.fit(x1train,y1train)
```

```
Out[323... ▼ AdaBoostClassifier
AdaBoostClassifier()
```

```
In [324... print('train score',model1.score(x1train,y1train))
print('test score',model1.score(x1test,y1test))
```

```
train score 0.998015873015873
test score 0.9603174603174603
```

```
In [352... from sklearn.metrics import confusion_matrix

# Confusion matrix for Model 1
conf_matrix_model1 = confusion_matrix(ytest, model1.predict(x1test))
print("Confusion Matrix (Model 1):")
print(conf_matrix_model1)
```

Confusion Matrix (Model 1):

```
[[6 1 4 7 1]
 [5 8 3 6 4]
 [3 4 7 7 6]
 [9 5 6 3 7]
 [1 7 4 6 6]]
```

```
In [325... from sklearn.metrics import classification_report

# Make predictions on the test set
y1pred = model1.predict(x1test)

# Generate classification report
print("Classification Report:")
print(classification_report(y1test, y1pred, zero_division=1))
```

Classification Report:

	precision	recall	f1-score	support
0	1.00	0.92	0.96	26
1	0.92	1.00	0.96	23
2	0.96	1.00	0.98	23
3	0.93	0.96	0.95	28
4	1.00	0.92	0.96	26
accuracy			0.96	126
macro avg	0.96	0.96	0.96	126
weighted avg	0.96	0.96	0.96	126

Feature Selection 2

```
In [326... from sklearn.feature_selection import SelectPercentile
from sklearn.feature_selection import mutual_info_classif
```

```
In [327... sp = SelectPercentile(mutual_info_classif, percentile=50)
select_feature2 = sp.fit(xtrain, ytrain)
```

```
In [328... selected_features2 = xtrain.columns[select_feature2.get_support()]
```

```
In [329... x2 = df[selected_features2]
```

```
In [330... x2train, x2test, y2train, y2test = train_test_split(x2, y, test_size=.20)
```

```
In [331... model2 = AdaBoostClassifier()
model2.fit(x2train, y2train)
```

```
Out[331... ▾ AdaBoostClassifier
AdaBoostClassifier()
```

```
In [332... print('train score',model2.score(x2train,y2train))
print('test score',model2.score(x2test,y2test))
```

```
train score 0.6150793650793651
test score 0.5396825396825397
```

```
In [353... # Confusion matrix for Model 2
conf_matrix_model2 = confusion_matrix(ytest, model2.predict(x2test))
print("\nConfusion Matrix (Model 2):")
print(conf_matrix_model2)
```

```
Confusion Matrix (Model 2):
```

```
[[ 4  0  0  8  7]
 [14  0  0  9  3]
 [15  0  0  9  3]
 [15  0  0 10  5]
 [12  0  0  8  4]]
```

```
In [333... from sklearn.metrics import classification_report

# Make predictions on the test set
y2pred = model2.predict(x2test)

# Generate classification report
print("Classification Report:")
print(classification_report(y2test, y2pred, zero_division=1))
```

```
Classification Report:
```

	precision	recall	f1-score	support
0	0.43	1.00	0.60	26
1	1.00	0.00	0.00	33
2	1.00	0.00	0.00	25
3	0.45	1.00	0.62	20
4	1.00	1.00	1.00	22
accuracy			0.54	126
macro avg	0.78	0.60	0.45	126
weighted avg	0.80	0.54	0.40	126

Feature importance

```
In [334... feature_importance = model2.feature_importances_
feature_importance = pd.DataFrame(feature_importance, columns = ['Importance'])
```

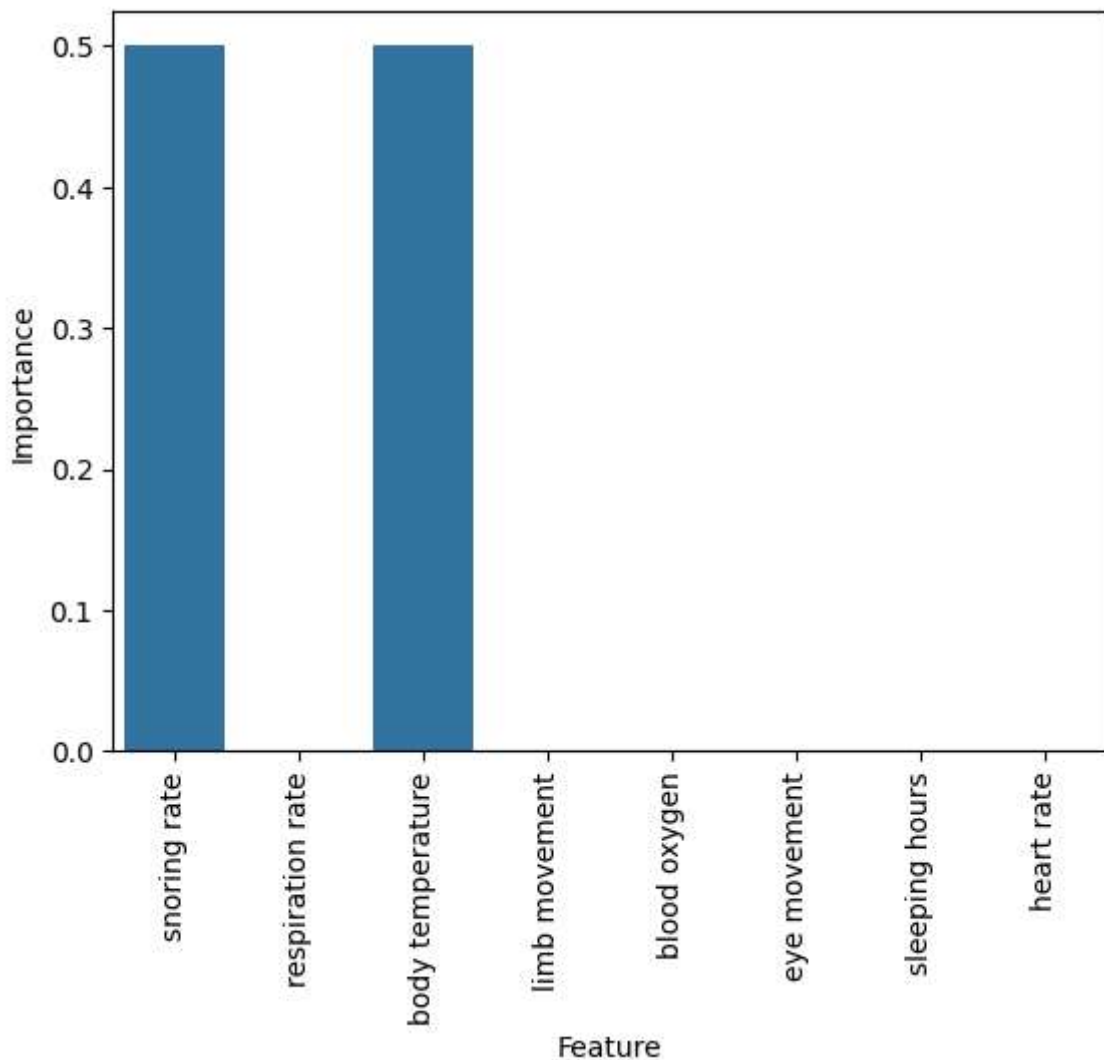
```
In [335... features = pd.DataFrame(xtrain.columns, columns = ['Feature'])
```

```
In [336... importance_df = pd.concat([features, feature_importance],axis=1)
```

```
In [337... print(importance_df)
```

	Feature	Importance
0	snoring rate	0.5
1	respiration rate	0.0
2	body temperature	0.5
3	limb movement	0.0
4	blood oxygen	NaN
5	eye movement	NaN
6	sleeping hours	NaN
7	heart rate	NaN

```
In [338... sns.barplot(x=importance_df['Feature'], y=importance_df['Importance'])  
plt.xticks(rotation=90) # Rotate x-axis labels  
plt.show()
```



```
In [339... # Sort features by importance  
importance_df_sorted = importance_df.sort_values(by='Importance', ascending=False)  
  
# Select top N features based on importance  
top_n_features = 2 # Specify the number of top features you want to select  
selected_features = importance_df_sorted['Feature'][:top_n_features]
```

```
In [340... # Create x3 DataFrame using the selected features
x3 = df[selected_features]
```

```
In [341... x3.head()
```

```
Out[341...      snoring rate  body temperature
0           93.80           91.840
1           91.64           91.552
2           60.00           96.000
3           85.76           90.768
4           48.12           97.872
```

```
In [342... x3train,x3test,y3train,y3test = train_test_split(x3,y,test_size=.20)
```

```
In [343... model3=AdaBoostClassifier()
model3.fit(x3train,y3train)
```

```
Out[343... ▼ AdaBoostClassifier
AdaBoostClassifier()
```

```
In [344... print('train score',model3.score(x3train,y3train))
print('test score',model3.score(x3test,y3test))
```

```
train score 0.6091269841269841
test score 0.5634920634920635
```

```
In [354... # Confusion matrix for Model 3
conf_matrix_model3 = confusion_matrix(y3test, model3.predict(x3test))
print("\nConfusion Matrix (Model 3):")
print(conf_matrix_model3)
```

```
Confusion Matrix (Model 3):
[[27  0  0  0  0]
 [ 1  0 26  0  0]
 [ 0  0 25  0  0]
 [ 0  0 28  0  0]
 [ 0  0  0  0 19]]
```

```
In [345... from sklearn.metrics import classification_report
```

```
# Make predictions on the test set
y3pred = model3.predict(x3test)
```

```
# Generate classification report
print("Classification Report:")
print(classification_report(y3test, y3pred, zero_division=1))
```


Classification Report:

	precision	recall	f1-score	support
0	0.96	1.00	0.98	27
1	1.00	0.00	0.00	27
2	0.32	1.00	0.48	25
3	1.00	0.00	0.00	28
4	1.00	1.00	1.00	19
accuracy			0.56	126
macro avg	0.86	0.60	0.49	126
weighted avg	0.86	0.56	0.46	126

SVM

In [346... `from sklearn import svm`

In [347... `classifier=svm.SVC(kernel='linear',gamma='auto',C=3)`
`classifier.fit(xtrain,ytrain)`

Out[347... `SVC`
`SVC(C=3, gamma='auto', kernel='linear')`

In [348... `print('train score',classifier.score(xtrain,ytrain))`
`print('test score',classifier.score(xtest,ytest))`

train score 1.0
test score 1.0

In [355... `# Confusion matrix for SVM`
`conf_matrix_svm = confusion_matrix(ytest, classifier.predict(xtest))`
`print("\nConfusion Matrix (SVM):")`
`print(conf_matrix_svm)`

Confusion Matrix (SVM):
[[19 0 0 0 0]
[0 26 0 0 0]
[0 0 27 0 0]
[0 0 0 30 0]
[0 0 0 0 24]]

In [349... `from sklearn.metrics import classification_report`

`# Make predictions on the test set`
`ypred = classifier.predict(xtest)`

`# Generate classification report`
`print("Classification Report:")`
`print(classification_report(ytest, ypred, zero_division=1))`

Classification Report:

	precision	recall	f1-score	support
0	1.00	1.00	1.00	19
1	1.00	1.00	1.00	26
2	1.00	1.00	1.00	27
3	1.00	1.00	1.00	30
4	1.00	1.00	1.00	24
accuracy			1.00	126
macro avg	1.00	1.00	1.00	126
weighted avg	1.00	1.00	1.00	126

Prediction

```
In [350... # Example with values for each feature
manual_example = np.array([[20, 18, 37, 2, 95, 1, 7, 85]]) # Assuming 8 features a

# Display the manual example
print("manual_example:")
print(manual_example)
```

```
manual_example:
[[20 18 37  2 95  1  7 85]]
```

```
In [351... # Prediction using Feature Selection 1
selected_features_model1 = x1.columns # Features used in model1
manual_example_model1_df = pd.DataFrame(manual_example, columns=x.columns) # Conve
manual_example_model1 = manual_example_model1_df[selected_features_model1] # Selec
prediction_model1 = model1.predict(manual_example_model1)
print("Prediction (Feature Selection 1):", prediction_model1)

# Prediction using Feature Selection 2
selected_features_model2 = x2.columns # Features used in model2
manual_example_model2_df = pd.DataFrame(manual_example, columns=x.columns) # Conve
manual_example_model2 = manual_example_model2_df[selected_features_model2] # Selec
prediction_model2 = model2.predict(manual_example_model2)
print("Prediction (Feature Selection 2):", prediction_model2)

# Prediction using Feature Importance (Model 3)
selected_features_model3 = x3.columns # Features used in model3
manual_example_model3_df = pd.DataFrame(manual_example, columns=x.columns) # Conve
manual_example_model3 = manual_example_model3_df[selected_features_model3] # Selec
prediction_model3 = model3.predict(manual_example_model3)
print("Prediction (Feature Importance):", prediction_model3)

# Prediction using SVM
prediction_svm = classifier.predict(manual_example_df) # No need to modify example
print("Prediction (SVM):", prediction_svm)
```

```
Prediction (Feature Selection 1): [2]  
Prediction (Feature Selection 2): [0]  
Prediction (Feature Importance): [0]  
Prediction (SVM): [0]
```