

```
In [56]: import shap
import pandas as pd
import numpy as np
from xgboost import XGBClassifier
import matplotlib.pyplot as plt
import seaborn as sns

from sklearn.compose import ColumnTransformer
from sklearn.model_selection import StratifiedKFold, GridSearchCV,
train_test_split
from sklearn.preprocessing import OneHotEncoder, StandardScaler
from sklearn.metrics import accuracy_score, precision_score, recall_score,
f1_score, confusion_matrix, roc_auc_score
```

```
In [57]: df = pd.read_csv('table.csv')
# replace C with 1 and P with 0
df['Label'] = df['Label'].replace({'C': 1, 'P': 0})
df
```

C:\Users\user\AppData\Local\Temp\ipykernel\_13268\3825214023.py:3: FutureWarning: Downcasting behavior in `replace` is deprecated and will be removed in a future version. To retain the old behavior, explicitly call `result.infer\_objects(copy=False)`. To opt-in to the future behavior, set `pd.set\_option('future.no\_silent\_downcasting', True)`

```
df['Label'] = df['Label'].replace({'C': 1, 'P': 0})
```

```
Out[57]:
```

	mean_value_ML	mean_value_AP	mean_distance_ML	mean_distance_AP	mean_distance_ML
0	0.046123	-1.461512	0.385651	0.543167	
1	0.042500	-0.365777	0.532939	0.484075	
2	0.496358	-1.401023	0.364302	0.400104	
3	0.314393	-0.549541	0.573516	0.486970	
4	1.412529	0.186249	0.929037	1.094830	
...	...	...	...	...	...
100	2.605348	-2.030942	0.335179	0.524191	
101	0.776783	1.222163	0.541288	0.659628	
102	-0.708497	-0.800798	0.466417	0.711986	
103	0.268076	-1.346882	0.930665	0.949494	
104	3.524215	-1.546033	0.291584	0.580360	

105 rows × 73 columns

```
In [58]: X = df.drop(['Label'], axis=1)
y = df['Label']
```

```
In [59]: cv = StratifiedKFold(n_splits=5, shuffle=True, random_state=42)

for i, (train_index, test_index) in enumerate(cv.split(X, y)):
    print(f'Results for split {i+1}')
```

```
X_train, X_test = X.iloc[train_index], X.iloc[test_index]
y_train, y_test = y.iloc[train_index], y.iloc[test_index]

clf = XGBClassifier(eval_metric='logloss', random_state=42)
clf.fit(X_train, y_train)

y_pred = clf.predict(X_test)

accuracy = accuracy_score(y_test, y_pred)
print(f'Accuracy: {accuracy:.4f}')

precision = precision_score(y_test, y_pred)
print(f'Precision: {precision:.4f}')

recall = recall_score(y_test, y_pred)
print(f'Recall: {recall:.4f}')

f1 = f1_score(y_test, y_pred)
print(f'F1 Score: {f1:.4f}')

cm = confusion_matrix(y_test, y_pred)
sns.heatmap(cm, annot=True, fmt="d", cmap="Blues", xticklabels=['Predicted
0', 'Predicted 1'], yticklabels=['Actual 0', 'Actual 1'])
plt.title('Confusion Matrix')
plt.show()

explainer = shap.TreeExplainer(clf)
shap_values = explainer.shap_values(X_test)

shap.summary_plot(shap_values, X_test)
```

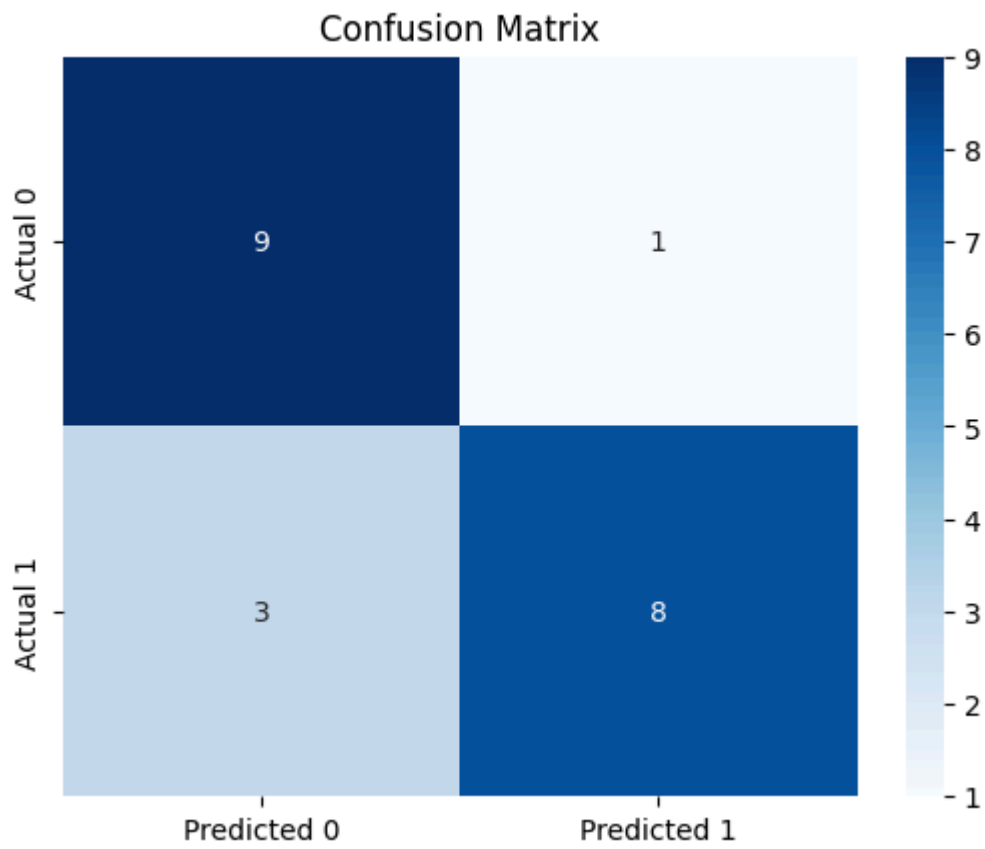
Results for split 1

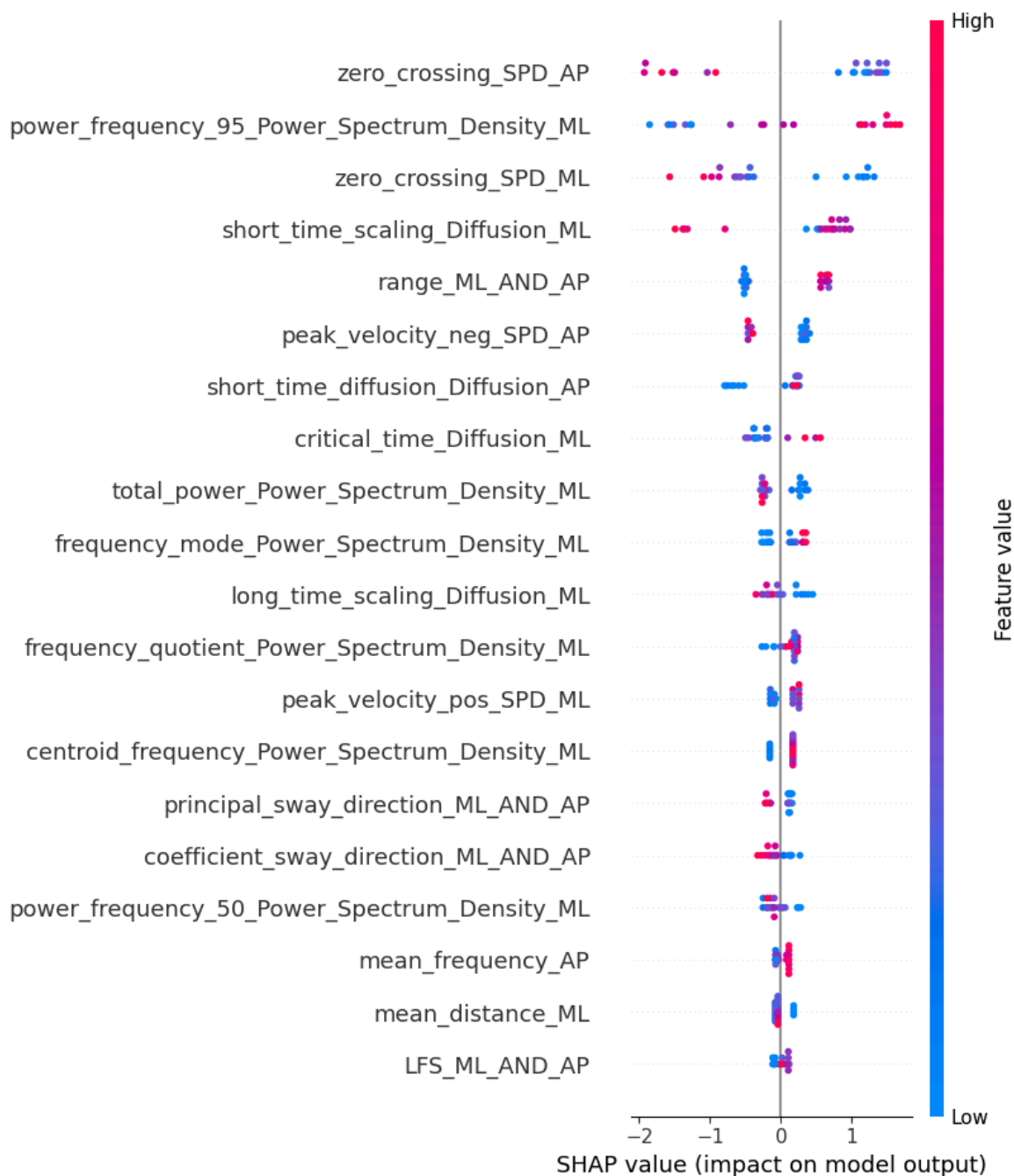
Accuracy: 0.8095

Precision: 0.8889

Recall: 0.7273

F1 Score: 0.8000





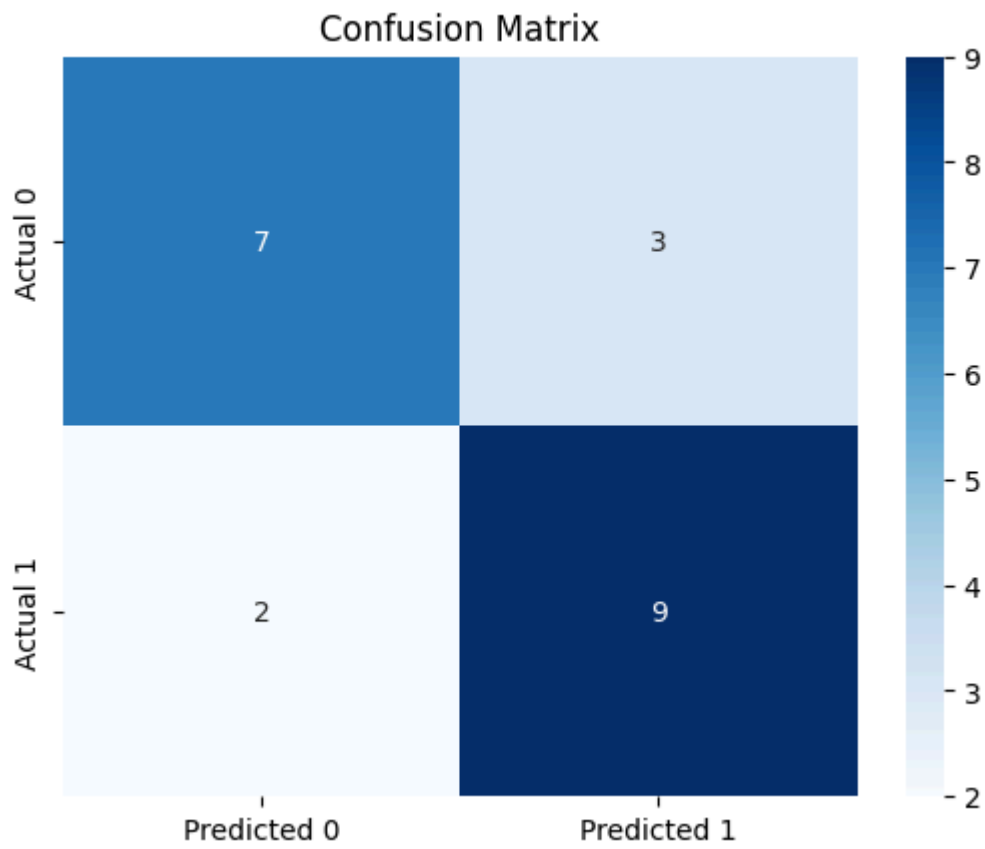
Results for split 2

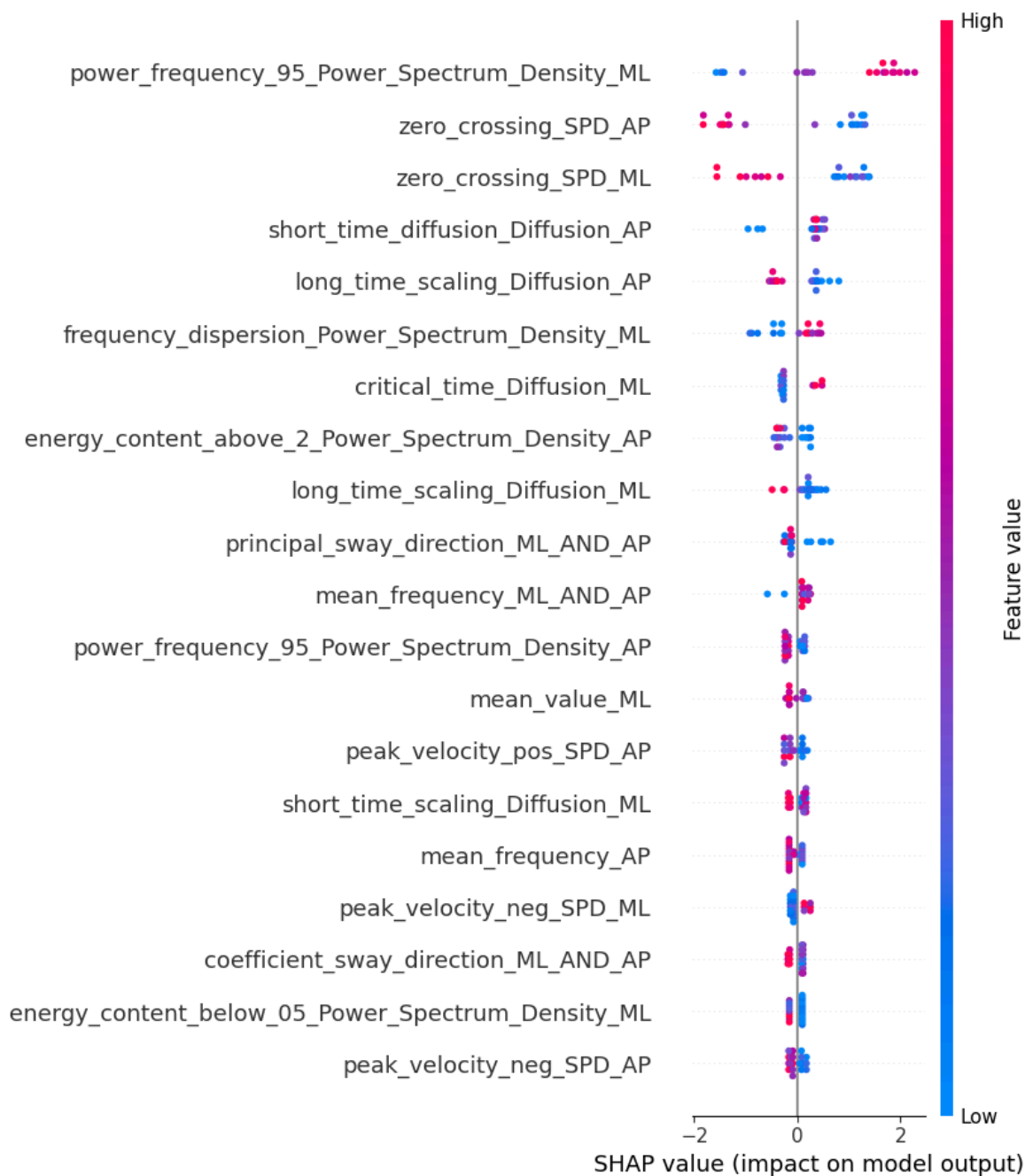
Accuracy: 0.7619

Precision: 0.7500

Recall: 0.8182

F1 Score: 0.7826





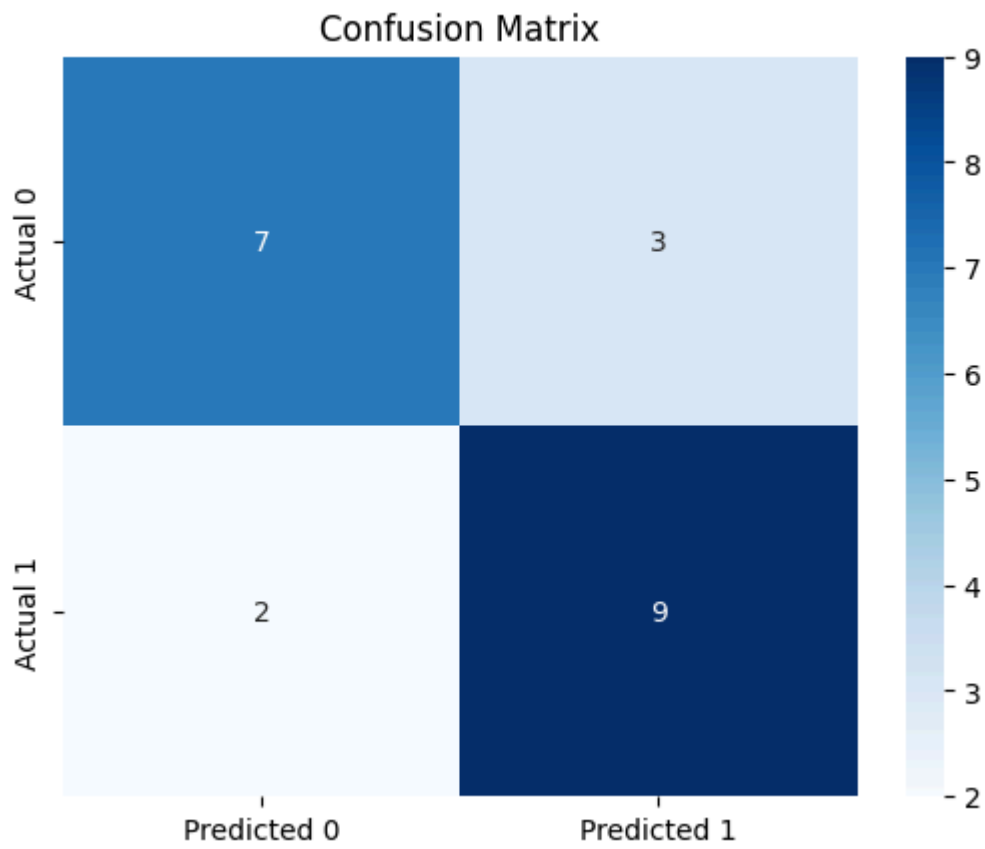
Results for split 3

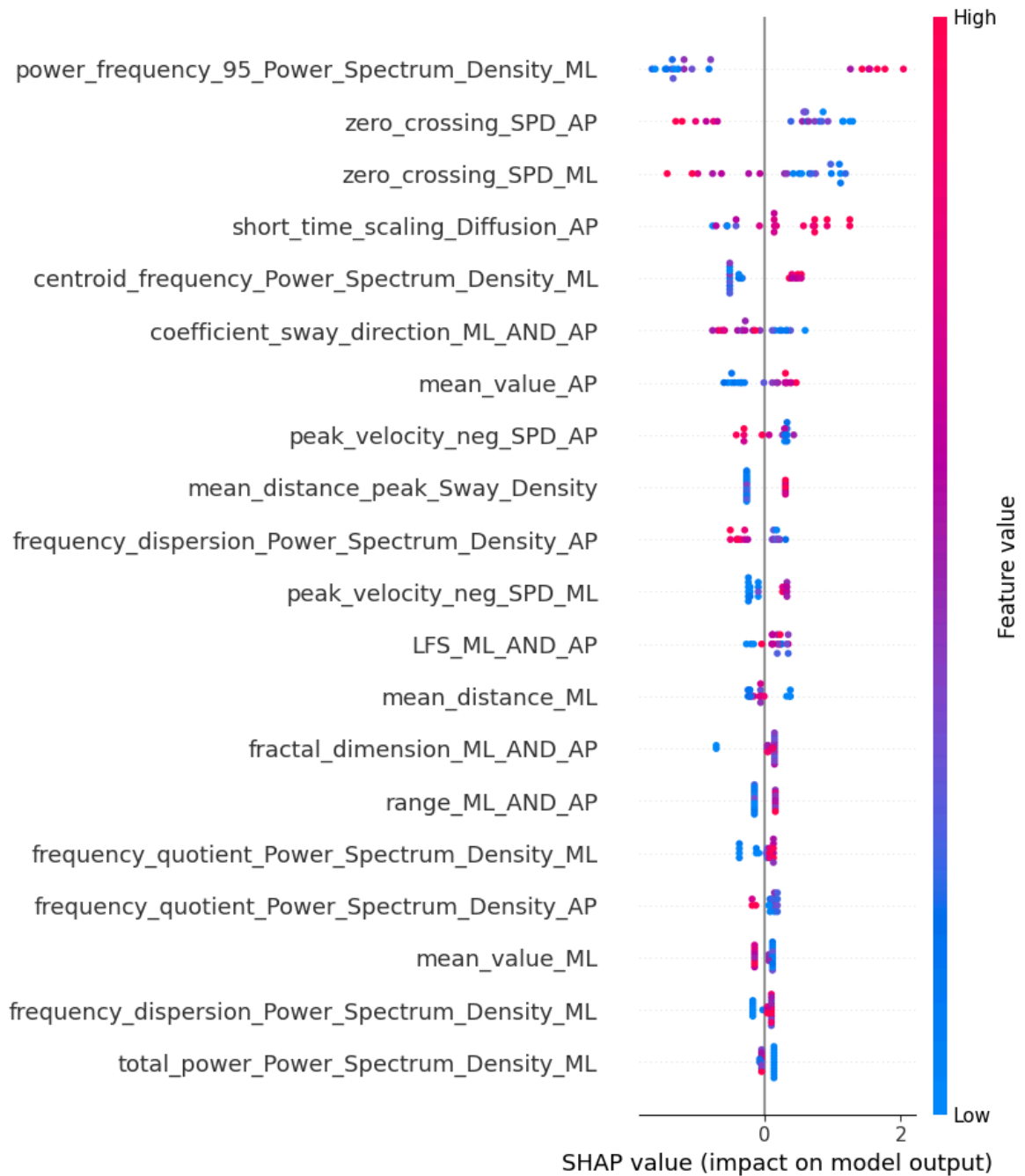
Accuracy: 0.7619

Precision: 0.7500

Recall: 0.8182

F1 Score: 0.7826





Results for split 4

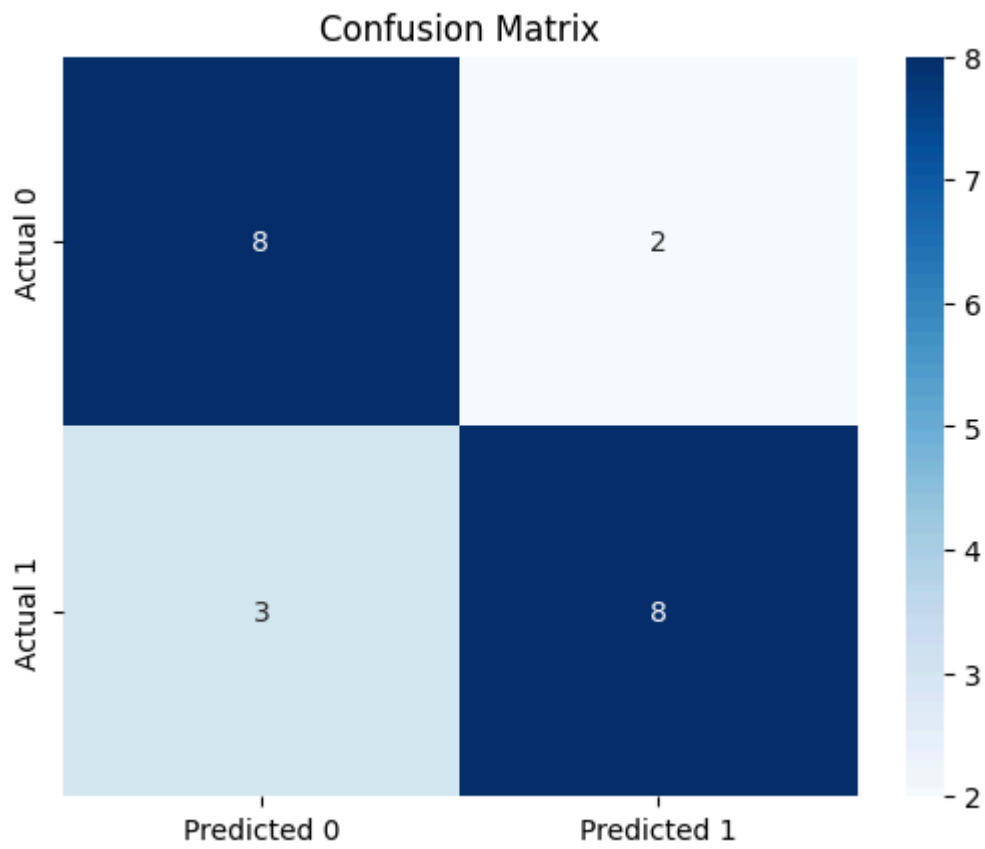
Accuracy: 0.7619

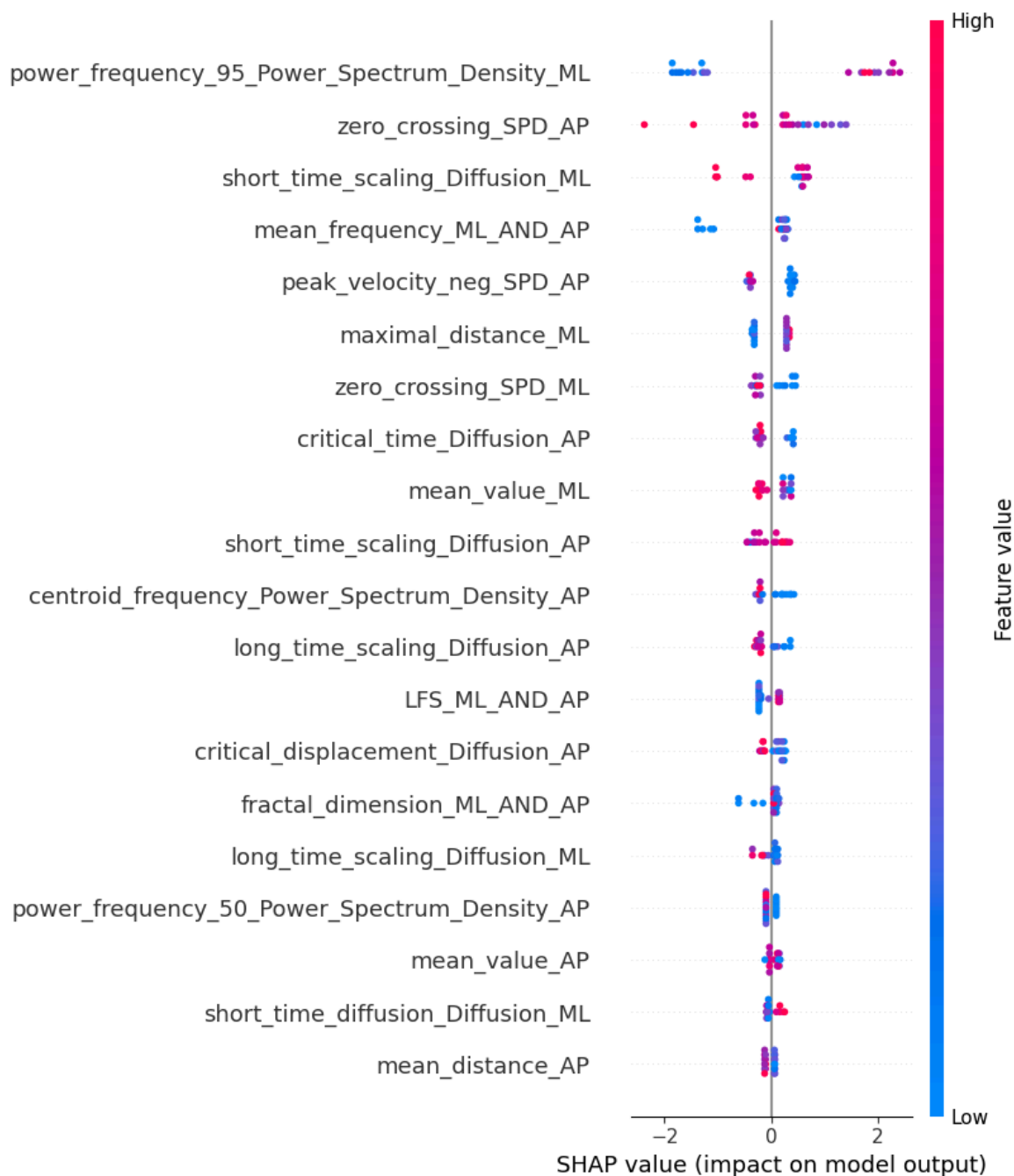
Precision: 0.8000

Recall: 0.7273

F1 Score: 0.7619







Results for split 5

Accuracy: 0.8571

Precision: 0.9000

Recall: 0.8182

F1 Score: 0.8571

