

1 is left unilateral

2 is right unilateral

3 is bilateral

```
In [ ]: import pandas as pd  
  
df = pd.read_csv('updated_complete_vol_visual.csv')
```

```

In [ ]: import pandas as pd
from sklearn.model_selection import train_test_split, cross_val_score, KFold
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, confusion_matrix
from sklearn.preprocessing import StandardScaler

PCA = df
PCA = PCA[PCA['PCA'] != 0]

X = pd.concat([PCA['PCAL_x'], PCA['PCAR_x']], axis=1)
y = PCA['PCA']

# Set up cross-validation:
kf = KFold(n_splits=5, shuffle=True, random_state=42)

model = LogisticRegression(multi_class='multinomial', solver='lbfgs', max_iter=1000)

# Perform cross-validation:
cv_scores = cross_val_score(model, X, y, cv=kf)

# Print cross-validated scores:
print("Cross-validated scores:", cv_scores)
print("Average cross-validation score:", cv_scores.mean())

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
model.fit(X_train, y_train)
y_pred = model.predict(X_test)

# Evaluate the model:
accuracy = accuracy_score(y_test, y_pred)
conf_matrix = confusion_matrix(y_test, y_pred)

print("Accuracy of the model on the test set:", accuracy)
print("")
print("Confusion Matrix:\n", conf_matrix)

# Create a DataFrame with the test set and predictions:
results_df = pd.DataFrame(X_test)
results_df['Actual'] = y_test
results_df['Predicted'] = y_pred

# Filter the DataFrame to only include misclassified instances:
misclassified = results_df[results_df['Actual'] != results_df['Predicted']]

print("")
print("Misclassified instances:")
print(misclassified)

```

Cross-validated scores: [0.92857143 0.92857143 0.92857143 0.88888889 1.
]

Average cross-validation score: 0.9349206349206348

Accuracy of the model on the test set: 0.9285714285714286

Confusion Matrix:

[[12 0 0]

[0 13 0]

[1 1 1]]

Misclassified instances:

	PCAL_x	PCAR_x	Actual	Predicted
410	975	10145	3.0	2.0
405	959	112	3.0	1.0

```

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Thalamus = df
Thalamus = Thalamus[Thalamus['PCA'] != 0]

X = pd.concat([Thalamus['ThalamusL_x'], Thalamus['ThalamusR_x']], axis=1)
y = Thalamus['Thalamus']

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```

Cross-validated scores: [0.82142857 0.75 0.82142857 0.92592593 0.92592593]

Average cross-validation score: 0.8489417989417989

Accuracy of the model on the test set: 0.8214285714285714

Confusion Matrix:

```
[[18  0  2]
```

```
[ 0  3  0]
```

```
[ 3  0  2]]
```

Misclassified instances:

	ThalamusL_x	ThalamusR_x	Actual	Predicted
406	0	161	2.0	0.0
87	0	473	2.0	0.0
411	0	343	2.0	0.0
24	0	543	0.0	2.0
414	0	876	0.0	2.0

```
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from sklearn.preprocessing import StandardScaler

cerebellum = df
cerebellum = cerebellum[cerebellum['PCA'] != 0]

X = pd.concat([cerebellum['cerebellumL_x'], cerebellum['cerebellumR_x']], axis=1)
y = cerebellum['cerebellum']

# Set up cross-validation:
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Cross-validated scores: [0.92857143 0.96428571 1. 0.85185185 0.8888 8889]

Average cross-validation score: 0.9267195767195766

Accuracy of the model on the test set: 0.9285714285714286

Confusion Matrix:

[[24 0 0]

[0 0 1]

[1 0 2]]

Misclassified instances:

	cerebellumL_x	cerebellumR_x	Actual	Predicted
406	31895	461	2.0	3.0
405	578	0	3.0	0.0