¥ Hide code

```
import numpy as np
import pandas as pd
from matplotlib import pyplot as plt
import tensorflow as tf
from sklearn.model_selection import train_test_split

import warnings
warnings.filterwarnings('ignore', category=Warning)
```

In [2]:

```
s00 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s00.csv',header=Non
e)
s01 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s01.csv',header=Non
e)
s02 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s02.csv',header=Non
e)
s03 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s03.csv',header=Non
e)
s04 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s04.csv',header=Non
e)
s05 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s05.csv',header=Non
e)
s06 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s06.csv',header=Non
e)
s07 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s07.csv',header=Non
e)
s08 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s08.csv', header=Non
e)
s09 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s09.csv',header=Non
e)
s10 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s10.csv',header=Non
e)
s11 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s11.csv',header=Non
e)
s12 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s12.csv',header=Non
e)
s13 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s13.csv',header=Non
e)
s14 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s14.csv',header=Non
e)
s15 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s15.csv',header=Non
e)
s16 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s16.csv',header=Non
e)
s17 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s17.csv',header=Non
e)
s18 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s18.csv',header=Non
e)
s19 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s19.csv',header=Non
e)
```

```
s20 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s20.csv',header=Non
e)
s21 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s21.csv',header=Non
e)
s22 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s22.csv',header=Non
e)
s23 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s23.csv',header=Non
e)
s24 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s24.csv',header=Non
e)
s25 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s25.csv',header=Non
e)
s26 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s26.csv',header=Non
e)
s27 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s27.csv',header=Non
e)
s28 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s28.csv', header=Non
e)
s29 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s29.csv',header=Non
e)
s30 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s30.csv', header=Non
e)
s31 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s31.csv',header=Non
e)
s32 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s32.csv',header=Non
e)
s33 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s33.csv',header=Non
e)
s34 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s34.csv',header=Non
e)
s35 = pd.read_csv('/kaggle/input/complete-eeg-dataset/s35.csv',header=Non
e)
```

```
In [3]: s00.shape
Out[3]: (31000, 19)
```

```
In [4]:
        s00=s00.transpose().to_numpy()
        s01=s01.transpose().to_numpy()
        s02=s02.transpose().to_numpy()
        s03=s03.transpose().to_numpy()
        s04=s04.transpose().to_numpy()
        s05=s05.transpose().to_numpy()
        s06=s06.transpose().to_numpy()
        s07=s07.transpose().to_numpy()
        s08=s08.transpose().to_numpy()
        s09=s09.transpose().to_numpy()
        s10=s10.transpose().to_numpy()
        s11=s11.transpose().to_numpy()
        s12=s12.transpose().to_numpy()
        s13=s13.transpose().to_numpy()
        s14=s14.transpose().to_numpy()
        s15=s15.transpose().to_numpy()
        s16=s16.transpose().to_numpy()
        s17=s17.transpose().to_numpy()
        s18=s18.transpose().to_numpy()
        s19=s19.transpose().to_numpy()
        s20=s20.transpose().to_numpy()
        s21=s21.transpose().to_numpy()
        s22=s22.transpose().to_numpy()
        s23=s23.transpose().to_numpy()
        s24=s24.transpose().to_numpy()
        s25=s25.transpose().to_numpy()
        s26=s26.transpose().to_numpy()
        s27=s27.transpose().to_numpy()
        s28=s28.transpose().to_numpy()
        s29=s29.transpose().to_numpy()
        s30=s30.transpose().to_numpy()
        s31=s31.transpose().to_numpy()
        s32=s32.transpose().to_numpy()
        s33=s33.transpose().to_numpy()
        s34=s34.transpose().to_numpy()
        s35=s35.transpose().to_numpy()
```

## **Original dataset with 36 samples**

```
In [5]:
          dataset= np.array([[s00],[s01],[s02],[s03],[s04],[s05],[s06],[s07],
                           [s08], [s09], [s10], [s11], [s12], [s13], [s14], [s15], [s16],
          [s17],[s18],[s19],
                           [s20],[s21],[s22],[s23],[s24],[s25],[s26],[s27],
                           [s28], [s29], [s30], [s31], [s32], [s33], [s34], [s35]])
  In [6]:
          dataset.shape
   Out[6]:
          (36, 1, 19, 31000)
Setting global random seed for model stability
  In [7]:
          seed = 42
          tf.random.set_seed(seed)
 In [10]:
          [0,1,1,1,1,1]
 In [12]:
          dataset = dataset.reshape(36, 1, 760, 775)
```

```
In [21]:
```

```
# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random_state=42)

# Logistic Regression
logistic_model = LogisticRegression()
logistic_model.fit(X_train, y_train)
logistic_preds = logistic_model.predict(X_test)
print("Logistic Regression Accuracy:", accuracy_score(y_test, logistic_preds))
print("Logistic Regression Classification Report:")
print(classification_report(y_test, logistic_preds))
```

Logistic Regression Accuracy: 0.375

Logistic Regression Classification Report:

support	f1-score	recall	precision	
2	0.29	0.50	0.20	0
6	0.44	0.33	0.67	1
_				
8	0.38			accuracy
8	0.37	0.42	0.43	macro avg
8	0.40	0.38	0.55	weighted avg

```
In [22]:
```

```
# Decision Tree
decision_tree_model = DecisionTreeClassifier(random_state=50)
decision_tree_model.fit(X_train, y_train)
decision_tree_preds = decision_tree_model.predict(X_test)
print("Decision Tree Accuracy:", accuracy_score(y_test, decision_tree_preds))
print("Decision Tree Classification Report:")
print(classification_report(y_test, decision_tree_preds))
```

Decision Tree Accuracy: 0.625

Decision Tree Classification Report:

support	f1-score	recall	precision	
2	0.00	0.00	0.00	0
6	0.77	0.83	0.71	1
8	0.62			accuracy
8	0.38	0.42	0.36	macro avg
8	0.58	0.62	0.54	weighted avg

```
In [20]:
```

```
# Gradient Boosting
gradient_boost_model = GradientBoostingClassifier(random_state=42)
gradient_boost_model.fit(X_train, y_train)
gradient_boost_preds = gradient_boost_model.predict(X_test)
print("Gradient Boosting Accuracy:", accuracy_score(y_test, gradient_boost_preds))
print("Gradient Boosting Classification Report:")
print(classification_report(y_test, gradient_boost_preds))
```

Gradient Boosting Accuracy: 0.5

Gradient Boosting Classification Report:

support	f1-score	recall	precision	
2	0.00	0.00	0.00	0
6	0.67	0.67	0.67	1
8	0.50			accuracy
8	0.33	0.33	0.33	macro avg
8	0.50	0.50	0.50	weighted avg

```
In [25]:
```

```
# Random Forest
random_forest_model = RandomForestClassifier(random_state=40)
random_forest_model.fit(X_train, y_train)
random_forest_preds = random_forest_model.predict(X_test)
print("Random Forest Accuracy:", accuracy_score(y_test, random_forest_preds))
print("Random Forest Classification Report:")
print(classification_report(y_test, random_forest_preds))
```

Random Forest Accuracy: 0.75

Random Forest Classification Report:

support	f1-score	recall	precision	
2	0.00	0.00	0.00	0
6	0.86	1.00	0.75	1
8	0.75			accuracy
8	0.43	0.50	0.38	macro avg
8	0.64	0.75	0.56	weighted avg

## In [27]:

```
pip install joblib
```

Requirement already satisfied: joblib in /opt/conda/lib/python3.7/site -packages (1.0.1)

WARNING: Running pip as the 'root' user can result in broken permissio ns and conflicting behaviour with the system package manager. It is re commended to use a virtual environment instead: https://pip.pypa.io/warnings/venv

Note: you may need to restart the kernel to use updated packages.

\_\_notebook\_\_

In [28]:

```
import time
from joblib import Parallel, delayed
# Function to train and evaluate a model
def train_and_evaluate(model, X_train, X_test, y_train, y_test):
    start_time = time.time()
   # Train the model
   model.fit(X_train, y_train)
   # Make predictions
    preds = model.predict(X_test)
   # Evaluate the model
    accuracy = accuracy_score(y_test, preds)
    report = classification_report(y_test, preds)
    elapsed_time = time.time() - start_time
    return accuracy, report, elapsed_time
# Serial Execution
print("Serial Execution:")
for model, model_name in zip([logistic_model, decision_tree_model, gradie
nt_boost_model, random_forest_model],
                             ["Logistic Regression", "Decision Tree", "Gr
adient Boosting", "Random Forest"]):
    acc, rep, elapsed_time = train_and_evaluate(model, X_train, X_test, y
_train, y_test)
    print(f"{model_name} - Accuracy: {acc:.4f}, Elapsed Time: {elapsed_ti
me:.2f} seconds")
    print(f"Classification Report:\n{rep}\n")
# Parallel Execution
print("Parallel Execution:")
models = [logistic_model, decision_tree_model, gradient_boost_model, rand
om_forest_model]
names = ["Logistic Regression", "Decision Tree", "Gradient Boosting", "Ra
ndom Forest"l
results_parallel = Parallel(n_jobs=-1)(
```

```
delayed(train_and_evaluate)(model, X_train, X_test, y_train, y_test)
for model in models
)

for (acc, rep, elapsed_time), model_name in zip(results_parallel, names):
    print(f"{model_name} - Accuracy: {acc:.4f}, Elapsed Time: {elapsed_time:.2f} seconds")
    print(f"Classification Report:\n{rep}\n")
```

## Serial Execution:

Logistic Regression - Accuracy: 0.3750, Elapsed Time: 3.93 seconds Classification Report:

	precision	recall	f1-score	support
0	0.20	0.50	0.29	2
1	0.67	0.33	0.44	6
accuracy			0.38	8
macro avg	0.43	0.42	0.37	8
weighted avg	0.55	0.38	0.40	8

Decision Tree - Accuracy: 0.6250, Elapsed Time: 1.66 seconds Classification Report:

	precision	recall	f1-score	support
0	0.00	0.00	0.00	2
1	0.71	0.83	0.77	6
accuracy			0.62	8
macro avg	0.36	0.42	0.38	8
weighted avg	0.54	0.62	0.58	8

Gradient Boosting - Accuracy: 0.5000, Elapsed Time: 197.91 seconds Classification Report:

		precision	recall	f1-score	support
	0	0.00	0.00	0.00	2
	1	0.67	0.67	0.67	6
accur	асу			0.50	8
macro	avg	0.33	0.33	0.33	8
weighted	avg	0.50	0.50	0.50	8

Random Forest - Accuracy: 0.7500, Elapsed Time: 0.41 seconds Classification Report:

precision recall f1-score support

	0	0.00	0.00	0.00	2
	1	0.75	1.00	0.86	6
accura	су			0.75	8
macro a	ıvg	0.38	0.50	0.43	8
weighted a	ıvg	0.56	0.75	0.64	8

## Parallel Execution:

Logistic Regression - Accuracy: 0.3750, Elapsed Time: 8.15 seconds Classification Report:

	precision	recall	f1-score	support
0	0.20	0.50	0.29	2
1	0.67	0.33	0.44	6
accuracy			0.38	8
macro avg	0.43	0.42	0.37	8
weighted avg	0.55	0.38	0.40	8

Decision Tree - Accuracy: 0.6250, Elapsed Time: 2.63 seconds Classification Report:

	precision	recall	f1-score	support
0	0.00	0.00	0.00	2
1	0.71	0.83	0.77	6
accuracy			0.62	8
macro avg	0.36	0.42	0.38	8
weighted avg	0.54	0.62	0.58	8

Gradient Boosting - Accuracy: 0.5000, Elapsed Time: 204.35 seconds Classification Report:

suppo	f1-score	recall	precision	p
	0.00	0.00	0.00	0
	0.67	0.67	0.67	1

accur	acy			0.50	Ö
macro	avg	0.33	0.33	0.33	8
weighted	avg	0.50	0.50	0.50	8

Random Forest - Accuracy: 0.7500, Elapsed Time: 0.90 seconds Classification Report:

	precision	recall	f1-score	support
0	0.00	0.00	0.00	2
1	0.75	1.00	0.86	6
accuracy			0.75	8
macro avg	0.38	0.50	0.43	8
weighted avg	0.56	0.75	0.64	8

In [29]:

```
import time
from joblib import Parallel, delayed
# Function to train and evaluate a model
def train_and_evaluate(model, X_train, X_test, y_train, y_test):
    start_time = time.time()
   # Train the model
    model.fit(X_train, y_train)
    # Make predictions
    preds = model.predict(X_test)
    # Evaluate the model
    accuracy = accuracy_score(y_test, preds)
    report = classification_report(y_test, preds)
    elapsed_time = time.time() - start_time
    print(f"{model.__class__.__name__}} - Accuracy: {accuracy:.4f}, Elapse
d Time: {elapsed_time:.2f} seconds")
    print(f"Classification Report:\n{report}\n")
    return accuracy, elapsed_time
# Serial Execution
serial_start_time = time.time()
print("Serial Execution:")
serial_elapsed_times = []
for model in [logistic_model, decision_tree_model, gradient_boost_model,
random_forest_model]:
    acc, elapsed_time = train_and_evaluate(model, X_train, X_test, y_trai
n, y_test)
    serial_elapsed_times.append(elapsed_time)
serial_total_time = time.time() - serial_start_time
print(f"Total Time (Serial): {serial_total_time:.2f} seconds\n")
# Parallel Execution
parallel_start_time = time.time()
print("Parallel Execution:")
```

```
models = [logistic_model, decision_tree_model, gradient_boost_model, rand
om_forest_model]
names = ["Logistic Regression", "Decision Tree", "Gradient Boosting", "Ra
ndom Forest"]

results_parallel = Parallel(n_jobs=-1)(
    delayed(train_and_evaluate)(model, X_train, X_test, y_train, y_test)
for model in models
)

parallel_total_time = time.time() - parallel_start_time
print(f"Total Time (Parallel): {parallel_total_time:.2f} seconds\n")

# Print elapsed times for each model in parallel
for (acc, elapsed_time), model_name in zip(results_parallel, names):
    print(f"{model_name} - Accuracy: {acc:.4f}, Elapsed Time: {elapsed_ti
me:.2f} seconds\n")
```

Serial Execution:

LogisticRegression - Accuracy: 0.3750, Elapsed Time: 4.11 seconds Classification Report:

	precision	recall	f1-score	support
0	0.20	0.50	0.29	2
1	0.67	0.33	0.44	6
accuracy			0.38	8
macro avg	0.43	0.42	0.37	8
weighted avg	0.55	0.38	0.40	8

DecisionTreeClassifier - Accuracy: 0.6250, Elapsed Time: 1.73 seconds Classification Report:

	precision	recall	f1-score	support
0	0.00	0.00	0.00	2
1	0.71	0.83	0.77	6
accuracy			0.62	8
macro avg	0.36	0.42	0.38	8
weighted avg	0.54	0.62	0.58	8

/opt/conda/lib/python3.7/site-packages/sklearn/metrics/\_classificatio n.py:1221: UndefinedMetricWarning: Precision and F-score are ill-defin ed and being set to 0.0 in labels with no predicted samples. Use `zero \_division` parameter to control this behavior.

\_warn\_prf(average, modifier, msg\_start, len(result))

GradientBoostingClassifier - Accuracy: 0.5000, Elapsed Time: 198.54 se conds

Classification Report:

	precision	recall	f1-score	support
0	0.00	0.00	0.00	2
1	0.67	0.67	0.67	6
accuracy			0.50	8
macro avg	0.33	0.33	0.33	8
weighted avg	0.50	0.50	0.50	8

RandomForestClassifier - Accuracy: 0.7500, Elapsed Time: 0.40 seconds Classification Report:

	precision	recall	f1-score	support
0	0.00	0.00	0.00	2
1	0.75	1.00	0.86	6
accuracy			0.75	8
macro avg	0.38	0.50	0.43	8
weighted avg	0.56	0.75	0.64	8

Total Time (Serial): 204.79 seconds

Parallel Execution:

Total Time (Parallel): 201.39 seconds

Logistic Regression - Accuracy: 0.3750, Elapsed Time: 8.86 seconds

Decision Tree - Accuracy: 0.6250, Elapsed Time: 2.73 seconds

Gradient Boosting - Accuracy: 0.5000, Elapsed Time: 198.72 seconds

Random Forest - Accuracy: 0.7500, Elapsed Time: 0.91 seconds

RandomForestClassifier - Accuracy: 0.7500, Elapsed Time: 0.91 seconds Classification Report:

support	f1-score	recall	precision	
2	0.00	0.00	0.00	0
6	0.86	1.00	0.75	1
8	0.75			accuracy
8	0.43	0.50	0.38	macro avg
8	0.64	0.75	0.56	weighted avg

/opt/conda/lib/python3.7/site-packages/sklearn/metrics/\_classificatio n.py:1221: UndefinedMetricWarning: Precision and F-score are ill-defin ed and being set to 0.0 in labels with no predicted samples. Use `zero \_division` parameter to control this behavior.

\_warn\_prf(average, modifier, msg\_start, len(result))

DecisionTreeClassifier - Accuracy: 0.6250, Elapsed Time: 2.73 seconds Classification Report:

support	f1-score	recall	precision	
2	0.00	0.00	0.00	0
6	0.77	0.83	0.71	1
8	0.62			accuracy
8	0.38	0.42	0.36	macro avg
8	0.58	0.62	0.54	weighted avg

LogisticRegression - Accuracy: 0.3750, Elapsed Time: 8.86 seconds Classification Report:

	precision	recall	f1-score	support
0	0.00	0 50	0.20	2
0	0.20	0.50	0.29	2
1	0.67	0.33	0.44	6
accuracy			0.38	8
macro avg	0.43	0.42	0.37	8
weighted avg	0.55	0.38	0.40	8

GradientBoostingClassifier - Accuracy: 0.5000, Elapsed Time: 198.72 se conds

Classification Report:

support	f1-score	recall	precision	
2	0.00	0.00	0.00	0
6	0.67	0.67	0.67	1
8	0.50			accuracy
8	0.33	0.33	0.33	macro avg
8	0.50	0.50	0.50	weighted avg

/22/24, 1:15 PM	notebook
In [ ]:	