Final year Project

Accuracy of models:

Code:

```
import pandas as pd
import numpy as np
import lightgbm as lgb
from sklearn.metrics import accuracy_score
elapsed_times = []
train_data_path = 'Trainset.csv'
train_df = pd.read_csv(train_data_path)
test data path = 'Testset1.csv'
test_df = pd.read_csv(test_data_path)
start_time = time.time()
X_train = train_df.drop('label', axis=1)
y_train = train_df['label']
X_test = test_df.drop('label', axis=1)
y_test = test_df['label']
clf = lgb.LGBMClassifier()
clf.fit(X_train, y_train)
predictions = clf.predict(X_test)
print("Accuracy: " + str(accuracy_score(y_test, predictions)))
```

Output --- Testset1: Accuracy: 0.999675

Output --- Testset2: Accuracy: 0.937175

1. KS test

code:

```
import pandas as pd
from scipy import stats
train_data = pd.read_csv('Trainset.csv')
test_data = pd.read_csv('Testset2.csv')
columns_to_test = train_data.columns
for column in columns_to_test:
   train_values = train_data[column]
    test_values = test_data[column]
    ks_stat, p_value = stats.ks_2samp(train_values, test_values)
    print(f"Column: {column}")
    print(f"KS Statistic: {ks_stat}")
    print(f"P-Value: {p_value}")
    if p_value < 0.05:
       print(f"Drift detected in column {column}!")
        print(f"No drift detected in column {column}.")
    print()
```

Output ---Testset1:

Column: label KS Statistic: 0.0 P-Value: 1.0

No drift detected in column label.

Column: orig_ip_bytes

KS Statistic: 0.0005124999999999991

P-Value: 1.0

No drift detected in column orig ip bytes.

Column: orig_pkts

KS Statistic: 0.000500000000000000000

P-Value: 1.0

No drift detected in column orig_pkts.

Output ---Testset2:

Column: label KS Statistic: 0.0 P-Value: 1.0

No drift detected in column label.

Column: orig ip bytes

Drift detected in column orig_ip_bytes!

Column: orig_pkts

KS Statistic: 0.000500000000000000000

P-Value: 1.0

No drift detected in column orig_pkts.

2.CUSUM test

```
import pandas as pd
import numpy as np

train_df = pd.read_csv('Trainset.csv')
test_df = pd.read_csv('Testset2.csv')
```

```
column_to_monitor = 'orig_ip_bytes'

train_mean = train_df[column_to_monitor].mean()
train_std = train_df[column_to_monitor].std()

cusum = np.cumsum((test_df[column_to_monitor] - train_mean) / train_std)

drift_index = np.argmax(cusum)

drift_indices = np.where(cusum > 5)[0]
print(f"Drifts detected at indices {drift_indices}!")

if np.max(cusum) > 5:
    print("Drift detected!")
else:
    print("No drift detected!")
```

```
Drifts detected at indices [50019 50020 50021 ... 79997 79998 79999]!
```

Output ---Testset2:

```
Drifts detected at indices [50019 50020 50021 ... 79997 79998 79999]!
label 3
orig_ip_bytes 139377944
orig_pkts 4806136
Name: 74748, dtype: int64
Drift detected!
PS C:\Users\HP\Desktop\final>
```

3. Logistic Regression

```
import pandas as pd
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score

train_df = pd.read_csv('Trainset.csv')
test_df = pd.read_csv('Testset2.csv')
```

```
feature_cols = train_df.columns
target_col = 'orig_ip_bytes'
clf = LogisticRegression()
clf.fit(train_df[feature_cols], train_df[target_col])
y_pred_train = clf.predict(train_df[feature_cols])
train_accuracy = accuracy_score(train_df[target_col], y_pred_train)
print(f"Training accuracy: {train_accuracy:.3f}")
y_pred_test = clf.predict(test_df[feature_cols])
test_accuracy = accuracy_score(test_df[target_col], y_pred_test)
print(f"Test accuracy: {test_accuracy:.3f}")
if test_accuracy < train_accuracy * 0.8:</pre>
   print("Drift detected!")
    drift_indices = []
    for i, (y_pred, y_true) in enumerate(zip(y_pred_test, test_df[target_col])):
        if y pred != y true:
            drift_indices.append(i)
    print(f"Drift indices: {drift_indices}")
else:
    print("No drift detected.")
```

```
Training accuracy: 0.970
Test accuracy: 0.969
```

Output ---Testset2:

```
Training accuracy: 0.970
Test accuracy: 0.913
```

4.AdWin

```
import pandas as pd
from river import drift
train df = pd.read csv('Trainset.csv')
test_df = pd.read_csv('Testset1.csv')
columns_to_monitor = train_df.columns
adwin_objects = []
for column in columns_to_monitor:
    adwin_objects.append(drift.ADWIN(delta=0.002))
drifts = []
for i, row in test_df.iterrows():
    drift_detected = False
    for j, column in enumerate(columns_to_monitor):
        adwin_objects[j].update(row[column])
        if adwin_objects[j].drift_detected:
            drift_detected = True
            break
    if drift_detected:
        drifts.append((i, row[columns_to_monitor]))
        for adwin in adwin_objects:
            adwin_objects = [drift.ADWIN(delta=0.002) for _ in
columns_to_monitor]
for drift in drifts:
    print('Drift detected at index {}: \n{}'.format(drift[0], drift[1]))
```

```
Drift detected at index 40031:
label 3.0
orig_ip_bytes 80.0
orig_pkts 2.0
Name: 40031, dtype: float64
```

Output ---Testset2:

```
Drift detected at index 40031:
label 3
orig_ip_bytes 80
orig_pkts 2
Name: 40031, dtype: int64
```

5.KSWin

```
import pandas as pd
from river.drift import KSWIN
train_data = pd.read_csv('Trainset.csv')
test data = pd.read csv('Testset2.csv')
def detect_drift(train_data, test_data, feature_column):
    kswin = KSWIN()
    drift_indices = []
    for value in train_data[feature_column]:
        kswin.update(value)
        for index, value in enumerate(test_data[feature_column]):
        kswin.update(value)
        if kswin.drift_detected:
            drift_indices.append(index)
            print(f"Drift detected at index {index} in feature {feature_column}")
    if not drift_indices:
        print(f"No drift detected in feature {feature_column}")
```

```
else:
    print(f"Drift detected at indices {drift_indices} in feature
{feature_column}")

for feature in train_data.columns:
    if feature != 'orig_ip_bytes':
        print(f"Checking drift in feature: {feature}")
        detect_drift(train_data, test_data, feature)
```

```
Checking drift in feature: label

Drift detected at index 13 in feature label

Drift detected at index 40013 in feature label

Drift detected at indices [13, 40013] in feature label

Checking drift in feature: orig_pkts

Drift detected at index 15 in feature orig_pkts

Drift detected at index 40012 in feature orig_pkts

Drift detected at indices [15, 40012] in feature orig_pkts
```

Output ---Testset2:

```
Checking drift in feature: label
Drift detected at index 13 in feature label
Drift detected at index 40013 in feature label
Drift detected at indices [13, 40013] in feature label
Checking drift in feature: orig_pkts
Drift detected at index 15 in feature orig_pkts
Drift detected at index 40012 in feature orig_pkts
Drift detected at indices [15, 40012] in feature orig_pkts
```

5. Page Hinkley

```
import pandas as pd
import numpy as np
def page_hinkley_test(data, threshold=50, lambda_=50, alpha=1 - 1e-4):
    mean = 0
    sum values = 0
    m_max = -np.inf
    change point = None
    for i, x in enumerate(data):
        sum values += x - mean - lambda_
        mean = (1 / (i + 1)) * ((i * mean) + x)
        m_t = sum_values / (i + 1)
        if m_t > m_max:
            m_max = m_t
        elif m_max - m_t > threshold:
            change_point = i
            break
    return change_point
train_data = pd.read_csv("Trainset.csv")
test_data = pd.read_csv("Testset2.csv")
```

```
columns_to_check = ['orig_ip_bytes', 'orig_pkts']

drift_results = {}
for col in columns_to_check:
    if test_data[col].dtype in (np.float64, np.int64):
        drift_index = page_hinkley_test(test_data[col].values)
        if drift_index is not None:
            drift_results[col] = f"Drift detected at index {drift_index}"
        else:
            drift_results[col] = "No drift detected"

print(drift_results)
```

```
{'orig_ip_bytes': 'Drift detected at index 1', 'orig_pkts': 'Drift detected at index 52199'}
```

Output ---Testset2:

{'orig_ip_bytes': 'Drift detected at index 1', 'orig_pkts': 'Drift detected at index 52199'}

| S.no | Method | Testset1 | Testset2 | Conclusion |
|------|-----------------|----------------|----------------|-------------------------|
| 1. | Accuracy | 0.999675 | 0.937175 | Since there is drift |
| | (Decision tree) | No drift | Drift | testset2, it has less |
| | | | | accuracy |
| 2. | KS test | No drift | Drift | Based on p-value, it is |
| | | | | verified |
| 3. | Cusum | No drift | Drift | Based on threshold |
| | | | | value, it is verified |
| 4. | Logistic | Test Acc:0.969 | Test Acc:0.913 | Since there is drift |
| | Regression | No drift | Drift | testset2, it has less |
| | | | | test accuracy |
| 5. | Adwin | drift | drift | Windowing method not |
| | | | | efficient for static |
| | | | | dataset |
| 6. | KsWin | drift | drift | Windowing method not |
| | | | | efficient for static |
| | | | | dataset. |
| 7. | Page Hinkley | drift | Drift | Not efficent |
| | | | | |