In [1]: #NIKHIL KUMAR NYALAM-11636071

```
In [2]: import pandas as pd
        import numpy as np
        import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        from sklearn.cluster import KMeans
        from sklearn.ensemble import IsolationForest
        from sklearn.preprocessing import StandardScaler
        from sklearn import preprocessing
        from sklearn.preprocessing import MinMaxScaler
        import numpy as np
        from sklearn.ensemble import IsolationForest
        import pandas as pd
        import matplotlib.pyplot as plt
        from sklearn.metrics import accuracy_score
        #1(ONE)
        def getData(filename):
            canData = [] # CANDATA LIST
            f = open(filename)
            read_file = f.readlines()
            speed = []
            rpm = []
            i=0
            for row in read_file:
              #POSITIONS ARE VERIFIED
                record = {
                     'stamp': row[1:18],
                     'PID': row[25:28],
                     'const1': row[29:33],
                     'change': row[33:41],
                     'value': int(row[41:45], 16),
                     'value2': 0,
                     'attack': 0
                }
                if record["PID"] == '254': # CAN ID SHOULD BE 254 FOR SPEED
                    if record["value"] >= 4095:
                         record["attack"] = 1
                    record['value'] = (record['value'] * 0.62137119) / 100
                    speed.append(record['value'])
                if record["PID"] == '115': # CANID SHOULD BE 115
                    if record["value"] >= 65535:
                         record["attack"] = 1
                    record['value'] = (record['value'] * 2)
                    rpm.append(record['value'])
                canData.append(record)
            f.close()
            return canData
        def dict to df(data dict):
            # DICTIONART TO DATAFRAME
            df = pd.DataFrame.from_dict(data_dict)
            #df = df.drop(columns=['stamp', 'const1', 'change', 'value2'])
```

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df = df.loc[(df['PID'] == '115') | (df['PID'] == '254')]
    df = df.reset index(drop=True)
    one hot = pd.get dummies(df['PID'])
    df = df.drop('PID', axis=1)
    df = df.join(one hot)
    df = df[['115', '254', 'value', 'attack']]
    df.rename(columns={'115': 'RPM', '254': 'Speed'}, inplace=True) #115 RO RP
    #df.loc[df['RPM'] == 1, 'RPM'] = df['value']
    #df.loc[df['Speed'] == 1, 'Speed'] = df['value']
    #df = df.drop(columns=['value'])
    return df
#normalisation of the column called value to bring in range 0,1
def normalize(df):
    values = df['value'].values
    values = values.reshape(-1, 1)
    min_max_scaler = MinMaxScaler()
    scaled values = min max scaler.fit transform(values)
    df['value'] = scaled_values
    return df
fff_injection_df = dict_to_df(getData("C:\\Users\\nyala\\Downloads\\CAN Bus lo
rpm_injection_df = dict_to_df(getData("C:\\Users\\nyala\\Downloads\\CAN Bus lo
no_injection_df = dict_to_df(getData("C:\\Users\\nyala\\Downloads\\CAN Bus log
fff injection df=normalize(fff injection df)
rpm_injection_df=normalize(rpm_injection_df)
no_injection_df=normalize(no_injection_df)
original_fff=fff_injection_df
original rpm=rpm injection df
original_no=no_injection_df
print(original_fff)
#dropping of column attack
column to drop='attack'
fff injection df = fff injection df.drop(column to drop, axis=1)
rpm injection df = rpm injection df.drop(column to drop, axis=1)
no injection df= no injection df.drop(column to drop, axis=1)
print("fff injection df :")
print(fff_injection_df)
print("rpm_injection_df :")
print(rpm injection df)
print("no injection df :")
print(no_injection_df)
```

			ı

	RPM	Speed	value	attack
0	1	•	0.353273	0
1	0		0.000000	0
2	1		0.352144	0
3	0		0.000000	0
4	1		0.352144	0
	• • •	• • • •		• • •
14430	0		0.003096	0
14431	0		0.014360	1
14432	0		0.014360	1
14433	1		0.511287	0
14434	0	1	0.014360	1
[14435	rows	x 4 co	olumns]	
fff_in	jecti	on_df	:	
	RPM	Speed	value	
0	1	0	0.353273	
1	0	1	0.000000	
2	1	0	0.352144	
3	0	1	0.000000	
4	1	0	0.352144	
 14430			0.003096	
14431	0		0.014360	
14431	0		0.014360	
14433	1		0.511287	
14434	0	1	0.014360	
[14435	rows	х 3 со	olumns]	
rpm_in	jecti	on_df	:	
	RPM	Speed	value	
0	1	0	0.005607	
1	0	1	0.000001	
2	1	0	0.005622	
3	0	1	0.000001	
4	1	0	0.005637	
4530	• • •	• • •		
4538	0	1	0.000021	
4539	1	0	1.000000	
4540	1	0	0.006690	
4541	1	0	1.000000	
4542	1	0	1.000000	
[4543]	rows	x 3 co	lumns]	

[4543 rows x 3 columns] no injection df :

	,		•
	RPM	Speed	value
0	1	0	0.247471
1	0	1	0.000000
2	1	0	0.249027
3	0	1	0.000000
4	1	0	0.249027
			• • •
1644	1	0	0.273152
1645	1	0	0.273152
1646	0	1	0.001182
1647	1	0	0.273152
1648	0	1	0.001141

```
[1649 rows x 3 columns]
```

```
In [3]: #2(TWO)
    def divide(df):
        df_speed=df[['Speed','value']]
        df_rpm=df[['RPM','value']]
        return df_speed,df_rpm
    #making 6 data frames
    ff_speed,ff_rpm=divide(fff_injection_df)
    rpm_speed,rpm_rpm=divide(rpm_injection_df)
    no_speed,no_rpm=divide(no_injection_df)
    #print(ff_speed)
    #print(ff_rpm)
```

```
In [4]: #2(Two)
        import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        from sklearn.cluster import KMeans
        import warnings
        warnings.filterwarnings('ignore')
        def perform kmeans and plot(df speed, df rpm, scenario name):
        #building the model k means clustering
            kmeans_speed = KMeans(n_clusters=2, random_state=0)
            df_speed['Speed_Cluster'] = kmeans_speed.fit_predict(df_speed[['Speed', 'v
            speed_centroids = kmeans_speed.cluster_centers_
        #finding out the centroid by the function cluster centers
            kmeans_rpm = KMeans(n_clusters=2, random_state=0)
            df_rpm['RPM_Cluster'] = kmeans_rpm.fit_predict(df_rpm[['RPM', 'value']])
            rpm centroids = kmeans rpm.cluster centers
            plt.figure(figsize=(12, 5))
            plt.subplot(1, 2, 1)
            plt.scatter(df_speed['Speed'], df_speed['value'], c=df_speed['Speed_Cluste
            plt.scatter(speed_centroids[:, 0], speed_centroids[:, 1], marker='X', s=20
            plt.title(f'{scenario name} - Speed Clusters')
            plt.xlabel('Speed')
            plt.ylabel('value')
            plt.legend()
            plt.subplot(1, 2, 2)
            plt.scatter(df_rpm['RPM'], df_rpm['value'], c=df_rpm['RPM_Cluster'], cmap=
            plt.scatter(rpm centroids[:, 0], rpm centroids[:, 1], marker='X', s=200, c
            plt.title(f'{scenario_name} - RPM Clusters')
            plt.xlabel('RPM')
            plt.ylabel('value')
            plt.legend()
            plt.tight_layout()
            plt.show()
            print('scenario name Speed Cluster Assignments:')
            print(df_speed[['Speed', 'value', 'Speed_Cluster']])
            print('Speed Centroids:')
            print(speed centroids)
            print('RPM Cluster Assignments:')
            print(df_rpm[['RPM', 'value', 'RPM_Cluster']])
            print('RPM Centroids:')
            print(rpm_centroids)
        perform_kmeans_and_plot(ff_speed, ff_rpm, 'FFF Injection')
        perform_kmeans_and_plot(rpm_speed, rpm_rpm, 'RPM Injection')
        perform kmeans and plot(no speed, no rpm, 'No Injection')
```

[[1.00000000000000 0.>>0010000 0.>] [-6.10622664e-16 6.41226135e-01]] RPM Cluster Assignments: RPMvalue RPM_Cluster 0 1 0.247471 1 1 0 0.000000 2 1 0.249027 1 3 0.000000 0 4 1 0.249027 1 1644 1 0.273152 1 1645 1 0.273152 1 0 1646 0 0.001182 1647 1 0.273152 1 1648 0 0.001141 [1649 rows x 3 columns] RPM Centroids: [[5.55111512e-16 6.55061868e-03] [1.00000000e+00 6.41226135e-01]]

```
In [5]: |#3(Three)
        from sklearn.ensemble import IsolationForest
        import pandas as pd
        import matplotlib.pyplot as plt
        #Isolation forest and scateer plot
        def fit isolation forest and plot(df speed, df rpm, scenario name):
            # Speed
            iso forest speed = IsolationForest(contamination=0.01, random state=0)
            df_speed['Speed_Outlier'] = iso_forest_speed.fit_predict(df_speed[['Speed'
            # RPM
            iso_forest_rpm = IsolationForest(contamination=0.01, random state=0)
            df_rpm['RPM_Outlier'] = iso_forest_rpm.fit_predict(df_rpm[['RPM', 'value']
            df_speed.to_csv('anomalies_speed1.csv', index=False)
            # Save anomalies in RPM dataset to a CSV file
            df_rpm.to_csv('anomalies_rpm1.csv', index=False)
            # plots
            plt.figure(figsize=(12, 5))
            # Scatter plot for Speed
            plt.subplot(1, 2, 1)
            plt.scatter(df_speed['Speed'], df_speed['value'], c=df_speed['Speed_Outlie']
            plt.title(f'{scenario name} - Speed Outliers')
            plt.xlabel('Speed')
            plt.ylabel('value')
            # Scatter plot for RPM
            plt.subplot(1, 2, 2)
            plt.scatter(df_rpm['RPM'], df_rpm['value'], c=df_rpm['RPM_Outlier'], cmap=
            plt.title(f'{scenario_name} - RPM Outliers')
            plt.xlabel('RPM')
            plt.ylabel('value')
            plt.tight_layout()
            plt.show()
            return df_speed, df_rpm
        def get anomalies(df speed, df rpm, scenario name):
            speed_anomalies = df_speed[df_speed['Speed_Outlier'] == -1]
            rpm_anomalies = df_rpm[df_rpm['RPM_Outlier'] == -1]
            print(f'Anomalies in {scenario_name} - Speed:')
            print(speed anomalies)
            print(f'Anomalies in {scenario_name} - RPM:')
            print(rpm anomalies)
```

```
ff_speed, ff_rpm = fit_isolation_forest_and_plot(ff_speed, ff_rpm, 'FFF Inject')
        rpm_speed, rpm_rpm = fit_isolation_forest_and_plot(rpm_speed, rpm_rpm, 'RPM In
        no_speed, no_rpm = fit_isolation_forest_and_plot(no_speed, no_rpm, 'No Injecti
        # Get anomalies for each scenario
        get_anomalies(ff_speed, ff_rpm, 'FFF Injection')
        get_anomalies(rpm_speed, rpm_rpm, 'RPM Injection')
        get_anomalies(no_speed, no_rpm, 'No Injection')
        681
                 0 0.996887
                                          1
                                                        -1
        Anomalies in No Injection - RPM:
                                         RPM_Outlier
             RPM
                     value RPM_Cluster
        1
               0.000000
                                      0
                                                  -1
        3
               0.000000
                                                  -1
                                      0
        5
               0.000000
                                      0
                                                  -1
        8
               0.000000
                                      0
                                                  -1
        10
               0.000000
                                      0
                                                  -1
        12
               0.000000
                                      0
                                                  -1
        658
               1 0.995331
                                      1
                                                  -1
                                      1
        662
                                                  -1
               1 0.996887
                                      1
                                                  -1
        667
               1 0.996887
        671
               1 0.996109
                                      1
                                                  -1
        672
               1 0.997665
                                      1
                                                  -1
                                      1
        676
               1 0.996109
                                                  -1
        678
               1 1.000000
                                      1
                                                  -1
        679
                                      1
                                                  -1
               1 0.996109
                                      1
                                                  -1
        681
               1 0.996887
        686
               1 0.995331
                                      1
                                                  -1
In [6]: #!pip install hmmlearn
In [ ]:
In [ ]:
```

```
In [7]: #4(fourjkmnm)
        import numpy as np
        from hmmlearn import hmm
        from sklearn.metrics import silhouette score
        import matplotlib.pyplot as plt
        import pandas as pd
        def hmm1(df,s):
                sequences = df[['Speed']].values
                #HMM model
                n components = 2 # Number of hidden states
                model = hmm.GaussianHMM(n_components=n_components, covariance_type="di
                # Train the HMM
                model.fit(sequences)
                # trained HMM is used to decode the most likely sequence of hidden sta
                decoded_states = model.predict(sequences)
                # Hidden States are visualized
                plt.figure(figsize=(12, 6))
                colors = ['blue' if state == 0 else 'red' for state in decoded_states]
                plt.scatter(sequences, df['value'], c=colors, s=50)
                plt.xlabel("Speed")
                plt.ylabel("Data Values")
                plt.title("Visualization of (0: Non-Attack, 1: Attack) "+s)
                plt.show()
                return 0
        def hmm2(df,s):
                sequences = df[['RPM']].values
                # HMM model
                n components = 2 # Number of hidden states
                model = hmm.GaussianHMM(n_components=n_components, covariance_type="di
                # Train the HMM
                model.fit(sequences)
                decoded_states = model.predict(sequences)
                plt.figure(figsize=(12, 6))
                colors = ['blue' if state == 0 else 'red' for state in decoded_states]
```

```
plt.scatter(sequences, df['value'], c=colors, s=50)
                 plt.xlabel("RPM")
                 plt.ylabel("Data Values")
                 plt.title("Visualization of Hidden States (0: Non-Attack, 1: Attack)"+
                 plt.show()
                  return 0
         hmm1(original_fff,'fff_injection-speed')
         hmm1(original_rpm,'rpm_injection-speed')
         hmm1(original_no, 'no_injection-speed')
         hmm2(original_fff,'fff_injection-rpm')
         hmm2(original_rpm,'rpm_injection-rpm')
         hmm2(original_no, 'no_injection-rpm')
           1.0
           0.8
         Data Values
6.0
           0.6
           0.2
           0.0
                0.0
                              0.2
                                            0.4
                                                          0.6
                                                                        0.8
                                                                                      1.0
                                                   RPM
Out[7]: 0
```

In []: