

```

1 import numpy as np
2 from numpy import savetxt
3 import pandas as pd
4 from pandas.plotting import register_matplotlib_converters
5 import seaborn as sns
6 from pylab import rcParams
7 import matplotlib.pyplot as plt
8 from matplotlib import rc
9 from sklearn.model_selection import train_test_split
10 from sklearn.preprocessing import MinMaxScaler, StandardScaler
11 from sklearn.metrics import r2_score
12 import tensorflow as tf
13 from tensorflow import keras
14
15 %matplotlib inline
16 %config InlineBackend.figure_format='retina'
17 register_matplotlib_converters()

```

```

1 source = '../Dataset.csv'
2 df = pd.read_csv(source, index_col=0)
3 df = df[['fc', 'n', 'fhy', 'load', 'ph', 'fsy', 'ds', 's', 'ps', 'dh', 'displacement']]
4 df.head()

```

```

1 train_size = int(len(df) * 0.8)
2 test_size = len(df) - train_size
3 train, test = df.iloc[0:train_size], df.iloc[train_size:len(df)]
4 print(len(train), len(test))

```

```

1 features = ['fc', 'n', 'fhy', 'ph', 'fsy', 'ds', 's', 'ps', 'dh', 'displacement']
2
3 feature_scaler = MinMaxScaler()
4 intensity_scaler = MinMaxScaler()
5
6 feature_scaler = feature_scaler.fit(train[features].to_numpy())
7 intensity_scaler = intensity_scaler.fit(train[['load']])
8
9 train.loc[:, features] = feature_scaler.transform(train[features].to_numpy())
10 train[['load']] = intensity_scaler.transform(train[['load']])
11
12 test.loc[:, features] = feature_scaler.transform(test[features].to_numpy())
13 test[['load']] = intensity_scaler.transform(test[['load']])

```

```

1 def training_data(X, y, time_steps=1):
2     Xs, ys = [], []
3     for i in range(len(X) - time_steps):
4         v = X.iloc[i:(i + time_steps)].values
5         Xs.append(v)
6         ys.append(y.iloc[i + time_steps])
7     return np.array(Xs), np.array(ys)

```

```

1 time_steps = 40 #Make predictions on the past 40 time_steps = 10h
2
3 # reshape to [samples, time_steps, n_features]
4
5 X_train, y_train = training_data(train, train.load, time_steps)
6 x_test, y_test = training_data(test, test.load, time_steps)
7
8 print(X_train.shape, y_train.shape)
9 print(x_test.shape, y_test.shape)

```

```

1 def model_vLSTM(units):
2     model = keras.Sequential()
3     #input
4     model.add(keras.layers.LSTM(
5         units=units,

```

```

6         activation="relu",
7         input_shape=(X_train.shape[1], X_train.shape[2])
8     ))
9     model.add(keras.layers.Dropout(0.2))
10    model.add(keras.layers.Dense(1))
11    model.compile(loss='mse', optimizer='adam')
12    return model

```

```

1 def model_tLSTM_h2(units):
2     model = keras.Sequential()
3     #Input Layer
4     model.add(keras.layers.LSTM(
5         units=units,
6         activation="relu",
7         return_sequences=True,
8         input_shape=(X_train.shape[1], X_train.shape[2])
9     ))
10    model.add(keras.layers.Dropout(0.2))
11    #Hidden Layer
12    model.add(keras.layers.LSTM(
13        units=units,
14    ))
15    model.add(keras.layers.Dropout(0.2))
16    #Output Layer
17    model.add(keras.layers.Dense(1))
18    model.compile(loss='mse', optimizer='adam')
19    return model

```

```

1 def model_tLSTM_h3(units):
2     model = keras.Sequential()
3     #Input Layer
4     model.add(keras.layers.LSTM(
5         units=units,
6         activation="relu",
7         return_sequences=True,
8         input_shape=(X_train.shape[1], X_train.shape[2])
9     ))
10    model.add(keras.layers.Dropout(0.2))
11    #Hidden Layers
12    model.add(keras.layers.LSTM(
13        units=units,
14        activation="relu",
15        return_sequences=True,
16    ))
17    model.add(keras.layers.Dropout(0.2))
18    model.add(keras.layers.LSTM(
19        units=units,
20        activation="relu",
21    ))
22    model.add(keras.layers.Dropout(0.2))
23    #Output Layer
24    model.add(keras.layers.Dense(1))
25    model.compile(loss='mse', optimizer='adam')
26    return model

```

```

1 def model_tLSTM_h4(units):
2     model = keras.Sequential()
3     #input
4     model.add(keras.layers.LSTM(
5         units=units,
6         activation="relu",
7         return_sequences=True,
8         input_shape=(X_train.shape[1], X_train.shape[2])
9     ))
10    model.add(keras.layers.Dropout(0.2))
11    model.add(keras.layers.LSTM(
12        units=units,
13        activation="relu",
14        return_sequences=True,
15    ))
16    model.add(keras.layers.Dropout(0.2))

```

```

17     model.add(keras.layers.LSTM(
18         units=units,
19         activation="relu",
20         return_sequences=True,
21     ))
22     model.add(keras.layers.Dropout(0.2))
23     model.add(keras.layers.LSTM(
24         units=units,
25         activation="relu",
26     ))
27     model.add(keras.layers.Dropout(0.2))
28     model.add(keras.layers.Dense(1))
29     model.compile(loss='mse', optimizer='adam')
30     return model

```

```

1 def model_BiLSTM(units):
2     model = keras.Sequential()
3     #Input Layer
4     model.add(keras.layers.Bidirectional(
5         keras.layers.LSTM(
6             units=units,
7             activation="relu",
8             input_shape=(X_train.shape[1], X_train.shape[2])
9         )))
10    model.add(keras.layers.Dropout(0.2))
11    #Hidden Layer
12    model.add(keras.layers.Dense(1))
13    model.compile(loss='mse', optimizer='adam')
14    return model

```

```

1 def model_GRU(units):
2     model = keras.Sequential()
3     #input
4     model.add(keras.layers.GRU(
5         units=units,
6         activation="relu",
7         input_shape=(X_train.shape[1], X_train.shape[2])
8     ))
9     model.add(keras.layers.Dropout(0.2))
10    model.add(keras.layers.Dense(1))
11    model.compile(loss='mse', optimizer='adam')
12    return model

```

```

1 def model_BiGRU(units):
2     model = keras.Sequential()
3     #input
4     model.add(keras.layers.Bidirectional(
5         keras.layers.GRU(
6             units=units,
7             activation="relu",
8             input_shape=(X_train.shape[1], X_train.shape[2])
9         )))
10    model.add(keras.layers.Dropout(0.2))
11    model.add(keras.layers.Dense(1))
12    model.compile(loss='mse', optimizer='adam')
13    return model

```

```

1 def model_BiGRU_h2(units):
2     model = keras.Sequential()
3     #input
4     model.add(keras.layers.Bidirectional(
5         keras.layers.GRU(
6             units=units,
7             return_sequences=True,
8             activation="relu",
9             input_shape=(X_train.shape[1], X_train.shape[2])
10        )))
11    model.add(keras.layers.Dropout(0.2))
12    #Hidden Layers

```

```

13     model.add(keras.layers.Bidirectional(
14         keras.layers.GRU(
15             units=units,
16             activation="relu",
17         )))
18     model.add(keras.layers.Dropout(0.2))
19     model.add(keras.layers.Dense(1))
20     model.compile(loss='mse', optimizer='adam')
21     return model

```

```

1 def fit_model(model):
2     early_stop = keras.callbacks.EarlyStopping(
3         monitor = 'val_loss',
4         min_delta = 0.0,
5         patience = 10)
6     history = model.fit(
7         X_train, y_train,
8         epochs = 100,
9         validation_split = 0.1,
10        batch_size = 64,
11        shuffle = False,
12        callbacks = [early_stop])
13     return history

```

```

1 # Plot train loss and validation loss
2 def plot_loss (history, model_name):
3     plt.figure(figsize = (8, 4))
4     plt.plot(history.history['loss'], color = 'blue', label='Train Loss')
5     plt.plot(history.history['val_loss'], color = 'red', label='Validation Loss')
6     plt.title('Train vs. Validation Loss for ' + model_name)
7     plt.ylabel('Loss')
8     plt.xlabel('epoch')
9     plt.legend(loc='upper right')
10    plt.savefig('G:/Failure mode prediction/Cycling column/Method2/' + 'Loss_' + model_name + '.jpg', format='jpg')

```

```

1 def save_model(model, model_name):
2     model.save('G:/Failure mode prediction/Cycling column/Method2/' + model_name)
3     return None

```

```

1 def prediction(model, x_test, y_test):
2     y_pred = model.predict(x_test)
3     y_train_inv = intensity_scaler.inverse_transform(y_train.reshape(1, -1))
4     y_test_inv = intensity_scaler.inverse_transform(y_test.reshape(1, -1))
5     y_pred_inv = intensity_scaler.inverse_transform(y_pred)
6     return y_pred, y_pred_inv, y_train_inv, y_test_inv

```

```

1 def plot_forecast(prediction_model, y_test, model_name):
2     y_pred, y_pred_inv, y_train_inv, y_test_inv = prediction_model
3     plt.plot(np.arange(0, len(y_train)), y_train_inv.flatten(), 'g', label="history")
4     plt.plot(np.arange(len(y_train), len(y_train) + len(y_test)), y_test_inv.flatten(), marker='.', label="Test data")
5     plt.plot(np.arange(len(y_train), len(y_train) + len(y_test)), y_pred_inv.flatten(), 'r', label="Prediction")
6     plt.ylabel('Load (kN)')
7     plt.xlabel('Time Step (s)')
8     plt.title('Prediction of applied load ' + model_name)
9     plt.legend()
10    plt.savefig('G:/Failure mode prediction/Cycling column/Method2/' + 'Forecast_' + model_name + '.jpg', format='jpg')
11    plt.show();

```

```

1 def plot_compare(prediction_model, y_test, model_name):
2
3     y_pred, y_pred_inv, y_train_inv, y_test_inv = prediction_model
4     plt.plot(y_test_inv.flatten(), marker='.', label="Test data")
5     plt.plot(y_pred_inv.flatten(), 'r', label="Prediction " + model_name)
6     plt.ylabel('Load (kN)')
7     plt.xlabel('Time Step (s)')
8     plt.title('Prediction of applied load ' + model_name)
9     plt.legend()

```

```

10 plt.savefig('G:/Failure mode prediction/Cycling column/Method2/'+TestPrediction_+model_name+'.jpg',
11 plt.show();

```

```

1 def plot_compare_models_2(prediction_model_1, prediction_model_2 ,y_test, model_name_1, model_name_2):
2
3     y_pred_1, y_pred_inv_1, y_train_inv_1, y_test_inv_1 = prediction_model_1
4     y_pred_2, y_pred_inv_2, y_train_inv_2, y_test_inv_2 = prediction_model_2
5
6     plt.plot(y_test_inv_1.flatten(), marker='.', label="Test data")
7     plt.plot(y_pred_inv_1.flatten(), 'r', label="Prediction "+model_name_1)
8     plt.plot(y_pred_inv_2.flatten(), 'g', label="Prediction "+model_name_2)
9     plt.ylabel('Load (kN)')
10    plt.xlabel('Time Step (s)')
11    plt.title('Test data vs Prediction data: '+model_name_1+' vs. '+model_name_2)
12    plt.legend()
13    plt.savefig('G:/Failure mode prediction/Cycling column/Method2/'+TestPrediction_+model_name_1+'vs'+model_name_2+'.jpg')
14    plt.show();

```

```

1 def plot_compare_models_3(prediction_model_1, prediction_model_2, prediction_model_3, y_test, model_name_1,
2
3     y_pred_1, y_pred_inv_1, y_train_inv_1, y_test_inv_1 = prediction_model_1
4     y_pred_2, y_pred_inv_2, y_train_inv_2, y_test_inv_2 = prediction_model_2
5     y_pred_3, y_pred_inv_3, y_train_inv_3, y_test_inv_3 = prediction_model_3
6
7     plt.plot(y_test_inv_1.flatten(), marker='.', label="Test data")
8     plt.plot(y_pred_inv_1.flatten(), 'r', label="Prediction "+model_name_1)
9     plt.plot(y_pred_inv_2.flatten(), 'g', label="Prediction "+model_name_2)
10    plt.plot(y_pred_inv_3.flatten(), 'gray', label="Prediction "+model_name_3)
11
12    plt.ylabel('Load (kN)')
13    plt.xlabel('Time Step (s)')
14    plt.title('Test data vs Prediction data: '+model_name_1+' vs. '+model_name_2+' vs. '+model_name_3)
15    plt.legend()
16    plt.savefig('G:/Failure mode prediction/Cycling column/Method2/'+TestPrediction_+model_name_1+'vs'+model_name_2+'vs'+model_name_3+'.jpg')
17    plt.show();

```

```

1 def evaluate_performance(prediction_model, y_test, model_name):
2     y_pred, y_pred_inv, y_train_inv, y_test_inv = prediction_model
3
4     y_test_inv = y_test_inv.reshape(-1,1)
5     errors = y_pred_inv - y_test_inv
6
7
8     mae = round(np.abs(errors).mean(),2)
9     mse = round(np.square(errors).mean(),2)
10    rmse = round(np.sqrt(mse),2)
11    r_sq = round(r2_score(y_test, y_pred),2)
12
13    performance = pd.DataFrame({"Model": [model_name], "MAE": [mae], "MSE": [mse], "RMSE": [rmse], "R^2": [r_sq]})
14    performance.to_csv('G:/Failure mode prediction/Cycling column/Method2/'+Measures_+model_name+'.csv')
15
16    print('Performance of '+model_name+' :')
17    print('MAE:\t', mae)
18    print('MSE:\t', mse)
19    print('RMSE:\t', rmse)
20    print('R^2:\t', r_sq)
21    return mae, mse, rmse, r_sq

```

```

1 vLSTM_n4 = model_vLSTM(4)
2 vLSTM_n8 = model_vLSTM(8)
3 vLSTM_n16 = model_vLSTM(16)
4 vLSTM_n32 = model_vLSTM(32)

```

```

1 # LSTM models with 32 hidden neurons.
2 tLSTM_n32h2 = model_tLSTM_h2(32)
3 tLSTM_n32h3 = model_tLSTM_h3(32)
4 tLSTM_n32h4 = model_tLSTM_h4(32)
5 GRU_n32 = model_GRU(32)

```

```
6 BiLSTM_n32 = model_BiLSTM(32)
7 BiGRU_n32 = model_BiGRU(32)
```

```
1 history_vLSTM_n4 = fit_model(vLSTM_n4)
2 history_vLSTM_n8 = fit_model(vLSTM_n8)
3 history_vLSTM_n16 = fit_model(vLSTM_n16)
4 history_vLSTM_n32 = fit_model(vLSTM_n32)
```

```
1 history_tLSTM_n32h2 = fit_model(tLSTM_n32h2)
2 history_tLSTM_n32h3 = fit_model(tLSTM_n32h3)
3 history_tLSTM_n32h4 = fit_model(tLSTM_n32h4)
4 history_GRU_n32 = fit_model(GRU_n32)
5 history_BiLSTM_n32 = fit_model(BiLSTM_n32)
6 history_BiGRU_n32 = fit_model(BiGRU_n32)
```

```
1 plot_loss(history_vLSTM_n4, 'vLSTM_n4')
2 plot_loss(history_vLSTM_n8, 'vLSTM_n8')
3 plot_loss(history_vLSTM_n16, 'vLSTM_n16')
4 plot_loss(history_vLSTM_n32, 'vLSTM_n32')
```

```
1 plot_loss(history_tLSTM_n32h2, 'tLSTM_n32h2')
2 plot_loss(history_tLSTM_n32h3, 'tLSTM_n32h3')
3 plot_loss(history_tLSTM_n32h4, 'tLSTM_n32h4')
4 plot_loss(history_GRU_n32, 'GRU_n32')
5 plot_loss(history_BiLSTM_n32, 'BiLSTM_n32')
6 plot_loss(history_BiGRU_n32, 'BiGRU_n32')
```

```
1 # Calculate predictions neurons
2 prediction_vLSTM_n4 = prediction(vLSTM_n4, x_test, y_test)
3 prediction_vLSTM_n8 = prediction(vLSTM_n8, x_test, y_test)
4 prediction_vLSTM_n16 = prediction(vLSTM_n16, x_test, y_test)
5 prediction_vLSTM_n32 = prediction(vLSTM_n32, x_test, y_test)
```

```
1 # Calculate predictions layers
2 prediction_tLSTM_n32h2 = prediction(tLSTM_n32h2, x_test, y_test)
3 prediction_tLSTM_n32h3 = prediction(tLSTM_n32h3, x_test, y_test)
4 prediction_tLSTM_n32h4 = prediction(tLSTM_n32h4, x_test, y_test)
5 prediction_GRU_n32 = prediction(GRU_n32, x_test, y_test)
6 prediction_BiLSTM_n32 = prediction(BiLSTM_n32, x_test, y_test)
7 prediction_BiGRU_n32 = prediction(BiGRU_n32, x_test, y_test)
```

```
1 plot_forecast(prediction_vLSTM_n4, y_test, 'vLSTM_n4')
2 plot_forecast(prediction_vLSTM_n8, y_test, 'vLSTM_n8')
3 plot_forecast(prediction_vLSTM_n16, y_test, 'vLSTM_n16')
4 plot_forecast(prediction_vLSTM_n32, y_test, 'vLSTM_n32')
```

```
1 plot_forecast(prediction_tLSTM_n32h2, y_test, 'tLSTM_n32h2')
2 plot_forecast(prediction_tLSTM_n32h3, y_test, 'tLSTM_n32h3')
3 plot_forecast(prediction_tLSTM_n32h4, y_test, 'tLSTM_n32h4')
4 plot_forecast(prediction_GRU_n32, y_test, 'GRU_n32')
5 plot_forecast(prediction_BiLSTM_n32, y_test, 'BiLSTM_n32')
6 plot_forecast(prediction_BiGRU_n32, y_test, 'BiGRU_n32')
```

```
1 plot_compare(prediction_vLSTM_n4, y_test, 'vLSTM_n4')
2 plot_compare(prediction_vLSTM_n8, y_test, 'vLSTM_n8')
3 plot_compare(prediction_vLSTM_n16, y_test, 'vLSTM_n16')
4 plot_compare(prediction_vLSTM_n32, y_test, 'vLSTM_n32')
```

```
1 plot_compare(prediction_tLSTM_n32h2, y_test, 'tLSTM_n32h2')
2 plot_compare(prediction_tLSTM_n32h3, y_test, 'tLSTM_n32h3')
3 plot_compare(prediction_tLSTM_n32h4, y_test, 'tLSTM_n32h4')
4 plot_compare(prediction_GRU_n32, y_test, 'GRU_n32')
```

```
5 plot_compare(prediction_BiLSTM_n32, y_test, 'BiLSTM_n32')
6 plot_compare(prediction_BiGRU_n32, y_test, 'BiGRU_n32')
```

```
1 evaluate_performance(prediction_vLSTM_n4, y_test, 'vLSTM_n4')
2 evaluate_performance(prediction_vLSTM_n8, y_test, 'vLSTM_n8')
3 evaluate_performance(prediction_vLSTM_n16, y_test, 'vLSTM_n16')
4 evaluate_performance(prediction_vLSTM_n32, y_test, 'vLSTM_n32')
```

```
1 evaluate_performance(prediction_tLSTM_n32h2, y_test, 'tLSTM_n32h2')
2 evaluate_performance(prediction_tLSTM_n32h3, y_test, 'tLSTM_n32h3')
3 evaluate_performance(prediction_tLSTM_n32h4, y_test, 'tLSTM_n32h4')
4 evaluate_performance(prediction_GRU_n32, y_test, 'GRU_n32')
5 evaluate_performance(prediction_BiLSTM_n32, y_test, 'BiLSTM_n32')
6 evaluate_performance(prediction_BiGRU_n32, y_test, 'BiGRU_n32')
```

```
1 dataframe = pd.DataFrame({'y_pred_inv':prediction_tLSTM_n32h2})
2 dataframe.to_csv("outputn32h2.csv", index=True, sep=',')
```

