

CMAPSS_FD002_1D_Conv

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```
[1]: import tensorflow as tf
from tensorflow.keras import layers, models
from tensorflow.keras.models import Sequential
import keras

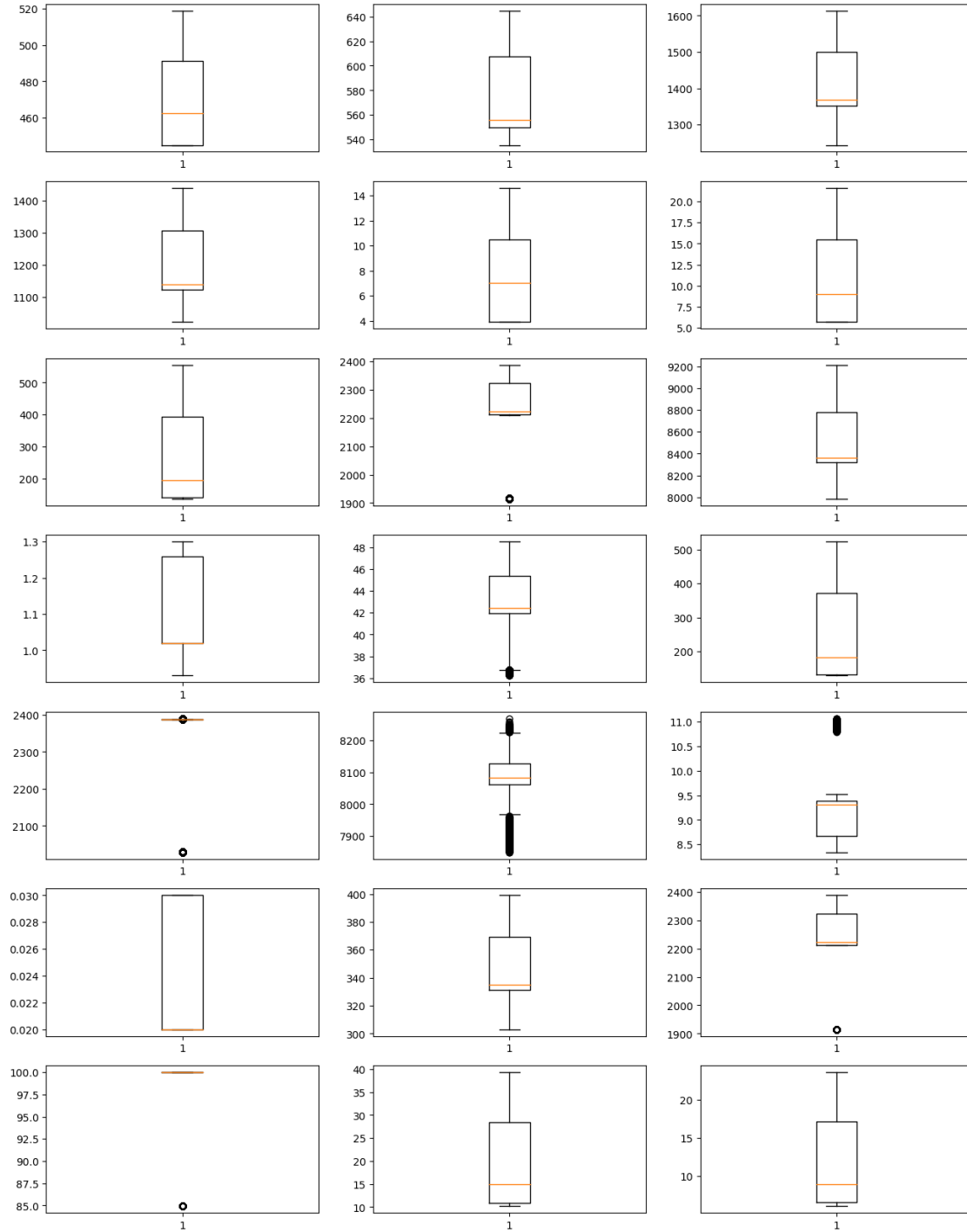
import numpy as np
import pandas as pd
import sklearn
from sklearn.model_selection import train_test_split

from sklearn.preprocessing import MinMaxScaler
from sklearn.metrics import mean_squared_error
import matplotlib.pyplot as plt

np.random.seed(100)
```

1 Data Preprocessing

```
[2]: train_data = pd.read_csv("train_FD002.txt", sep= "\s+", header = None)
plt.figure(figsize = (16, 21))
for i in range(21):
    temp_data = train_data.iloc[:,i+5]
    plt.subplot(7,3,i+1)
    plt.boxplot(temp_data)
plt.show()
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[3]: def process_targets(data_length, early_rul = None):
    if early_rul == None:
        return np.arange(data_length-1, -1, -1)
    else:
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        early_rul_duration = data_length - early_rul
        if early_rul_duration <= 0:
            return np.arange(data_length-1, -1, -1)
        else:
            return np.append(early_rul*np.ones(shape = (early_rul_duration,)),
↪np.arange(early_rul-1, -1, -1))

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[4]: def process_input_data_with_targets(input_data, target_data = None,
↪window_length = 1, shift = 1):
    num_batches = int(np.floor((len(input_data) - window_length)/shift)) + 1
    num_features = input_data.shape[1]
    output_data = np.repeat(np.nan, repeats = num_batches * window_length *
↪num_features).reshape(num_batches, window_length,
    num_features)
    if target_data is None:
        for batch in range(num_batches):
            output_data[batch, :, :] = input_data[(0+shift*batch):
↪(0+shift*batch+window_length), :]
        return output_data
    else:
        output_targets = np.repeat(np.nan, repeats = num_batches)
        for batch in range(num_batches):
            output_data[batch, :, :] = input_data[(0+shift*batch):
↪(0+shift*batch+window_length), :]
            output_targets[batch] = target_data[(shift*batch +
↪(window_length-1))]
        return output_data, output_targets

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[5]: def process_test_data(test_data_for_an_engine, window_length, shift,
↪num_test_windows = 1):
    max_num_test_batches = int(np.floor((len(test_data_for_an_engine) -
↪window_length)/shift)) + 1
    if max_num_test_batches < num_test_windows:
        required_len = (max_num_test_batches - 1)* shift + window_length
        batched_test_data_for_an_engine =
↪process_input_data_with_targets(test_data_for_an_engine[-required_len:, :],
        target_data = None,
    window_length = window_length, shift = shift)
    return batched_test_data_for_an_engine, max_num_test_batches
    else:
        required_len = (num_test_windows - 1) * shift + window_length
        batched_test_data_for_an_engine =
↪process_input_data_with_targets(test_data_for_an_engine[-required_len:, :],

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    ↪target_data = None,

    ↪window_length = window_length, shift = shift)
    return batched_test_data_for_an_engine, num_test_windows

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[6]: test_data = pd.read_csv("test_FD002.txt", sep = "\s+", header = None)
true_rul = pd.read_csv("RUL_FD002.txt", sep = '\s+', header = None)

window_length = 20
shift = 1
early_rul = 130
processed_train_data = []
processed_train_targets = []
num_test_windows = 5
processed_test_data = []
num_test_windows_list = []

columns_to_be_dropped = [0,1,2,3,4,5,9,10,14,20,22,23]

train_data_first_column = train_data[0]
test_data_first_column = test_data[0]

scaler = MinMaxScaler(feature_range = (-1,1))
train_data = scaler.fit_transform(train_data.drop(columns =
    ↪columns_to_be_dropped))
test_data = scaler.transform(test_data.drop(columns = columns_to_be_dropped))

train_data = pd.DataFrame(data = np.c_[train_data_first_column, train_data])
test_data = pd.DataFrame(data = np.c_[test_data_first_column, test_data])

num_train_machines = len(train_data[0].unique())
num_test_machines = len(test_data[0].unique())

for i in np.arange(1, num_train_machines + 1):
    temp_train_data = train_data[train_data[0] == i].drop(columns = [0]).values

    if (len(temp_train_data) < window_length):
        print("Train engine {} doesn't have enough data for window_length of
    ↪{}".format(i, window_length))
        raise AssertionError("Window length is larger than number of data
    ↪points for some engines. "
                               "Try decreasing window length.")

    temp_train_targets = process_targets(data_length = temp_train_data.
    ↪shape[0], early_rul = early_rul)

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    data_for_a_machine, targets_for_a_machine =
    ↪process_input_data_with_targets(temp_train_data, temp_train_targets,

    ↪window_length = window_length, shift = shift)

    processed_train_data.append(data_for_a_machine)
    processed_train_targets.append(targets_for_a_machine)

processed_train_data = np.concatenate(processed_train_data)
processed_train_targets = np.concatenate(processed_train_targets)

for i in np.arange(1, num_test_machines + 1):
    temp_test_data = test_data[test_data[0] == i].drop(columns = [0]).values

    if (len(temp_test_data) < window_length):
        print("Test engine {} doesn't have enough data for window_length of {}".
    ↪format(i, window_length))
        raise AssertionError("Window length is larger than number of data
    ↪points for some engines. "
                               "Try decreasing window length.")

    test_data_for_an_engine, num_windows = process_test_data(temp_test_data,
    ↪window_length = window_length, shift = shift,

    num_test_windows =
    ↪num_test_windows)

    processed_test_data.append(test_data_for_an_engine)
    num_test_windows_list.append(num_windows)

processed_test_data = np.concatenate(processed_test_data)
true_rul = true_rul[0].values

index = np.random.permutation(len(processed_train_targets))
processed_train_data, processed_train_targets = processed_train_data[index],
    ↪processed_train_targets[index]

print("Processed trianing data shape: ", processed_train_data.shape)
print("Processed training ruls shape: ", processed_train_targets.shape)
print("Processed test data shape: ", processed_test_data.shape)
print("True RUL shape: ", true_rul.shape)

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Processed trianing data shape: (48819, 20, 14)
Processed training ruls shape: (48819,)
Processed test data shape: (1289, 20, 14)
True RUL shape: (259,)

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[7]: processed_train_data, processed_val_data, processed_train_targets,
      ↪ processed_val_targets = train_test_split(processed_train_data,
      ↪
      ↪ processed_train_targets,
      ↪
      ↪ test_size = 0.2,
      ↪
      ↪ random_state = 83)
print("Processed train data shape: ", processed_train_data.shape)
print("Processed validation data shape: ", processed_val_data.shape)
print("Processed train targets shape: ", processed_train_targets.shape)
print("Processed validation targets shape: ", processed_val_targets.shape)
```

```
Processed train data shape: (39055, 20, 14)
Processed validation data shape: (9764, 20, 14)
Processed train targets shape: (39055,)
Processed validation targets shape: (9764,)
```

2 1D CNN model

```
[8]: def scheduler(epoch):
      if epoch < 50:
          return 0.001
      else:
          return 0.0001
```

```
[9]: runs = 5

test_rmse = np.zeros(runs)

for i in range(runs):

    model = models.Sequential([
        layers.Conv1D(32, 3, padding = 'same', activation='relu', input_shape =
        ↪ (window_length, processed_train_data.shape[2])),
        layers.MaxPooling1D(2),
        layers.Conv1D(64, 3, padding = 'same', activation='relu'),
        layers.MaxPooling1D(2),
        layers.Conv1D(128, 3, padding = 'same', activation='relu'),
        layers.MaxPooling1D(2),
        layers.Flatten(),
        layers.Dense(window_length*processed_train_data.shape[2],
        ↪ activation='relu'),
        layers.Dropout(0.2),
        layers.Dense(1)
    ])

    # Train the model
    model.compile(optimizer='adam', loss='mse', metrics=['accuracy'])
    model.fit(processed_train_data, processed_train_targets,
              validation_data=(processed_val_data, processed_val_targets),
              epochs=50, batch_size=32, callbacks=[scheduler])

    # Evaluate the model
    test_rmse[i] = model.evaluate(processed_val_data, processed_val_targets,
                                  verbose=0)
```

```

model.compile(optimizer='adam',
              loss='mean_squared_error',
              metrics=[keras.metrics.RootMeanSquaredError()])

callback = tf.keras.callbacks.LearningRateScheduler(scheduler, verbose =  

↳False)

history = model.fit(processed_train_data, processed_train_targets, epochs =  

↳150,
                    validation_data = (processed_val_data,  

↳processed_val_targets),
                    callbacks = callback,
                    batch_size = 512, verbose = False)

rul_pred = model.predict(processed_test_data).reshape(-1)
preds_for_each_engine = np.split(rul_pred, np.cumsum(num_test_windows_list)[:  

↳-1])

indices_of_last_examples = np.cumsum(num_test_windows_list) - 1
preds_for_last_example = np.  

↳concatenate(preds_for_each_engine)[indices_of_last_examples]

RMSE = np.sqrt(mean_squared_error(true_rul, preds_for_last_example))

test_rmse[i] = RMSE

print('RMSE in trial', i+1, ':', test_rmse[i])

print(f'Avg test RMSE over 5 iterations : {test_rmse.mean()}')

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```

41/41 [=====] - 0s 2ms/step
RMSE in trial 1 : 28.63465403455697
41/41 [=====] - 0s 4ms/step
RMSE in trial 2 : 30.190094467753276
41/41 [=====] - 0s 2ms/step
RMSE in trial 3 : 29.214307074236984
41/41 [=====] - 0s 2ms/step
RMSE in trial 4 : 29.324645655655218
41/41 [=====] - 0s 2ms/step
RMSE in trial 5 : 31.009542220258325
Avg test RMSE over 5 iterations : 29.674648690492155

```

[10]: model.summary()

Model: "sequential_4"

Layer (type)	Output Shape	Param #
=====		

conv1d_12 (Conv1D)	(None, 20, 32)	1376
max_pooling1d_12 (MaxPooling1D)	(None, 10, 32)	0
conv1d_13 (Conv1D)	(None, 10, 64)	6208
max_pooling1d_13 (MaxPooling1D)	(None, 5, 64)	0
conv1d_14 (Conv1D)	(None, 5, 128)	24704
max_pooling1d_14 (MaxPooling1D)	(None, 2, 128)	0
flatten_4 (Flatten)	(None, 256)	0
dense_8 (Dense)	(None, 280)	71960
dropout_4 (Dropout)	(None, 280)	0
dense_9 (Dense)	(None, 1)	281

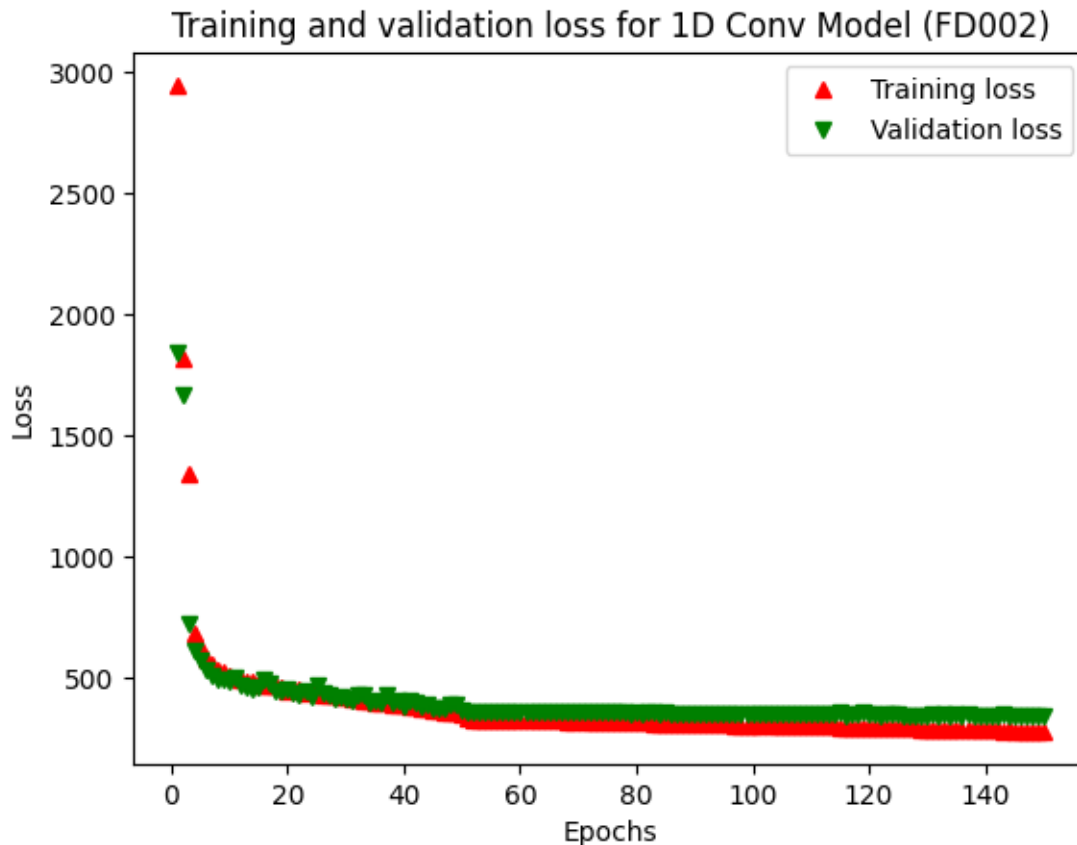
```
=====
Total params: 104529 (408.32 KB)
Trainable params: 104529 (408.32 KB)
Non-trainable params: 0 (0.00 Byte)
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```

```
[11]: loss = history.history['loss']
      val_loss = history.history['val_loss']

      epochs = range(1, len(loss) + 1)

      plt.plot(epochs, loss, 'r^', label='Training loss')
      plt.plot(epochs, val_loss, 'gv', label='Validation loss')
      plt.title('Training and validation loss for 1D Conv Model (FD002)')
      plt.xlabel('Epochs')
      plt.ylabel('Loss')
      plt.legend()

      plt.show()
```

```
[12]: def compute_s_score(rul_true, rul_pred):
        diff = rul_pred - rul_true
        return np.sum(np.where(diff < 0, np.exp(-diff/13)-1, np.exp(diff/10)-1))
```

```
[13]: s_score = compute_s_score(true_rul, preds_for_last_example)
        print("S-score: ", s_score)
```

S-score: 33331.88055886224

```
[14]: y_pred_series = pd.Series(preds_for_last_example.flatten())
        y_test_series = pd.Series(true_rul.flatten())

        df = pd.DataFrame({'y_pred_fd002': y_pred_series, 'y_test_fd002':
            ↳y_test_series})

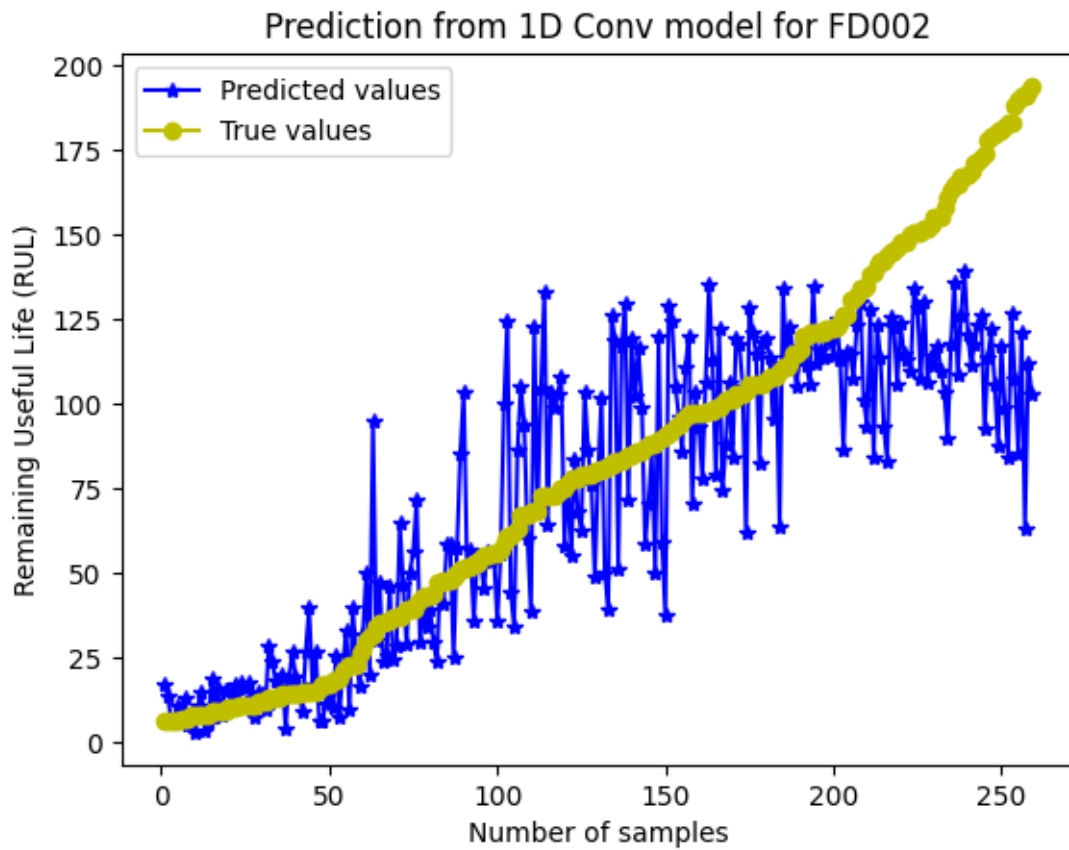
        df = df.sort_values('y_test_fd002')

        number_of_sample = range(1, len(true_rul) + 1)

        plt.plot(number_of_sample, df['y_pred_fd002'], 'b*-', label='Predicted values')
```

```
plt.plot(number_of_sample, df['y_test_fd002'], 'yo-', label='True values')
plt.title('Prediction from 1D Conv model for FD002')
plt.xlabel('Number of samples')
plt.ylabel('Remaining Useful Life (RUL)')
plt.legend()

plt.show()
```



[14]: