

C4W2_Assignment

April 4, 2023

1 Week 2: Predicting time series

Welcome! In the previous assignment you got some exposure to working with time series data, but you didn't use machine learning techniques for your forecasts. This week you will be using a deep neural network to create forecasts to see how this technique compares with the ones you already tried out. Once again all of the data is going to be generated.

Let's get started!

```
[1]: import numpy as np
import tensorflow as tf
import matplotlib.pyplot as plt
from dataclasses import dataclass

# import warnings
# warnings.filterwarnings('ignore')
```

1.1 Generating the data

The next cell includes a bunch of helper functions to generate and plot the time series:

```
[2]: def plot_series(time, series, format="-", start=0, end=None):
    plt.plot(time[start:end], series[start:end], format)
    plt.xlabel("Time")
    plt.ylabel("Value")
    plt.grid(False)

def trend(time, slope=0):
    return slope * time

def seasonal_pattern(season_time):
    """Just an arbitrary pattern, you can change it if you wish"""
    return np.where(season_time < 0.1,
                    np.cos(season_time * 6 * np.pi),
                    2 / np.exp(9 * season_time))

def seasonality(time, period, amplitude=1, phase=0):
    """Repeats the same pattern at each period"""
    season_time = ((time + phase) % period) / period
```

```

    return amplitude * seasonal_pattern(season_time)

def noise(time, noise_level=1, seed=None):
    rnd = np.random.RandomState(seed)
    return rnd.randn(len(time)) * noise_level

```

You will be generating time series data that greatly resembles the one from last week but with some differences.

Notice that this time all the generation is done within a function and global variables are saved within a dataclass. This is done to avoid using global scope as it was done in during the previous week.

If you haven't used dataclasses before, they are just Python classes that provide a convenient syntax for storing data. You can read more about them in the [docs](#).

```

[3]: def generate_time_series():
    # The time dimension or the x-coordinate of the time series
    time = np.arange(4 * 365 + 1, dtype="float32")

    # Initial series is just a straight line with a y-intercept
    y_intercept = 10
    slope = 0.005
    series = trend(time, slope) + y_intercept

    # Adding seasonality
    amplitude = 50
    series += seasonality(time, period=365, amplitude=amplitude)

    # Adding some noise
    noise_level = 3
    series += noise(time, noise_level, seed=51)

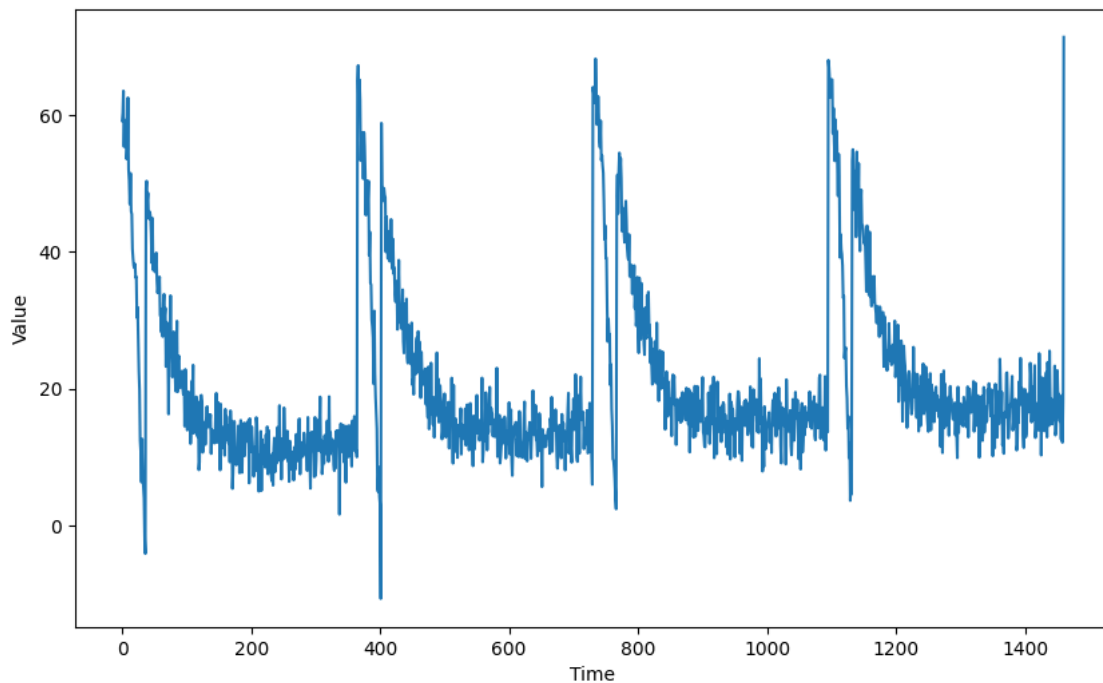
    return time, series

# Save all "global" variables within the G class (G stands for global)
@dataclass
class G:
    TIME, SERIES = generate_time_series()
    SPLIT_TIME = 1100
    WINDOW_SIZE = 20
    BATCH_SIZE = 32
    SHUFFLE_BUFFER_SIZE = 1000

# Plot the generated series
plt.figure(figsize=(10, 6))

```

```
plot_series(G.TIME, G.SERIES)
plt.show()
```



1.2 Splitting the data

Since you already coded the `train_val_split` function during last week's assignment, this time it is provided for you:

```
[4]: def train_val_split(time, series, time_step=G.SPLIT_TIME):

    time_train = time[:time_step]
    series_train = series[:time_step]
    time_valid = time[time_step:]
    series_valid = series[time_step:]

    return time_train, series_train, time_valid, series_valid

# Split the dataset
time_train, series_train, time_valid, series_valid = train_val_split(G.TIME, G.
↪SERIES)
```

1.3 Processing the data

As you saw on the lectures you can feed the data for training by creating a dataset with the appropriate processing steps such as **windowing**, **flattening**, **batching** and **shuffling**. To do so complete the `windowed_dataset` function below.

Notice that this function receives a `series`, `window_size`, `batch_size` and `shuffle_buffer` and the last three of these default to the “global” values defined earlier.

Be sure to check out the [docs](#) about TF Datasets if you need any help.

```
[5]: def windowed_dataset(series, window_size=G.WINDOW_SIZE, batch_size=G.  
    ↪BATCH_SIZE, shuffle_buffer=G.SHUFFLE_BUFFER_SIZE):  
  
    ### START CODE HERE  
  
    # Create dataset from the series  
    dataset = tf.data.Dataset.from_tensor_slices(series)  
  
    # Slice the dataset into the appropriate windows  
    dataset = dataset.window(window_size + 1, shift=1, drop_remainder=True)  
  
    # Flatten the dataset  
    dataset = dataset.flat_map(lambda window: window.batch(window_size + 1))  
  
    # Shuffle it  
    dataset = dataset.shuffle(shuffle_buffer)  
  
    # Split it into the features and labels  
    dataset = dataset.map(lambda window: (window[:-1], window[-1]))  
  
    # Batch it  
    dataset = dataset.batch(batch_size).prefetch(1)  
  
    ### END CODE HERE  
  
    return dataset
```

To test your function you will be using a `window_size` of 1 which means that you will use each value to predict the next one. This for 5 elements since a `batch_size` of 5 is used and no shuffle since `shuffle_buffer` is set to 1.

Given this, the batch of features should be identical to the first 5 elements of the `series_train` and the batch of labels should be equal to elements 2 through 6 of the `series_train`.

```
[6]: # Test your function with windows size of 1 and no shuffling  
test_dataset = windowed_dataset(series_train, window_size=1, batch_size=5,  
    ↪shuffle_buffer=1)  
  
# Get the first batch of the test dataset
```

```

batch_of_features, batch_of_labels = next((iter(test_dataset)))

print(f"batch_of_features has type: {type(batch_of_features)}\n")
print(f"batch_of_labels has type: {type(batch_of_labels)}\n")
print(f"batch_of_features has shape: {batch_of_features.shape}\n")
print(f"batch_of_labels has shape: {batch_of_labels.shape}\n")
print(f"batch_of_features is equal to first five elements in the series: {np.
    ↪allclose(batch_of_features.numpy().flatten(), series_train[:5])}\n")
print(f"batch_of_labels is equal to first five labels: {np.
    ↪allclose(batch_of_labels.numpy(), series_train[1:6])}")

```

Metal device set to: Apple M1 Pro

batch_of_features has type: <class
'tensorflow.python.framework.ops.EagerTensor'>

batch_of_labels has type: <class 'tensorflow.python.framework.ops.EagerTensor'>

batch_of_features has shape: (5, 1)

batch_of_labels has shape: (5,)

batch_of_features is equal to first five elements in the series: True

batch_of_labels is equal to first five labels: True

2023-04-04 13:39:35.258938: I

tensorflow/core/common_runtime/pluggable_device/pluggable_device_factory.cc:305]
Could not identify NUMA node of platform GPU ID 0, defaulting to 0. Your kernel
may not have been built with NUMA support.

2023-04-04 13:39:35.259301: I

tensorflow/core/common_runtime/pluggable_device/pluggable_device_factory.cc:271]
Created TensorFlow device (/job:localhost/replica:0/task:0/device:GPU:0 with 0
MB memory) -> physical PluggableDevice (device: 0, name: METAL, pci bus id:
<undefined>)

2023-04-04 13:39:35.320145: W

tensorflow/core/platform/profile_utils/cpu_utils.cc:128] Failed to get CPU
frequency: 0 Hz

Expected Output:

batch_of_features has type: <class 'tensorflow.python.framework.ops.EagerTensor'>

batch_of_labels has type: <class 'tensorflow.python.framework.ops.EagerTensor'>

batch_of_features has shape: (5, 1)

batch_of_labels has shape: (5,)

batch_of_features is equal to first five elements in the series: True

batch_of_labels is equal to first five labels: True

1.4 Defining the model architecture

Now that you have a function that will process the data before it is fed into your neural network for training, it is time to define your layer architecture.

Complete the `create_model` function below. Notice that this function receives the `window_size` since this will be an important parameter for the first layer of your network.

Hint: - You will only need `Dense` layers. - Do not include `Lambda` layers. These are not required and are incompatible with the `HDF5` format which will be used to save your model for grading. - The training should be really quick so if you notice that each epoch is taking more than a few seconds, consider trying a different architecture.

```
[7]: def create_model(window_size=G.WINDOW_SIZE):  
  
    ### START CODE HERE  
  
    model = tf.keras.models.Sequential([  
        tf.keras.layers.Dense(10, input_shape=[window_size], activation="relu"),  
        tf.keras.layers.Dense(10, activation="relu"),  
        tf.keras.layers.Dense(1)  
    ])  
  
    model.compile(loss="mse", optimizer=tf.keras.optimizers.  
↳SGD(learning_rate=1e-6, momentum=0.9))  
  
    ### END CODE HERE  
  
    return model
```

```
[8]: # Apply the processing to the whole training series  
dataset = windowed_dataset(series_train)  
  
# Save an instance of the model  
model = create_model()  
  
# Train it  
model.fit(dataset, epochs=100)
```

Epoch 1/100

2023-04-04 13:41:11.300951: I

tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:113]

Plugin optimizer for device_type GPU is enabled.

34/34 [=====] - 2s 8ms/step - loss: 192.9555

Epoch 2/100

34/34 [=====] - 0s 4ms/step - loss: 96.5044
 Epoch 3/100
 34/34 [=====] - 0s 3ms/step - loss: 84.9203
 Epoch 4/100
 34/34 [=====] - 0s 3ms/step - loss: 80.0639
 Epoch 5/100
 34/34 [=====] - 0s 4ms/step - loss: 76.6586
 Epoch 6/100
 34/34 [=====] - 0s 4ms/step - loss: 73.6693
 Epoch 7/100
 34/34 [=====] - 0s 3ms/step - loss: 71.1519
 Epoch 8/100
 34/34 [=====] - 0s 3ms/step - loss: 68.6055
 Epoch 9/100
 34/34 [=====] - 0s 3ms/step - loss: 66.5648
 Epoch 10/100
 34/34 [=====] - 0s 4ms/step - loss: 64.4103
 Epoch 11/100
 34/34 [=====] - 0s 4ms/step - loss: 62.6638
 Epoch 12/100
 34/34 [=====] - 0s 4ms/step - loss: 61.1362
 Epoch 13/100
 34/34 [=====] - 0s 4ms/step - loss: 59.1069
 Epoch 14/100
 34/34 [=====] - 0s 4ms/step - loss: 57.6609
 Epoch 15/100
 34/34 [=====] - 0s 3ms/step - loss: 55.9538
 Epoch 16/100
 34/34 [=====] - 0s 3ms/step - loss: 54.5744
 Epoch 17/100
 34/34 [=====] - 0s 3ms/step - loss: 53.2879
 Epoch 18/100
 34/34 [=====] - 0s 3ms/step - loss: 52.1563
 Epoch 19/100
 34/34 [=====] - 0s 3ms/step - loss: 51.0263
 Epoch 20/100
 34/34 [=====] - 0s 3ms/step - loss: 50.0063
 Epoch 21/100
 34/34 [=====] - 0s 3ms/step - loss: 48.7900
 Epoch 22/100
 34/34 [=====] - 0s 3ms/step - loss: 47.8667
 Epoch 23/100
 34/34 [=====] - 0s 3ms/step - loss: 47.0502
 Epoch 24/100
 34/34 [=====] - 0s 3ms/step - loss: 46.2533
 Epoch 25/100
 34/34 [=====] - 0s 4ms/step - loss: 45.4643
 Epoch 26/100

```

34/34 [=====] - 0s 4ms/step - loss: 44.9326
Epoch 27/100
34/34 [=====] - 0s 3ms/step - loss: 44.2529
Epoch 28/100
34/34 [=====] - 0s 3ms/step - loss: 43.6375
Epoch 29/100
34/34 [=====] - 0s 3ms/step - loss: 42.9935
Epoch 30/100
34/34 [=====] - 0s 3ms/step - loss: 42.5225
Epoch 31/100
34/34 [=====] - 0s 3ms/step - loss: 42.1094
Epoch 32/100
34/34 [=====] - 0s 3ms/step - loss: 41.6153
Epoch 33/100
34/34 [=====] - 0s 3ms/step - loss: 41.2127
Epoch 34/100
34/34 [=====] - 0s 3ms/step - loss: 40.7935
Epoch 35/100
34/34 [=====] - 0s 3ms/step - loss: 40.3341
Epoch 36/100
34/34 [=====] - 0s 3ms/step - loss: 39.9758
Epoch 37/100
34/34 [=====] - 0s 3ms/step - loss: 39.7668
Epoch 38/100
34/34 [=====] - 0s 3ms/step - loss: 39.3478
Epoch 39/100
34/34 [=====] - 0s 3ms/step - loss: 39.1001
Epoch 40/100
34/34 [=====] - 0s 3ms/step - loss: 38.8186
Epoch 41/100
34/34 [=====] - 0s 3ms/step - loss: 38.5580
Epoch 42/100
34/34 [=====] - 0s 3ms/step - loss: 38.2768
Epoch 43/100
34/34 [=====] - 0s 3ms/step - loss: 38.0937
Epoch 44/100
34/34 [=====] - 0s 3ms/step - loss: 37.8107
Epoch 45/100
34/34 [=====] - 0s 4ms/step - loss: 37.5670
Epoch 46/100
34/34 [=====] - 0s 4ms/step - loss: 37.3549
Epoch 47/100
34/34 [=====] - 0s 3ms/step - loss: 37.1396
Epoch 48/100
34/34 [=====] - 0s 4ms/step - loss: 36.9231
Epoch 49/100
34/34 [=====] - 0s 4ms/step - loss: 36.7452
Epoch 50/100

```


34/34 [=====] - 0s 4ms/step - loss: 36.5021
Epoch 51/100
34/34 [=====] - 0s 4ms/step - loss: 36.2846
Epoch 52/100
34/34 [=====] - 0s 4ms/step - loss: 36.1398
Epoch 53/100
34/34 [=====] - 0s 4ms/step - loss: 35.9257
Epoch 54/100
34/34 [=====] - 0s 4ms/step - loss: 35.8002
Epoch 55/100
34/34 [=====] - 0s 3ms/step - loss: 35.6089
Epoch 56/100
34/34 [=====] - 0s 3ms/step - loss: 35.4566
Epoch 57/100
34/34 [=====] - 0s 4ms/step - loss: 35.2966
Epoch 58/100
34/34 [=====] - 0s 4ms/step - loss: 35.0986
Epoch 59/100
34/34 [=====] - 0s 3ms/step - loss: 34.9641
Epoch 60/100
34/34 [=====] - 0s 3ms/step - loss: 34.8190
Epoch 61/100
34/34 [=====] - 0s 3ms/step - loss: 34.6415
Epoch 62/100
34/34 [=====] - 0s 3ms/step - loss: 34.5077
Epoch 63/100
34/34 [=====] - 0s 4ms/step - loss: 34.3641
Epoch 64/100
34/34 [=====] - 0s 3ms/step - loss: 34.3022
Epoch 65/100
34/34 [=====] - 0s 3ms/step - loss: 34.2151
Epoch 66/100
34/34 [=====] - 0s 3ms/step - loss: 34.0677
Epoch 67/100
34/34 [=====] - 0s 3ms/step - loss: 33.9512
Epoch 68/100
34/34 [=====] - 0s 3ms/step - loss: 33.7981
Epoch 69/100
34/34 [=====] - 0s 4ms/step - loss: 33.7210
Epoch 70/100
34/34 [=====] - 0s 3ms/step - loss: 33.6431
Epoch 71/100
34/34 [=====] - 0s 3ms/step - loss: 33.5049
Epoch 72/100
34/34 [=====] - 0s 3ms/step - loss: 33.4439
Epoch 73/100
34/34 [=====] - 0s 3ms/step - loss: 33.2859
Epoch 74/100

34/34 [=====] - 0s 4ms/step - loss: 33.2298
Epoch 75/100
34/34 [=====] - 0s 4ms/step - loss: 33.1201
Epoch 76/100
34/34 [=====] - 0s 4ms/step - loss: 33.0063
Epoch 77/100
34/34 [=====] - 0s 4ms/step - loss: 32.9386
Epoch 78/100
34/34 [=====] - 0s 4ms/step - loss: 32.7927
Epoch 79/100
34/34 [=====] - 0s 3ms/step - loss: 32.7944
Epoch 80/100
34/34 [=====] - 0s 4ms/step - loss: 32.7835
Epoch 81/100
34/34 [=====] - 0s 4ms/step - loss: 32.6446
Epoch 82/100
34/34 [=====] - 0s 4ms/step - loss: 32.4793
Epoch 83/100
34/34 [=====] - 0s 4ms/step - loss: 32.3602
Epoch 84/100
34/34 [=====] - 0s 3ms/step - loss: 32.3372
Epoch 85/100
34/34 [=====] - 0s 3ms/step - loss: 32.2192
Epoch 86/100
34/34 [=====] - 0s 4ms/step - loss: 32.0530
Epoch 87/100
34/34 [=====] - 0s 4ms/step - loss: 32.0610
Epoch 88/100
34/34 [=====] - 0s 3ms/step - loss: 31.9021
Epoch 89/100
34/34 [=====] - 0s 3ms/step - loss: 31.8364
Epoch 90/100
34/34 [=====] - 0s 3ms/step - loss: 31.7226
Epoch 91/100
34/34 [=====] - 0s 3ms/step - loss: 31.6180
Epoch 92/100
34/34 [=====] - 0s 3ms/step - loss: 31.5166
Epoch 93/100
34/34 [=====] - 0s 4ms/step - loss: 31.4925
Epoch 94/100
34/34 [=====] - 0s 3ms/step - loss: 31.4000
Epoch 95/100
34/34 [=====] - 0s 3ms/step - loss: 31.3349
Epoch 96/100
34/34 [=====] - 0s 3ms/step - loss: 31.2168
Epoch 97/100
34/34 [=====] - 0s 4ms/step - loss: 31.1975
Epoch 98/100

```

34/34 [=====] - 0s 4ms/step - loss: 31.0816
Epoch 99/100
34/34 [=====] - 0s 4ms/step - loss: 31.0525
Epoch 100/100
34/34 [=====] - 0s 3ms/step - loss: 30.9963

```

[8]: <keras.callbacks.History at 0x17fb09900>

1.5 Evaluating the forecast

Now it is time to evaluate the performance of the forecast. For this you can use the `compute_metrics` function that you coded in the previous assignment:

```

[9]: def compute_metrics(true_series, forecast):

    mse = tf.keras.metrics.mean_squared_error(true_series, forecast).numpy()
    mae = tf.keras.metrics.mean_absolute_error(true_series, forecast).numpy()

    return mse, mae

```

At this point only the model that will perform the forecast is ready but you still need to compute the actual forecast.

For this, run the cell below which uses the `generate_forecast` function to compute the forecast. This function generates the next value given a set of the previous `window_size` points for every point in the validation set.

```

[10]: def generate_forecast(series=G.SERIES, split_time=G.SPLIT_TIME, window_size=G.
    ↪WINDOW_SIZE):
    forecast = []
    for time in range(len(series) - window_size):
        forecast.append(model.predict(series[time:time + window_size][np.
    ↪newaxis]))

    forecast = forecast[split_time-window_size:]
    results = np.array(forecast)[: , 0, 0]
    return results

# Save the forecast
dnn_forecast = generate_forecast()

# Plot it
plt.figure(figsize=(10, 6))
plot_series(time_valid, series_valid)
plot_series(time_valid, dnn_forecast)

```

2023-04-04 13:41:45.494960: I
tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:113]

Plugin optimizer for device_type GPU is enabled.

```
1/1 [=====] - 0s 272ms/step
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[illegible]

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```

[illegible]


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```

[illegible]

[illegible]

[illegible]

[illegible]

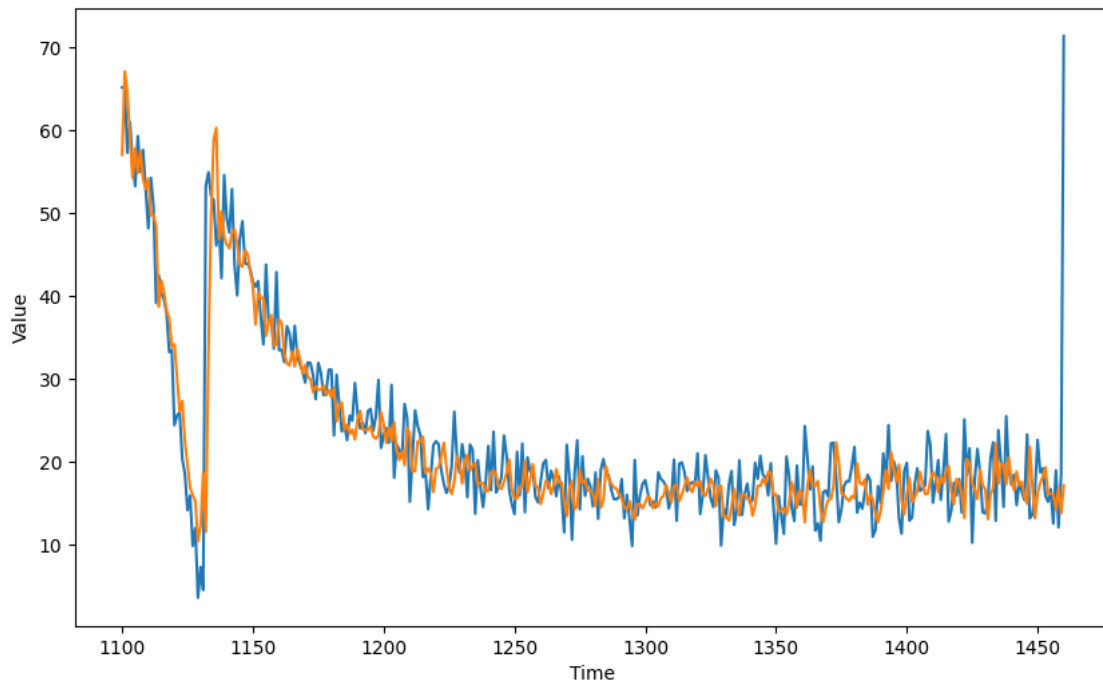
[illegible]

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[illegible]

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```



Expected Output:

A series similar to this one:

```
[11]: mse, mae = compute_metrics(series_valid, dnn_forecast)

print(f"mse: {mse:.2f}, mae: {mae:.2f} for forecast")
```

```
mse: 30.35, mae: 3.47 for forecast
```

To pass this assignment your forecast should achieve an MSE of 30 or less.

- If your forecast didn't achieve this threshold try re-training your model with a different architecture or tweaking the optimizer's parameters.
- If your forecast did achieve this threshold run the following cell to save your model in a HDF5 file which will be used for grading and after doing so, submit your assignment for grading.
- Make sure you didn't use `Lambda` layers in your model since these are incompatible with the HDF5 format which will be used to save your model for grading.
- This environment includes a dummy `my_model.h5` file which is just a dummy model trained for one epoch. **To replace this file with your actual model you need to run the next cell before submitting for grading.**

```
[12]: # Save your model in HDF5 format  
model.save('my_model.h5')
```

Congratulations on finishing this week's assignment!

You have successfully implemented a neural network capable of forecasting time series while also learning how to leverage Tensorflow's Dataset class to process time series data!

Keep it up!