Colab only includes TensorFlow 2.x; %tensorflow_version has no effect. 2.12.0

```
In [2]: from tensorflow.keras.layers import Input, SimpleRNN, GRU, LSTM, Dense, Flatten
from tensorflow.keras.models import Model
from tensorflow.keras.optimizers import SGD, Adam

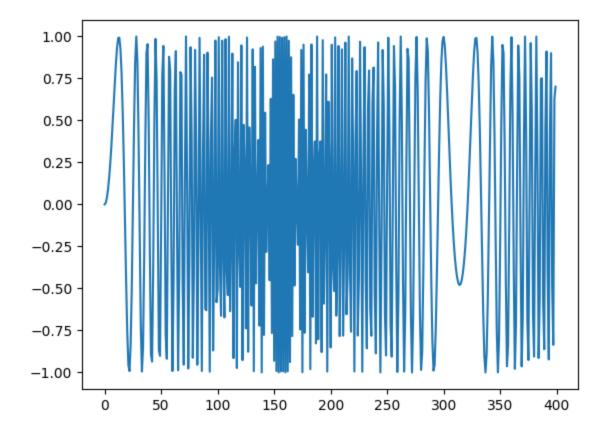
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
In [3]: # make the original data
series = np.sin((0.1*np.arange(400))**2)
```

This is a time series of the form:

$$x(t) = \sin(\omega t^2)$$

```
In [4]: # plot it
   plt.plot(series)
   plt.show()
```



```
In [5]: ### build the dataset
    # let's see if we can use T past values to predict the next value
    T = 10
    D = 1
    X = []
    Y = []
    for t in range(len(series) - T):
        x = series[t:t+T]
        X.append(x)
        y = series[t+T]
        Y.append(y)

X = np.array(X).reshape(-1, T) # make it N x T
Y = np.array(Y)
N = len(X)
print("X.shape", X.shape, "Y.shape", Y.shape)
```

X.shape (390, 10) Y.shape (390,)

```
In [6]: ### try autoregressive linear model
i = Input(shape=(T,))
x = Dense(1)(i)
model = Model(i, x)
model.compile(
    loss='mse',
    optimizer=Adam(learning_rate=0.01),
)

# train the RNN
r = model.fit(
    X[:-N//2], Y[:-N//2],
```

```
epochs=80,
validation_data=(X[-N//2:], Y[-N//2:]),
)
```

```
Epoch 1/80
801
Epoch 2/80
91
Epoch 3/80
07
Epoch 4/80
Epoch 5/80
Epoch 6/80
36
Epoch 7/80
23
Epoch 8/80
51
Epoch 9/80
88
Epoch 10/80
Epoch 11/80
21
Epoch 12/80
66
Epoch 13/80
7/7 [============= ] - 0s 8ms/step - loss: 0.5109 - val loss: 0.55
95
Epoch 14/80
80
Epoch 15/80
Epoch 16/80
Epoch 17/80
7/7 [============== ] - 0s 8ms/step - loss: 0.4913 - val loss: 0.56
37
Epoch 18/80
06
Epoch 19/80
7/7 [========================= ] - 0s 9ms/step - loss: 0.4861 - val loss: 0.57
```

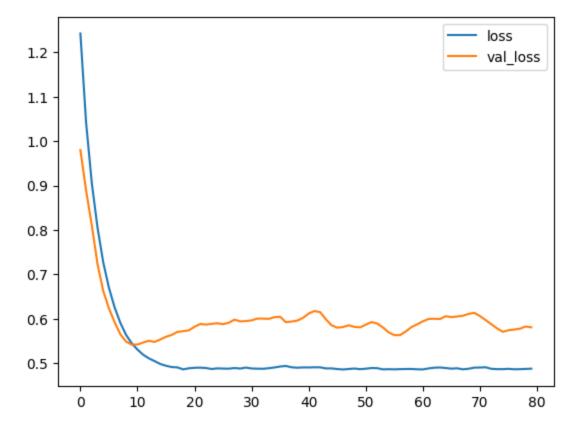
```
20
Epoch 20/80
7/7 [============= ] - 0s 10ms/step - loss: 0.4885 - val loss: 0.5
742
Epoch 21/80
Epoch 22/80
7/7 [============== ] - 0s 9ms/step - loss: 0.4900 - val loss: 0.58
Epoch 23/80
70
Epoch 24/80
7/7 [==========] - 0s 10ms/step - loss: 0.4869 - val loss: 0.5
887
Epoch 25/80
Epoch 26/80
80
Epoch 27/80
Epoch 28/80
7/7 [===========] - 0s 10ms/step - loss: 0.4891 - val_loss: 0.5
981
Epoch 29/80
43
Epoch 30/80
7/7 [===========] - 0s 16ms/step - loss: 0.4902 - val_loss: 0.5
950
Epoch 31/80
Epoch 32/80
Epoch 33/80
05
Epoch 34/80
99
Epoch 35/80
037
Epoch 36/80
47
Epoch 37/80
23
Epoch 38/80
```

```
Epoch 39/80
Epoch 40/80
21
Epoch 41/80
120
Epoch 42/80
176
Epoch 43/80
152
Epoch 44/80
7/7 [==========] - 0s 12ms/step - loss: 0.4883 - val loss: 0.5
Epoch 45/80
7/7 [===========] - 0s 14ms/step - loss: 0.4883 - val_loss: 0.5
854
Epoch 46/80
801
Epoch 47/80
7/7 [==========] - 0s 13ms/step - loss: 0.4859 - val loss: 0.5
814
Epoch 48/80
7/7 [===========] - 0s 15ms/step - loss: 0.4869 - val_loss: 0.5
855
Epoch 49/80
7/7 [==========] - 0s 19ms/step - loss: 0.4880 - val loss: 0.5
819
Epoch 50/80
7/7 [==========] - 0s 12ms/step - loss: 0.4866 - val_loss: 0.5
807
Epoch 51/80
7/7 [===========] - 0s 13ms/step - loss: 0.4876 - val_loss: 0.5
868
Epoch 52/80
7/7 [==========] - 0s 11ms/step - loss: 0.4892 - val_loss: 0.5
927
Epoch 53/80
7/7 [===========] - 0s 14ms/step - loss: 0.4889 - val_loss: 0.5
894
Epoch 54/80
806
Epoch 55/80
7/7 [===========] - 0s 12ms/step - loss: 0.4865 - val_loss: 0.5
697
Epoch 56/80
632
```

```
Epoch 57/80
633
Epoch 58/80
7/7 [==========] - 0s 11ms/step - loss: 0.4869 - val_loss: 0.5
715
Epoch 59/80
7/7 [===========] - 0s 13ms/step - loss: 0.4870 - val_loss: 0.5
814
Epoch 60/80
7/7 [==========] - 0s 12ms/step - loss: 0.4863 - val_loss: 0.5
874
Epoch 61/80
7/7 [==========] - 0s 12ms/step - loss: 0.4860 - val loss: 0.5
Epoch 62/80
7/7 [==========] - 0s 11ms/step - loss: 0.4882 - val_loss: 0.5
996
Epoch 63/80
002
Epoch 64/80
7/7 [===========] - 0s 19ms/step - loss: 0.4904 - val_loss: 0.5
995
Epoch 65/80
059
Epoch 66/80
Epoch 67/80
052
Epoch 68/80
069
Epoch 69/80
7/7 [============== ] - 0s 10ms/step - loss: 0.4875 - val loss: 0.6
109
Epoch 70/80
36
Epoch 71/80
Epoch 72/80
Epoch 73/80
7/7 [============= ] - 0s 10ms/step - loss: 0.4874 - val loss: 0.5
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Epoch 74/80
84
Epoch 75/80
7/7 [========================== ] - 0s 9ms/step - loss: 0.4866 - val loss: 0.57
```

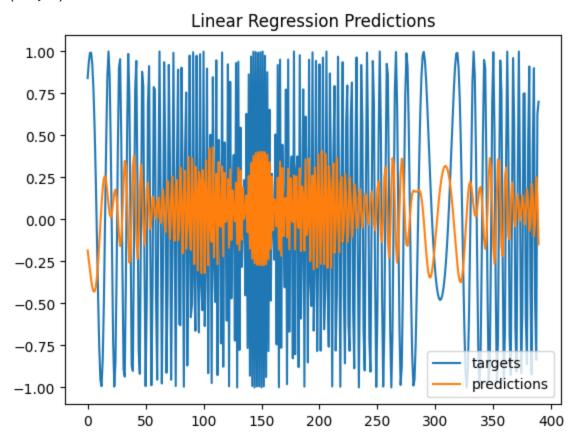
```
98
      Epoch 76/80
      41
      Epoch 77/80
      Epoch 78/80
      7/7 [=======
                    ========== ] - 0s 9ms/step - loss: 0.4866 - val loss: 0.57
      Epoch 79/80
                    =========] - 0s 10ms/step - loss: 0.4871 - val_loss: 0.5
      7/7 [=======
      827
      Epoch 80/80
      7/7 [=======
                    ========] - 0s 8ms/step - loss: 0.4877 - val_loss: 0.58
      99
In [7]: # Plot loss per iteration
      import matplotlib.pyplot as plt
      plt.plot(r.history['loss'], label='loss')
      plt.plot(r.history['val_loss'], label='val_loss')
      plt.legend()
```

Out[7]: <matplotlib.legend.Legend at 0x7fceb2dfdea0>



```
plt.plot(Y, label='targets')
plt.plot(predictions, label='predictions')
plt.title("Linear Regression Predictions")
plt.legend()
plt.show()
```

13/13 [=======] - 0s 2ms/step (390, 1)



```
In [9]: # This is the code we had before - it does the same thing
# One-step forecast using true targets

validation_target = Y[-N//2:]
validation_predictions = []

# index of first validation input
i = -N//2

while len(validation_predictions) < len(validation_target):
    p = model.predict(X[i].reshape(1, -1))[0,0] # 1x1 array -> scalar
    i += 1

# update the predictions list
validation_predictions.append(p)

plt.plot(validation_target, label='forecast target')
plt.plot(validation_predictions, label='forecast prediction')
plt.legend()
```

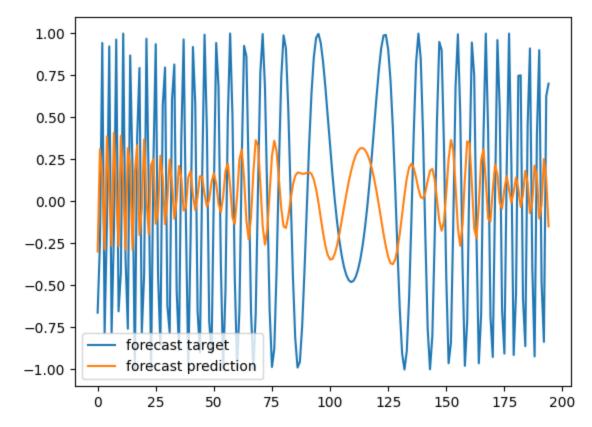
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1/1	[=======]	-	0s	23ms/step
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```

Out[9]: <matplotlib.legend.Legend at 0x7fce8a1e3b20>



```
In [10]: # Multi-step forecast
validation_target = Y[-N//2:]
```

```
validation_predictions = []

# first validation input
last_x = X[-N//2] # 1-D array of length T

while len(validation_predictions) < len(validation_target):
    p = model.predict(last_x.reshape(1, -1))[0,0] # 1x1 array -> scalar

# update the predictions list
    validation_predictions.append(p)

# make the new input
    last_x = np.roll(last_x, -1)
    last_x[-1] = p

plt.plot(validation_target, label='forecast target')
    plt.plot(validation_predictions, label='forecast prediction')
    plt.legend()
```

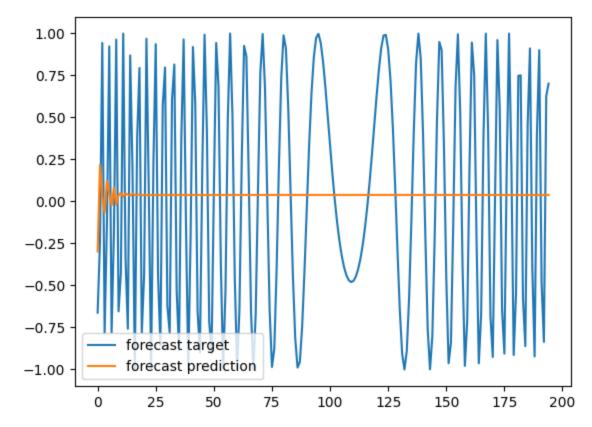
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1/1 [======] - 0s 18ms/step
1/1 [=======] - 0s 18ms/step
1/1 [=======] - 0s 17ms/step
```

Out[10]: <matplotlib.legend.Legend at 0x7fce8a0ee320>



In [11]: ### Now try RNN/LSTM model
X = X.reshape(-1, T, 1) # make it N x T x D

```
# make the RNN
i = Input(shape=(T, D))
x = LSTM(10)(i)
x = Dense(1)(x)
model = Model(i, x)
model.compile(
    loss='mse',
    optimizer=Adam(learning_rate=0.05),
)

# train the RNN
r = model.fit(
    X[:-N//2], Y[:-N//2],
    batch_size=32,
    epochs=200,
    validation_data=(X[-N//2:], Y[-N//2:]),
)
```

```
Epoch 1/200
294
Epoch 2/200
7/7 [==========] - 0s 10ms/step - loss: 0.4724 - val_loss: 0.5
237
Epoch 3/200
7/7 [==========] - 0s 12ms/step - loss: 0.4564 - val_loss: 0.5
494
Epoch 4/200
568
Epoch 5/200
712
Epoch 6/200
499
Epoch 7/200
334
Epoch 8/200
587
Epoch 9/200
250
Epoch 10/200
Epoch 11/200
020
Epoch 12/200
023
Epoch 13/200
7/7 [============== ] - 0s 10ms/step - loss: 0.0316 - val loss: 0.1
114
Epoch 14/200
033
Epoch 15/200
056
Epoch 16/200
Epoch 17/200
7/7 [============== ] - 0s 14ms/step - loss: 0.0110 - val loss: 0.1
114
Epoch 18/200
104
Epoch 19/200
7/7 [======================== ] - 0s 11ms/step - loss: 0.0068 - val loss: 0.1
```

```
182
Epoch 20/200
7/7 [============== ] - 0s 10ms/step - loss: 0.0067 - val loss: 0.1
089
Epoch 21/200
Epoch 22/200
7/7 [============== ] - 0s 11ms/step - loss: 0.0058 - val loss: 0.1
218
Epoch 23/200
514
Epoch 24/200
7/7 [=========================== ] - 0s 11ms/step - loss: 0.0156 - val loss: 0.1
356
Epoch 25/200
Epoch 26/200
153
Epoch 27/200
Epoch 28/200
965
Epoch 29/200
982
Epoch 30/200
977
Epoch 31/200
986
Epoch 32/200
Epoch 33/200
968
Epoch 34/200
081
Epoch 35/200
995
Epoch 36/200
973
Epoch 37/200
094
Epoch 38/200
```

```
061
Epoch 39/200
063
Epoch 40/200
117
Epoch 41/200
7/7 [==========] - 0s 10ms/step - loss: 0.0032 - val_loss: 0.0
932
Epoch 42/200
129
Epoch 43/200
109
Epoch 44/200
Epoch 45/200
179
Epoch 46/200
074
Epoch 47/200
151
Epoch 48/200
092
Epoch 49/200
021
Epoch 50/200
092
Epoch 51/200
136
Epoch 52/200
130
Epoch 53/200
135
Epoch 54/200
Epoch 55/200
081
Epoch 56/200
125
```

```
Epoch 57/200
124
Epoch 58/200
010
Epoch 59/200
005
Epoch 60/200
004
Epoch 61/200
Epoch 62/200
970
Epoch 63/200
003
Epoch 64/200
946
Epoch 65/200
949
Epoch 66/200
Epoch 67/200
079
Epoch 68/200
826
Epoch 69/200
7/7 [============== ] - 0s 21ms/step - loss: 0.0049 - val loss: 0.0
771
Epoch 70/200
023
Epoch 71/200
7/7 [============== ] - 0s 22ms/step - loss: 0.0069 - val loss: 0.0
793
Epoch 72/200
Epoch 73/200
7/7 [============== ] - 0s 21ms/step - loss: 0.0055 - val loss: 0.0
875
Epoch 74/200
043
Epoch 75/200
7/7 [======================== ] - 0s 19ms/step - loss: 0.0029 - val loss: 0.0
```

```
875
Epoch 76/200
7/7 [============== ] - 0s 22ms/step - loss: 0.0037 - val loss: 0.1
045
Epoch 77/200
Epoch 78/200
Epoch 79/200
943
Epoch 80/200
960
Epoch 81/200
946
Epoch 82/200
940
Epoch 83/200
Epoch 84/200
982
Epoch 85/200
042
Epoch 86/200
128
Epoch 87/200
041
Epoch 88/200
Epoch 89/200
7/7 [===========] - 0s 11ms/step - loss: 8.9212e-04 - val_loss:
0.1052
Epoch 90/200
023
Epoch 91/200
065
Epoch 92/200
7/7 [============= ] - 0s 11ms/step - loss: 9.7914e-04 - val_loss:
0.1097
Epoch 93/200
126
Epoch 94/200
```

```
920
Epoch 95/200
Epoch 96/200
146
Epoch 97/200
7/7 [==========] - 0s 11ms/step - loss: 0.0021 - val_loss: 0.1
045
Epoch 98/200
138
Epoch 99/200
7/7 [============= ] - 0s 11ms/step - loss: 7.7199e-04 - val_loss:
0.1061
Epoch 100/200
Epoch 101/200
7/7 [===========] - 0s 11ms/step - loss: 9.3456e-04 - val_loss:
0.1126
Epoch 102/200
96
Epoch 103/200
119
Epoch 104/200
014
Epoch 105/200
Epoch 106/200
019
Epoch 107/200
183
Epoch 108/200
251
Epoch 109/200
149
Epoch 110/200
Epoch 111/200
7/7 [==========] - 0s 12ms/step - loss: 8.7829e-04 - val_loss:
0.1201
Epoch 112/200
109
```

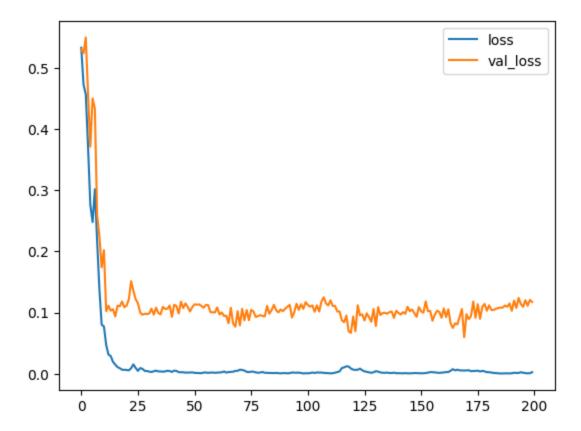
```
Epoch 113/200
109
Epoch 114/200
020
Epoch 115/200
020
Epoch 116/200
875
Epoch 117/200
844
Epoch 118/200
953
Epoch 119/200
695
Epoch 120/200
670
Epoch 121/200
937
Epoch 122/200
Epoch 123/200
121
Epoch 124/200
955
Epoch 125/200
7/7 [============== ] - 0s 10ms/step - loss: 0.0059 - val loss: 0.0
973
Epoch 126/200
871
Epoch 127/200
7/7 [============= ] - 0s 10ms/step - loss: 0.0033 - val loss: 0.0
Epoch 128/200
926
Epoch 129/200
7/7 [============== ] - 0s 11ms/step - loss: 0.0018 - val loss: 0.0
850
Epoch 130/200
061
Epoch 131/200
7/7 [======================== ] - 0s 12ms/step - loss: 0.0044 - val loss: 0.0
```

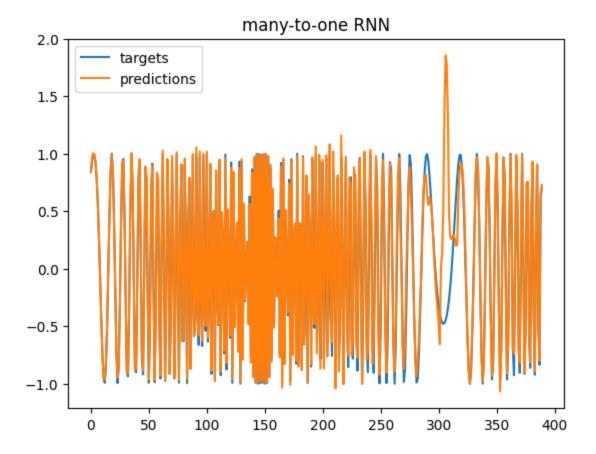
```
781
Epoch 132/200
7/7 [============== ] - 0s 10ms/step - loss: 0.0037 - val loss: 0.1
091
Epoch 133/200
Epoch 134/200
Epoch 135/200
997
Epoch 136/200
983
Epoch 137/200
Epoch 138/200
019
Epoch 139/200
Epoch 140/200
031
Epoch 141/200
995
Epoch 142/200
7/7 [===========] - 0s 12ms/step - loss: 9.6197e-04 - val_loss:
0.0970
Epoch 143/200
7/7 [==========] - 0s 11ms/step - loss: 8.8874e-04 - val_loss:
0.1009
Epoch 144/200
Epoch 145/200
7/7 [===========] - 0s 11ms/step - loss: 8.4202e-04 - val_loss:
0.1097
Epoch 146/200
7/7 [==========] - 0s 12ms/step - loss: 9.1971e-04 - val_loss:
0.1024
Epoch 147/200
051
Epoch 148/200
993
Epoch 149/200
931
Epoch 150/200
```

```
091
Epoch 151/200
7/7 [==========] - 0s 10ms/step - loss: 9.7763e-04 - val_loss:
0.1021
Epoch 152/200
999
Epoch 153/200
7/7 [==========] - 0s 11ms/step - loss: 0.0014 - val_loss: 0.1
184
Epoch 154/200
023
Epoch 155/200
029
Epoch 156/200
Epoch 157/200
963
Epoch 158/200
032
Epoch 159/200
013
Epoch 160/200
917
Epoch 161/200
Epoch 162/200
928
Epoch 163/200
055
Epoch 164/200
822
Epoch 165/200
751
Epoch 166/200
Epoch 167/200
808
Epoch 168/200
932
```

```
Epoch 169/200
060
Epoch 170/200
599
Epoch 171/200
974
Epoch 172/200
893
Epoch 173/200
Epoch 174/200
182
Epoch 175/200
916
Epoch 176/200
139
Epoch 177/200
899
Epoch 178/200
Epoch 179/200
142
Epoch 180/200
030
Epoch 181/200
7/7 [============== ] - 0s 11ms/step - loss: 0.0028 - val loss: 0.1
119
Epoch 182/200
046
Epoch 183/200
043
Epoch 184/200
Epoch 185/200
7/7 [============= ] - 0s 10ms/step - loss: 8.7073e-04 - val loss:
0.1075
Epoch 186/200
7/7 [==========] - 0s 12ms/step - loss: 7.5564e-04 - val_loss:
0.1084
Epoch 187/200
```

```
0.1081
    Epoch 188/200
    7/7 [===========] - 0s 13ms/step - loss: 8.9494e-04 - val loss:
    0.1117
    Epoch 189/200
    7/7 [==========] - 0s 10ms/step - loss: 8.1483e-04 - val_loss:
    Epoch 190/200
    7/7 [==========] - 0s 11ms/step - loss: 7.9022e-04 - val loss:
    0.1151
    Epoch 191/200
    030
    Epoch 192/200
    197
    Epoch 193/200
    Epoch 194/200
    240
    Epoch 195/200
    Epoch 196/200
    094
    Epoch 197/200
    203
    Epoch 198/200
    7/7 [==========] - 0s 10ms/step - loss: 9.0915e-04 - val_loss:
    0.1112
    Epoch 199/200
    208
    Epoch 200/200
    174
In [12]: # plot some data
    plt.plot(r.history['loss'], label='loss')
    plt.plot(r.history['val_loss'], label='val_loss')
    plt.legend()
    plt.show()
```





```
In [14]: # Multi-step forecast
forecast = []
input_ = X[-N//2]
while len(forecast) < len(Y[-N//2:]):
    # Reshape the input_ to N x T x D
    f = model.predict(input_.reshape(1, T, 1))[0,0]
    forecast.append(f)

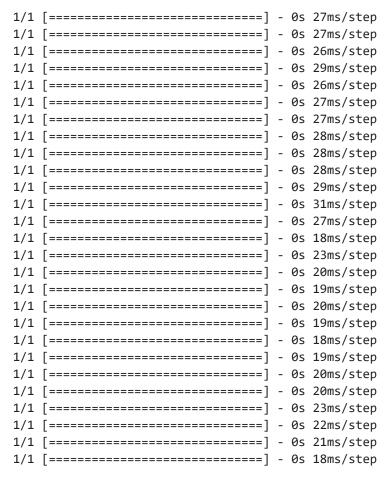
# make a new input with the latest forecast
    input_ = np.roll(input_, -1)
    input_[-1] = f

plt.plot(Y[-N//2:], label='targets')
plt.plot(forecast, label='forecast')
plt.title("RNN Forecast")
plt.legend()
plt.show()</pre>
```

1/1	[======]	-	0s	27ms/step
1/1	[======]	-	0s	19ms/step
1/1	[======]	-	0s	20ms/step
1/1	[======]	-	0s	19ms/step
1/1	[======]	-	0s	25ms/step
1/1	[=======]	-	0s	20ms/step
1/1	[=======]	-	0s	19ms/step
1/1	[=======]	-	0s	21ms/step
1/1	[=======]	-	0s	19ms/step
1/1	[=======]	-	0s	21ms/step
1/1	[=======]	-	0s	19ms/step
1/1	[=======]	_	0s	18ms/step
1/1	[=======]	_		20ms/step
1/1	[=======]			18ms/step
1/1	-			18ms/step
1/1				20ms/step
1/1	[======]			19ms/step
1/1	[=======]		0s	20ms/step
1/1	[=======]		0s	18ms/step
1/1	[=======]			20ms/step
1/1		_		18ms/step
1/1			0s	20ms/step
1/1	[=======]			18ms/step
1/1				•
٠.			0s	19ms/step
1/1				20ms/step
1/1			0s	19ms/step
1/1			0s	19ms/step
1/1		-		20ms/step
1/1	[=======]			19ms/step
1/1	[======]			21ms/step
1/1				23ms/step
1/1				20ms/step
1/1				21ms/step
1/1	[======]			18ms/step
	[======]			•
1/1				
1/1				•
1/1	-			
1/1	_			
	[======]			
1/1				
1/1	-			•
1/1	_			
1/1	-			
1/1				•
1/1	[======]	-	0s	•
1/1	[======]	-	0s	18ms/step
1/1	-			19ms/step
1/1	-			•
1/1	[]			•
1/1	[]			•
1/1	-			
1/1	_			
1/1	_			
1/1	_			
1/1	[]	-	0s	20ms/step

1/	1 [========]	-	0s	18ms/step
1/	1 [=========]	-	0s	22ms/step
1/	1 [=========]	-	0s	21ms/step
1/	1 [=========]	-	0s	19ms/step
1/	1 [=========]	-	0s	18ms/step
1/	1 [=========]	-	0s	22ms/step
1/	1 [========]	-	0s	21ms/step
1/	1 [=========]	-	0s	19ms/step
1/	1 [========]	-	0s	18ms/step
1/	1 [========]	-	0s	20ms/step
1/	1 [========]	-	0s	20ms/step
1/	1 [========]	-	0s	19ms/step
1/	1 [========]	-	0s	21ms/step
1/	1 [========]	-	0s	21ms/step
1/	1 [========]	-	0s	20ms/step
1/	1 [========]	-	0s	18ms/step
1/	1 [========]	-	0s	27ms/step
1/	1 [========]	-	0s	18ms/step
1/	1 [========]	-	0s	18ms/step
1/	1 [========]	-	0s	19ms/step
1/	1 [========]	-	0s	19ms/step
1/	1 [========]	-	0s	24ms/step
1/	1 [========]	-	0s	20ms/step
1/	1 [========]	-	0s	25ms/step
1/	1 [========]	-	0s	21ms/step
1/	1 [========]	-	0s	22ms/step
1/	1 [========]	-	0s	20ms/step
1/	1 [========]	-	0s	18ms/step
	1 [=========]			•
	1 [=========]			•
1/	1 [=========]	-	0s	22ms/step
1/	1 [=========]	-	0s	20ms/step
1/	1 [=========]	-	0s	19ms/step
1/	1 [========]	-	0s	20ms/step
1/	1 [========]	-	0s	20ms/step
	1 [========]			
	1 [========]			•
	1 [========]			•
	1 [========]			
	1 [========]			•
	1 [=======]			•
	1 [=======]			•
	1 [======]			
	1 [======]			
	1 [=======]			
	1 [=======]			
	1 [========]			
	1 [========]			
	1 [==========]			•
	1 [=========]			
	1 [==========]			
	1 [==========]			•
	1 [=========]			•
	1 [=========]			
	1 [=======]			
1/	1 [=======]	-	0s	20ms/step

1/1	. [======]	-	0s	18ms/step
1/1	. [=======]	-	0s	18ms/step
1/1	. [======]	-	0s	18ms/step
1/1	. [======]	-	0s	21ms/step
1/1	. [=======]	-	0s	21ms/step
1/1	. [======]	-	0s	19ms/step
1/1	. [=======]	-	0s	25ms/step
1/1	. [=======]	-	0s	18ms/step
1/1	. [=======]	-	0s	22ms/step
1/1	. [=======]	-	0s	18ms/step
1/1	. [=======]	-	0s	23ms/step
1/1	. [=======]	-	0s	19ms/step
1/1	. [=======]	-	0s	26ms/step
1/1	. [=======]	-	0s	20ms/step
1/1	. [=======]	-	0s	19ms/step
1/1	. [=======]	-	0s	20ms/step
1/1	. [=======]	-	0s	19ms/step
1/1	. [=======]	-	0s	25ms/step
1/1	. [=======]	-	0s	18ms/step
1/1	. [=======]	-	0s	19ms/step
1/1	. [=======]	-	0s	22ms/step
1/1	. [=======]	-	0s	22ms/step
1/1	. [=======]	-	0s	18ms/step
1/1	. [=======]	-	0s	18ms/step
1/1	. []	-	0s	18ms/step
1/1	. [=======]	-	0s	19ms/step
1/1				18ms/step
1/1				20ms/step
1/1				20ms/step
1/1				17ms/step
1/1	. [======]	-	0s	19ms/step
1/1	. []	-	0s	20ms/step
1/1	. []	-	0s	18ms/step
1/1	. []	-	0s	22ms/step
	. [======]			
	. [======]			•
	. [======]			
	. [======]			
	. [======]			
	. [======]			
	. [======]			
	. [======]			
1/1	-			
1/1				
1/1	-			
	. [=======]			
	. [========]			
	. [========]			
1/1				
	. [========]			
	. [=======]			
1/1	2			
	. [=======]			
	. [========]			
	. [========]			
1/1	. [=======]	-	0 S	28ms/step



RNN Forecast

