

Neural Network Laboratory Work – 6

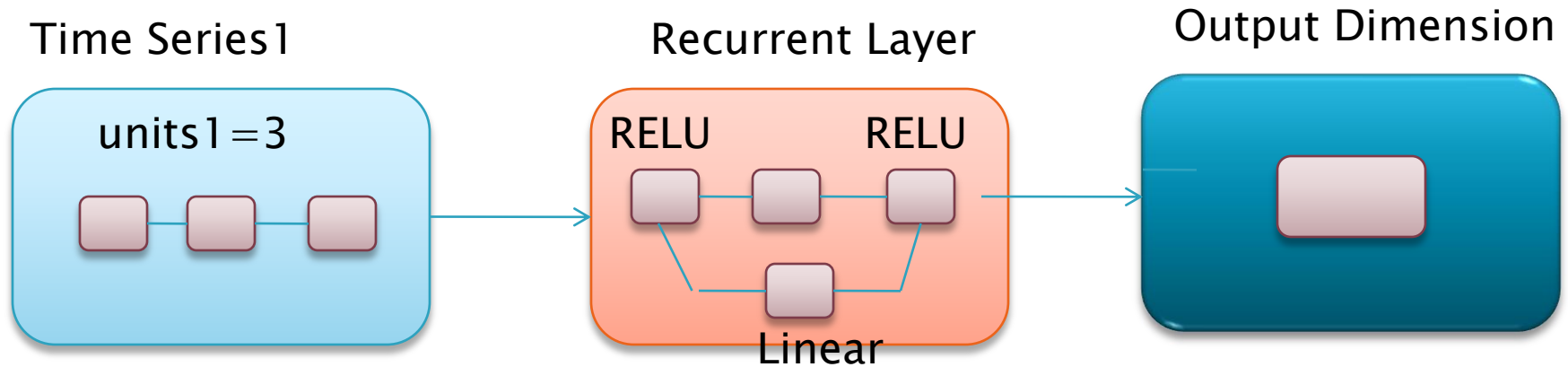
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RNN USAGE

- ▶ **Recurrent neural networks** (RNN) are a class of **neural networks** that are helpful in modeling sequence data.
- ▶ Derived from feedforward **networks**, RNNs exhibit similar behavior to how human brains function.
- ▶ **Recurrent neural networks** produce predictive results in sequential data that other algorithms can't.

RNN Layers

- ▶ RNN: recurrent connections, information is kept in the hidden states.
- ▶ RNN: neurons in the hidden layer have recurrent layers .
- ▶ RNN inputs are values of time series x_1, x_2, x_3 at time moments t and at the previous time moments $t-1, t-2, \dots, t-d$.



RNN – Parameters

- ▶ `model.add(SimpleRNN(units=100, input_shape=in_dim, activation="relu"))`
- ▶ `#units`: Positive integer, dimensionality of the output space
- ▶ `model.add(Dense(units=16, activation="relu"))`
- ▶ `#adding feed forward layer, 16 neurons in a hidden layer`
- ▶ `model.add(Dense(out_dim, activation='linear'))`

RNN and Dense Layer – Parameters, Output and Shape

Layer (type)	Output Shape	Param #
=====	=====	=====
simple_rnn_3 (SimpleRNN)	(None, 100)	10400
dense_5 (Dense)	(None, 16)	1616
dense_6 (Dense)	(None, 2)	34
=====	=====	=====
Total params: 12,050		
Trainable params: 12,050		
Non-trainable params: 0		

Loss per Epochs – 50 (Epochs)

```
Epoch 1/50
797/797 [=====] - 1s 1ms/step - loss: 5.5789
Epoch 2/50
797/797 [=====] - 0s 181us/step - loss: 0.7100
Epoch 3/50
797/797 [=====] - 0s 180us/step - loss: 0.5907
Epoch 4/50
797/797 [=====] - 0s 181us/step - loss: 0.5611
Epoch 5/50
797/797 [=====] - 0s 195us/step - loss: 0.5650
Epoch 6/50
797/797 [=====] - 0s 173us/step - loss: 0.6004
Epoch 7/50
797/797 [=====] - 0s 185us/step - loss: 0.5968
Epoch 8/50
797/797 [=====] - 0s 175us/step - loss: 0.5496
Epoch 9/50
797/797 [=====] - 0s 185us/step - loss: 0.5509
Epoch 10/50
```

Loss per Epochs – 50 (Epochs)

```
Epoch 41/50  
797/797 [=====] - 0s 175us/step - loss: 0.5534  
Epoch 42/50  
797/797 [=====] - 0s 173us/step - loss: 0.5269  
Epoch 43/50  
797/797 [=====] - 0s 183us/step - loss: 0.5523  
Epoch 44/50  
797/797 [=====] - 0s 165us/step - loss: 0.5396  
Epoch 45/50  
797/797 [=====] - 0s 189us/step - loss: 0.5494  
Epoch 46/50  
797/797 [=====] - 0s 173us/step - loss: 0.5368  
Epoch 47/50  
797/797 [=====] - 0s 168us/step - loss: 0.5735  
Epoch 48/50  
797/797 [=====] - 0s 173us/step - loss: 0.5368  
Epoch 49/50  
797/797 [=====] - 0s 161us/step - loss: 0.5237  
Epoch 50/50  
797/797 [=====] - 0s 164us/step - loss: 0.5550
```

MSE, RMSE and Correlation – Training and Testing Data

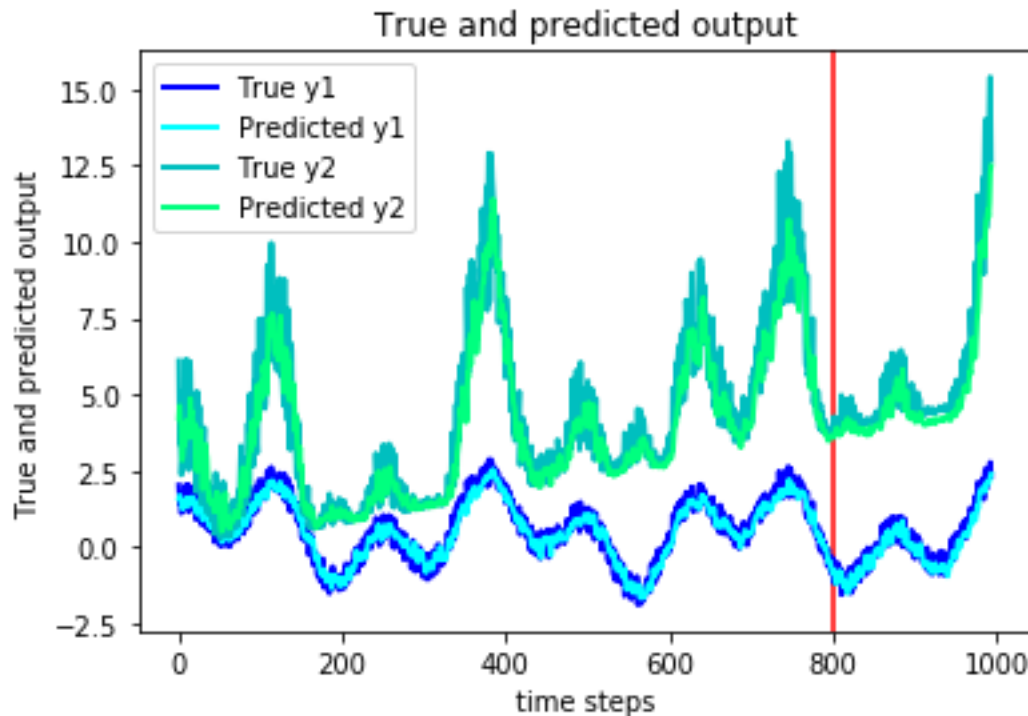
Train time series

	y1	y2
MSE	0.12	1.02
RMSE	0.34	1.01
Corr coef	0.95	0.94
Corr coef p	0.00	0.00

Test time series

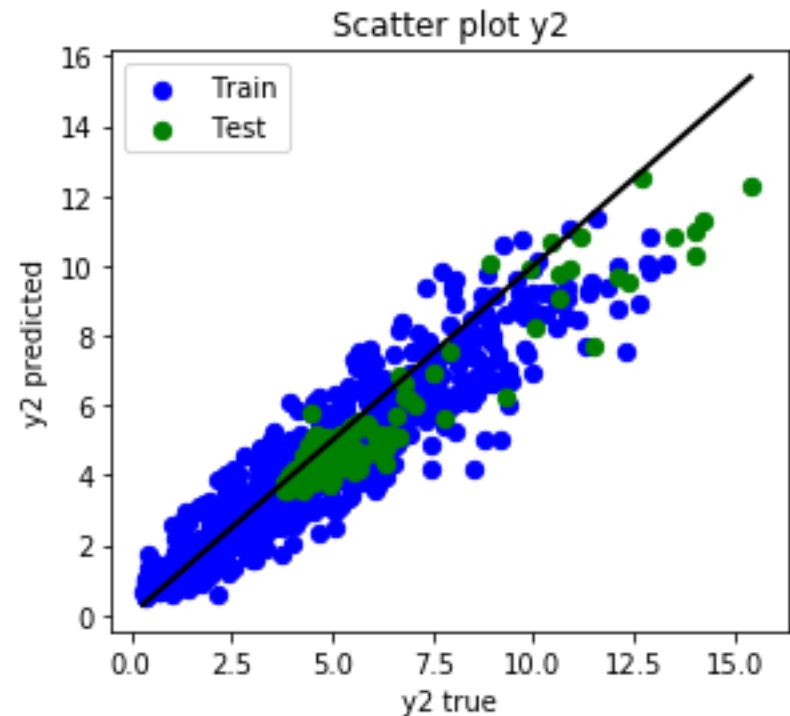
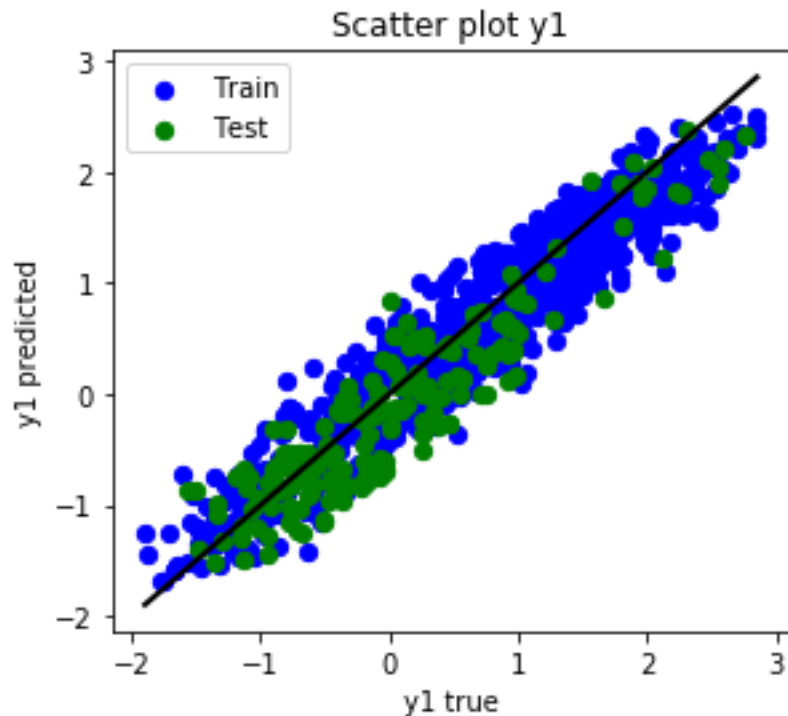
	y1	y2
MSE	0.14	0.84
RMSE	0.38	0.92
Corr coef	0.93	0.96
Corr coef p	0.00	0.00

outputs and predictions as a function of time



Blue – Plot of Y1
Aqua – Predicted y1
Green – Predicted y2

Outputs and Prediction



Input and Output Dimension, Unit Steps

- ▶ $N=2000$ #number of time steps
- ▶ $n=400$ #number of time steps for training
- ▶ $\text{step} = 5$

EPOCHS Loss

```
395/395 [=====] - 0s 202us/step - loss: 0.4789
Epoch 42/50
395/395 [=====] - 0s 222us/step - loss: 0.4603
Epoch 43/50
395/395 [=====] - 0s 205us/step - loss: 0.4038
Epoch 44/50
395/395 [=====] - 0s 220us/step - loss: 0.4076
Epoch 45/50
395/395 [=====] - 0s 210us/step - loss: 0.4579
Epoch 46/50
395/395 [=====] - 0s 217us/step - loss: 0.3920
Epoch 47/50
395/395 [=====] - 0s 187us/step - loss: 0.4113
Epoch 48/50
395/395 [=====] - 0s 187us/step - loss: 0.3842
Epoch 49/50
395/395 [=====] - 0s 197us/step - loss: 0.4123
Epoch 50/50
395/395 [=====] - 0s 202us/step - loss: 0.3937
```

Train and Test Series – Coefficient

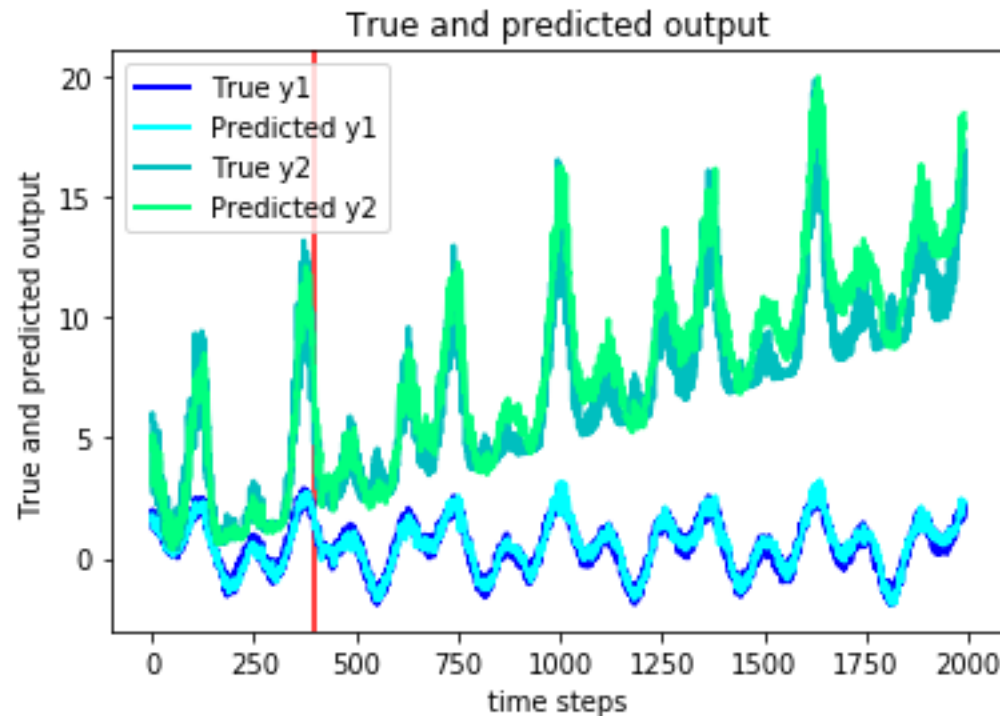
Train time series

	y1	y2
MSE	0.08	0.60
RMSE	0.28	0.77
Corr coef	0.97	0.97
Corr coef p	0.00	0.00

Test time series

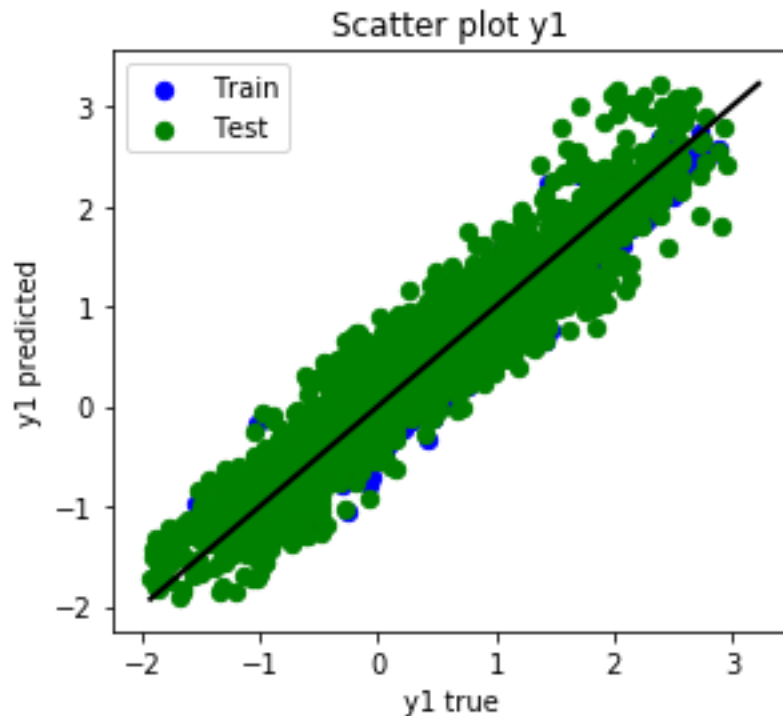
	y1	y2
MSE	0.14	2.44
RMSE	0.38	1.56
Corr coef	0.93	0.95
Corr coef p	0.00	0.00

outputs and predictions as a function of time



Blue – Plot of Y1
Aqua – Predicted y1
Green – Predicted y2

Outputs and Prediction



Thank You