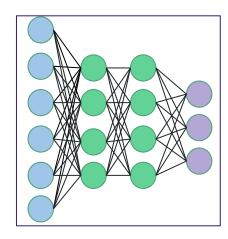
Modelling a time series using a feedforward neural network

Tom Monks
University of Exeter

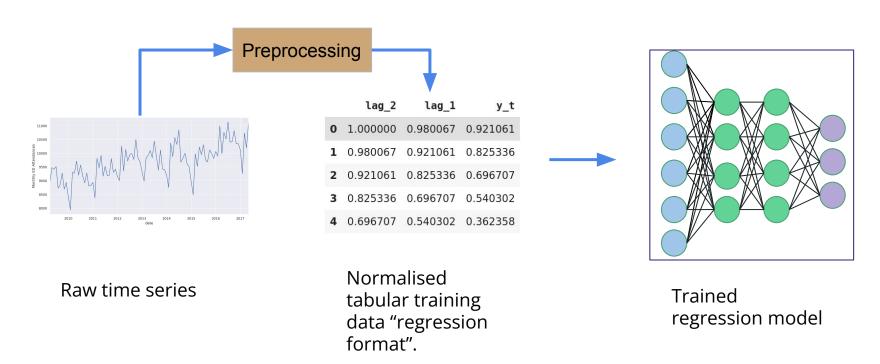
Overview

By the end of the lecture you will have gained knowledge in:

- Preprocessing time series ready for neural networks
- Methods for forecasting with feedforward neural networks
 - Iterative method
 - Director method
 - Vector output method
- How to reduce overfitting using ensemble learning



Feedforward neural networks for time series



Autoregressive model

Raw time series

10

15

12

18

21

11

16

Processed time series

lag_2	lag_1	Y

Let's build a training dataset for a lag 2 AR model.

Autoregressive model

Raw time series

lag_2	lag_1	Υ
10	15	12

Sliding window.

Autoregressive model

Raw time series

lag_2	lag_1	Υ
10	15	12
15	12	18

Autoregressive model

Raw time series

lag_2	lag_1	Υ
10	15	12
15	12	18
12	18	21

Autoregressive model

Raw time series

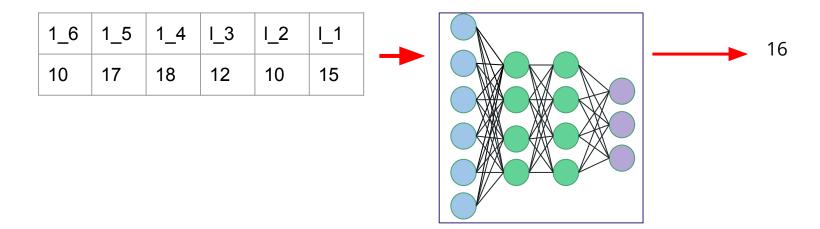
lag_2	lag_1	Υ
10	15	12
15	12	18
12	18	21
18	21	11

Autoregressive model

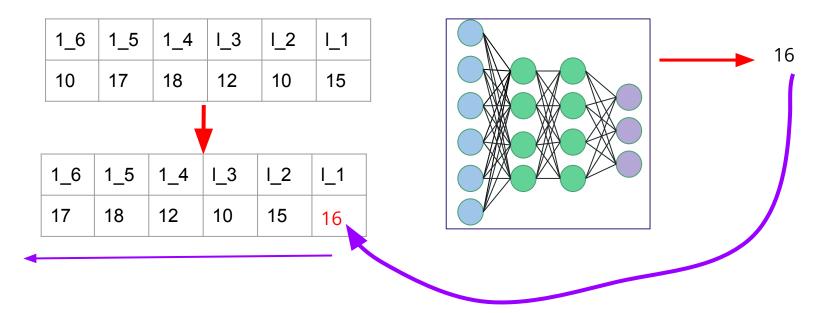
Raw time series

lag_2	lag_1	Υ
10	15	12
15	12	18
12	18	21
18	21	11
21	11	16

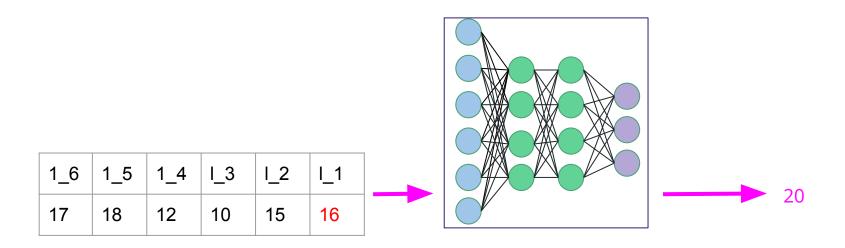
Iterative Forecasting



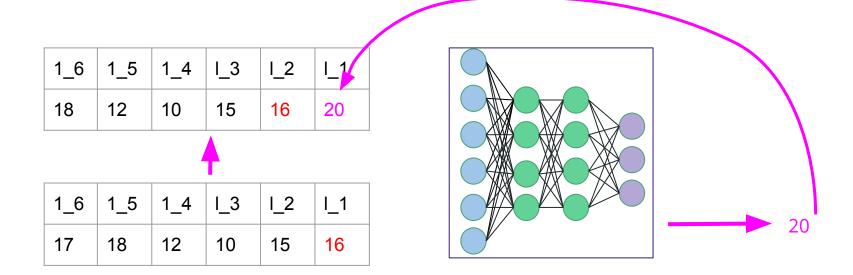
Multi-step forecasting: iterative approach



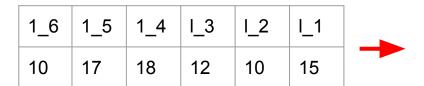
Multi-step forecasting: iterative approach



Multi-step forecasting: iterative approach

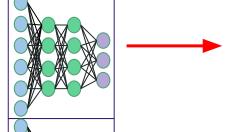


Multi-step forecasting: direct method



1_6	1_5	1_4	I_3	I_2	I_1	
10	17	18	12	10	15	

1_6	1_5	1_4	I_3	l_2	l_1
10	17	18	12	10	15



t+1	t+2	t+3
16		

	t-
	

t+1	t+2	t+3
	20	

t+1	t+2	t+3
		18

Preprocessing for direct method

Autoregressive model

Raw time series

lag_2	lag_1	Υ
10	15	18

For a model where h = 2

Preprocessing for direct method

Autoregressive model

Raw time series

lag_2	lag_1	Υ
10	15	21

For a model where h = 3

Forecasting a vector

1_6	1_5	1_4	I_3	I_2	l_1		t+1	t+2	t+3
10	17	18	12	10	15		16	20	18

Preprocessing for vector method

Autoregressive model

Raw time series

lag_2	lag_1	Y{t+1}	Y{t+2}	Y{t+3}
10	15	12	18	21

For a model where h = 3