

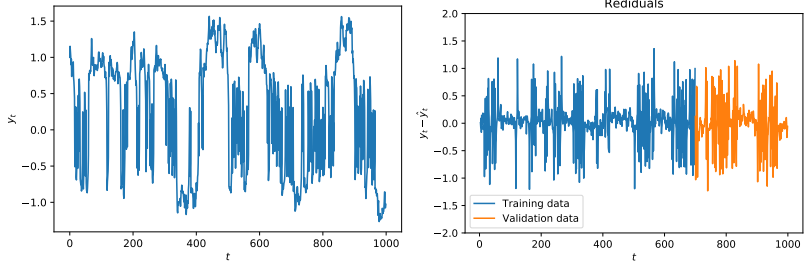
Time Series and Sequence Learning

Discussion seminar for Lecture 3

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lecture3a – Detecting nonlinearity

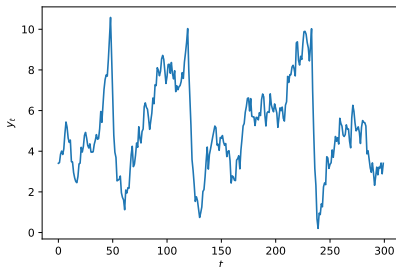


Left: Data. Right: Residuals for a fitted AR(3) model.

Discussion questions:

1. How can we see from the **residuals** that the data is not well explained by the model?
2. Is it possible to see that that data is not well explained by a *linear* process by **just looking at the data**? How?

lecture3a – Detecting nonlinearity



Discussion questions:

1. Does the data above **appear** to be generated by a linear $AR(p)$ model?

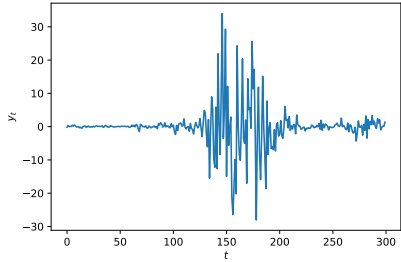
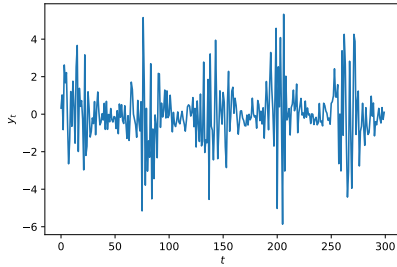
It is not a stationary AR, and it does not have the same pattern along the time

2. If $\{y_t\}_{t \geq 1}$ is a **stationary** process generated by an $AR(p)$ model with coefficients (a_1, \dots, a_p) , then what can be said about the **negated process** $\{-y_t\}_{t \geq 1}$

Yes, it is a mirror, it also means it change the ar's sign, so it is still linear stationary

AR model can have pattern, but only a same pattern and constantly pattern, not wide and narrow wave at the same time

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Left: Non-stationary (cyclic variance).

Right: Stationary but nonlinear (stochastic variance).

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When we are asked the question “*Is this time series stationary?*”, the meaning is often “*Could this time series have been generated by a stationary linear process?*”

In practice it is hard to distinguish between non-stationarity and nonlinearity!

Nonlinear auto-regressive model, NAR(p):

$$y_t = f_{\theta}(y_{t-1}, y_{t-2}, \dots, y_{t-p}) + \varepsilon_t, \quad \varepsilon_t \sim \mathcal{N}(0, \sigma_{\varepsilon}^2),$$

Using an 2-layer NN to model f_{θ} :

$$\mathbf{h}_t = \sigma(W^{(1)} H_{t-1} + b^{(1)})$$

$$y_t = W^{(2)} \mathbf{h}_t + b^{(2)} + \varepsilon_t.$$

Discussion questions:

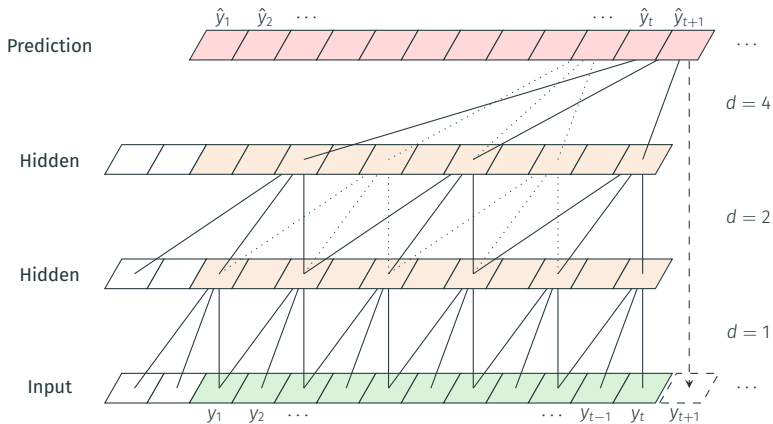
1. Contrary to a linear AR(p) model it is customary to include “bias terms” $b^{(j)}$ in the MLP-NAR. Why? Why don't we include a bias/intercept in the linear AR model?
in ar model, we will sum y_t and it will sum to inf
2. How is the squared loss function
activate function will help to not over the limit

$$L(\theta) = \frac{1}{n-p} \sum_{t=p+1}^n (y_t - f_{\theta}(y_{t-1}, y_{t-2}, \dots, y_{t-p}))^2$$

related to the **data likelihood**?

we can not predict the first p value, we just skip first p values
in NN network in the next slide using d , several values will not be used or predicted.

lecture3{b,c} - TCN



Discussion questions:

1. Which statements are correct?
 - a) NAR is a special case of TCN.
 - b) TCN is a special case of NAR.
 - c) TCN and NAR are equivalent model classes.
2. How is the receptive field of a TCN related to the order p of a $\text{NAR}(p)$ model? P is same as receptive value
3. Will the receptive field always increase exponentially with depth for a TCN with dilated convolutions?