

Time Series and Sequence Learning

Validation, Order selection

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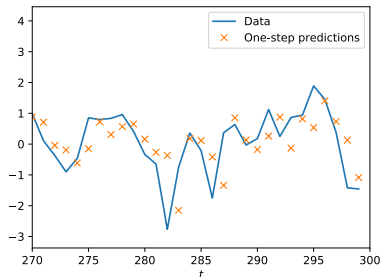
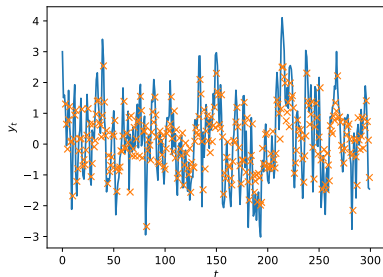
ex) Toy model

We simulate an AR(3) model for $n = 300$ time steps,

$$y_t = 0.9y_{t-1} - 0.4y_{t-2} + 0.2y_{t-3} + \varepsilon_t, \quad \varepsilon_t \sim \mathcal{N}(0, 1)$$

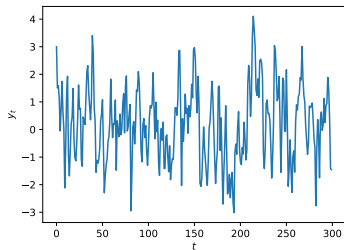
Estimating the model parameters with OLS gives:

$$\hat{\theta} = (0.84, -0.33, 0.16) \text{ and } \hat{\sigma}_\varepsilon^2 = 0.95.$$



Order selection

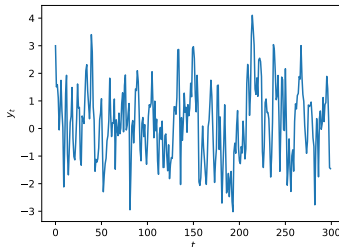
In practice we only observe the data.



How do we know which model order p to pick?!

Order selection

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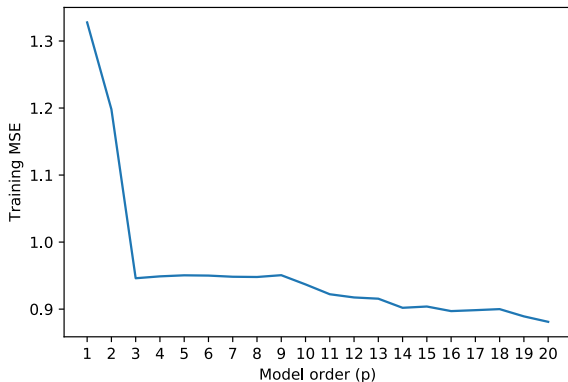
How do we know which model order p to pick?!

Two approaches:

1. Try to figure it out *before* fitting the model (“exploratory data analysis”)
2. Estimate **multiple models of different orders** and perform model selection by **validation**!

ex) Toy model, cont'd

1. Look for the “bend” in training error plot



plot the MSE of training data 1st

Residual analysis

2. Look at the residuals!

The **model assumption** is

$$y_t = \theta^\top \phi_t + \varepsilon_t,$$

$$\varepsilon_t \stackrel{\text{iid}}{\sim} \mathcal{N}(0, \sigma_\varepsilon^2).$$

Residual analysis

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The **model assumption** is

$$y_t = \theta^\top \phi_t + \varepsilon_t, \quad \varepsilon_t \stackrel{\text{iid}}{\sim} \mathcal{N}(0, \sigma_\varepsilon^2).$$

Hence, if the model is accurate, we expect

$$y_t - \hat{\theta}^\top \phi_t \approx \varepsilon_t.$$

The residuals should be **white Gaussian noise**!

1. Auto-correlation
2. QQ-plots for marginal Gaussianity
3. ...

ex) Toy model, cont'd

$X_t = \phi * X_{t-1} + \epsilon_t$

$\phi(h) = \Phi^{h(h)}$ at lag h

$\hat{\phi}(h) = \frac{\sum_{t=1}^{n-h} (x_t - \bar{x})(x_{t+h} - \bar{x})}{\sum_{t=1}^n (x_t - \bar{x})^2}$

When N is big, $\hat{\phi}(h)$ is approximately normal

so $\hat{\phi}(h) \sim N(\phi(h), 1/n)$ $\phi(h)$ is true autocorr

and var of $\hat{\phi}(h)$ is appr $1/n$, because

$\text{var}(\hat{\phi}(h)) \sim 1/n$, so $\text{sqr}(\text{var}) = 1/\text{sqr}(n)$

Reason:

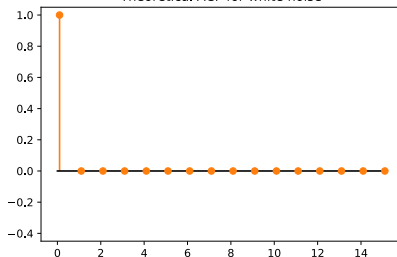
Asymptotic Normality of Sample Autocorrelations

样本自相关的渐近正态性

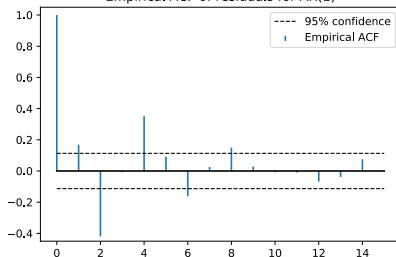
$$\hat{\phi}(h) \sim N(0, \frac{1}{\sqrt{n}}) \quad h > 0, \quad n \text{ large}$$

Estimated model: AR(1)

Theoretical ACF for white noise

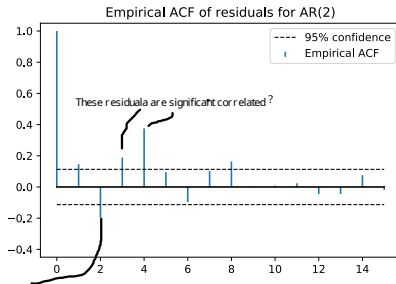
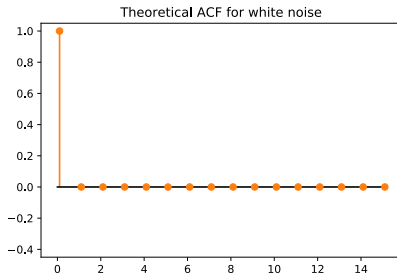


Empirical ACF of residuals for AR(1)



ex) Toy model, cont'd

Estimated model: AR(2)

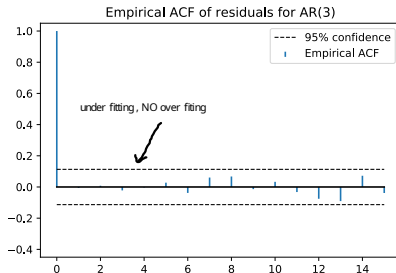
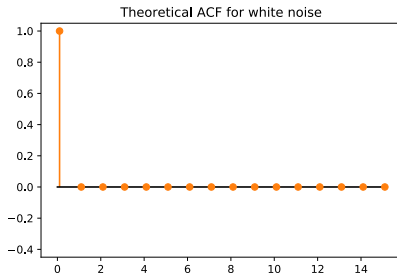


This could indicate that there is some degree of autocorrelation at those lags which is not explained by the AR(1) model.

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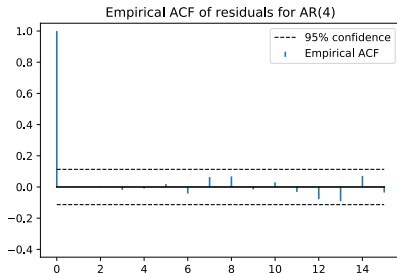
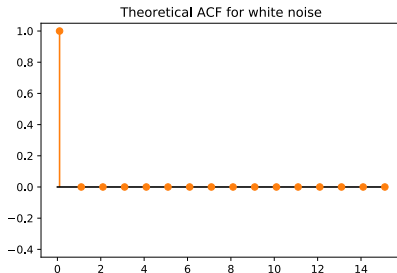
ex) Toy model, cont'd

Estimated model: AR(3)



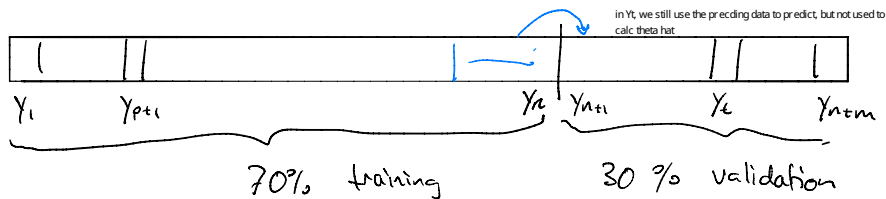
ex) Toy model, cont'd

Estimated model: AR(4)



Prediction error validation

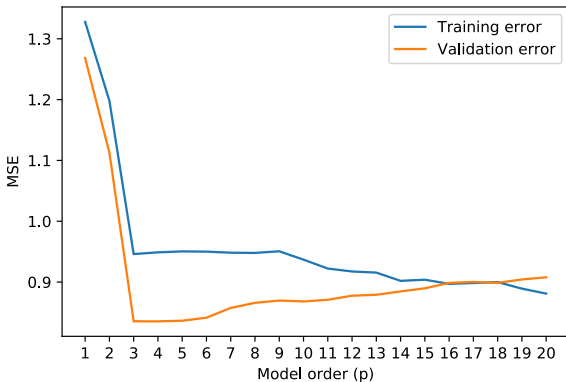
3. Evaluate on held-out validation data!



Validation mean-squared error, using one-step-ahead predictions:

$$\text{Val-MSE}(\hat{\theta}) = \frac{1}{m} \sum_{t=n+1}^{n+m} (y_t - \hat{\theta}^T \phi_t)^2$$

ex) Toy model, cont'd



Testing the model

