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9. Fitting autoregressive model

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Exercises due Nov 10, 2021 17:29 IST Completed

Fitting autoregressive model[Start of transcript. Skip to the end.](#)

Prof Jegelka: We were talking about fitting statistical models, and now we can make things very concrete by really looking how we could fit an autoregressive model. So this will actually be somewhat related

**Video**[Download video file](#)**Transcripts**[Download SubRip \(.srt\) file](#)[Download Text \(.txt\) file](#)**AR (p) Parameter Estimation**

For a given autoregressive order p , the **AR (p)** model has $p + 1$ parameters that need to be estimated from data:

$$\phi_1, \phi_2, \dots, \phi_p, \sigma_W^2.$$

We can estimate these parameters using the method of moments approach, which tells us to find $p + 1$ moments that can be estimated from data, and $p + 1$ equations that relate these estimable moments and the unknown parameters. Assuming that the time series $\{X_t\}$ is stationary, we can estimate the autocovariance function $\{\gamma_X(h)\}_{h=0}^p$ to get the required $p + 1$ moments of the series. So the first step in estimation is to compute the autocovariances:

$$\hat{\gamma}_X(0), \hat{\gamma}_X(1), \hat{\gamma}_X(2), \dots, \hat{\gamma}_X(p),$$

which we discussed before.

The second step is to find $p + 1$ equations that relate these moments to the unknown parameters above:

$$(\gamma_X(0), \gamma_X(1), \gamma_X(2), \dots, \gamma_X(p)) = \Gamma(\phi_1, \phi_2, \dots, \phi_p, \sigma_W^2).$$

We will then have $p + 1$ equations with $p + 1$ unknowns, which will in general have a unique solution, so we obtain a plug-in estimator:

$$(\hat{\phi}_1, \hat{\phi}_2, \dots, \hat{\phi}_p, \hat{\sigma}_W^2) = \Gamma^{-1}(\hat{\gamma}_X(0), \hat{\gamma}_X(1), \hat{\gamma}_X(2), \dots, \hat{\gamma}_X(p)).$$

The consistency and asymptotic normality of the resulting estimator $(\hat{\phi}_1, \hat{\phi}_2, \dots, \hat{\phi}_p, \hat{\sigma}_W^2)$ follows by the continuous mapping theorem and the Delta method from the corresponding properties of the estimator of the autocovariance function.

The equations Γ are known as the **Yule-Walker equations** and can be obtained as follows. Start from the definition of the **AR (p)** model

$$X_t = \phi_1 X_{t-1} + \phi_2 X_{t-2} + \dots + \phi_p X_{t-p} + W_t,$$

multiply both sides of the equation by the column vector

$$(X_t, X_{t-1}, \dots, X_{t-p})^\top$$

and take expectation of both side. We obtain:

$$\begin{aligned} \mathbf{E}[X_t X_t] &= \mathbf{E}[\phi_1 X_t X_{t-1} + \phi_2 X_t X_{t-2} + \dots + \phi_p X_t X_{t-p} + X_t W_t] \\ \mathbf{E}[X_{t-1} X_t] &= \mathbf{E}[\phi_1 X_{t-1} X_{t-1} + \phi_2 X_{t-1} X_{t-2} + \dots + \phi_p X_{t-1} X_{t-p} + X_{t-1} W_t] \\ &\vdots \\ \mathbf{E}[X_{t-p} X_t] &= \mathbf{E}[\phi_1 X_{t-p} X_{t-1} + \phi_2 X_{t-p} X_{t-2} + \dots + \phi_p X_{t-p} X_{t-p} + X_{t-p} W_t] \end{aligned}$$

In terms of the autocovariance function, this give us the following equations:

$$\begin{aligned} \gamma_X(0) &= \phi_1 \gamma_X(1) + \phi_2 \gamma_X(2) + \dots + \phi_p \gamma_X(p) + \sigma_W^2 \\ \gamma_X(1) &= \phi_1 \gamma_X(0) + \phi_2 \gamma_X(1) + \dots + \phi_p \gamma_X(p-1) \\ &\vdots \\ \gamma_X(p) &= \phi_1 \gamma_X(p-1) + \phi_2 \gamma_X(p-2) + \dots + \phi_p \gamma_X(0) \end{aligned}$$

ACF and AR(p) model

1/1 point (graded)

Does the autocovariance function tell us the order p of the **AR (p)** model?

☐ True

☒ False



Solution:

False. Unfortunately, the ACF does not tell us the order of the **AR (p)** model, because the ACF is non-zero for all h and decays exponentially fast.

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Yule-Walker equations

0/1 point (graded)

Is the moment function $\Gamma(\cdot)$ in Yule-Walker equations continuous?

☐ True ✓

☒ False



Solution:

True. The autocovariance terms are related to the parameters of the model via additions and multiplications, which are continuous operations. More formally, if we let

$$\mathbf{r} = \begin{pmatrix} \gamma_X(0) \\ \dots \\ \gamma_X(p) \end{pmatrix} \quad \mathbf{R} = \begin{pmatrix} \gamma(1) & \gamma(2) & \dots & \gamma(p) & 1 \\ \gamma(0) & \gamma(1) & \dots & \gamma(p-1) & 0 \\ \vdots & & & & \\ \gamma(p-1) & \gamma(p-2) & \dots & \gamma(0) & 0 \end{pmatrix} \quad \Phi = \begin{pmatrix} \phi_1 \\ \dots \\ \phi_p \\ \sigma_W^2 \end{pmatrix}$$

then the Yule-Walker equation can be written as

$$\mathbf{r} = \mathbf{R}\Phi$$

so that the function Γ is multiplication by the matrix \mathbf{R} .

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[STAFF] Is the moment function continuous?

question posted 2 months ago by [vstone9](#)

I'm not quite convinced with the solution. is Γ a matrix which is a set of equations $\gamma(h)$?

This post is visible to everyone.

Syed_SB

2 months ago - marked as answer 2 months ago by [vstone9](#)

Γ is a function/transformation which maps parameters to the the covariances.

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