

Stochastic Differential Equations - Parameter Estimation

Nathan Esau

Feb 8, 2017

We can equate the covariance function and expected value function of a stochastic differential equation and its corresponding time series model. Then we solve for the parameters of the SDE in terms of the time series model. This is known as the **principle of covariance equivalence**.

1 Ornstein-Uhlenbeck process

The AR(1) time series model corresponds to the OU model. The parameter estimates are shown below. Note that Δ is known.

1.1 AR(1) to OU

- Known parameters: ϕ_1, σ_a
- Unknown parameters: α, σ

$$\alpha = \frac{-\ln(\phi_1)}{\Delta} \tag{1}$$

$$\sigma^2 = \frac{2\alpha\sigma_a^2}{1 - \phi_1^2} \tag{2}$$

1.2 OU to AR(1)

- Known parameters: α, σ
- Unknown parameters: ϕ_1, σ_a

$$\phi_1 = e^{-\alpha\Delta} \tag{3}$$

$$\sigma_a^2 = \frac{\sigma^2}{2\alpha}(1 - \phi_1^2) \tag{4}$$

2 Second Order SDE

- Let μ_1 and μ_2 be the eigenvalues of $A = \begin{pmatrix} \alpha_1 & \alpha_0 \\ 1 & 0 \end{pmatrix}$, i.e. $\mu_1 = \frac{1}{2} \left(\alpha_1 + \sqrt{\alpha_1^2 + 4\alpha_0} \right)$
and $\mu_2 = \frac{1}{2} \left(\alpha_1 - \sqrt{\alpha_1^2 + 4\alpha_0} \right)$
- Let λ_1 and λ_2 be the roots of the AR(2) characteristic equation, i.e. $\lambda_1 = \frac{1}{2} \left(\phi_1 + \sqrt{\phi_1^2 + 4\phi_2} \right)$
and $\lambda_2 = \frac{1}{2} \left(\phi_1 - \sqrt{\phi_1^2 + 4\phi_2} \right)$

2.1 ARMA(2,1) to Second Order SDE

The simplified estimation procedure is described below.

- Known parameters: $\phi_1, \phi_2 \implies \lambda_1, \lambda_2$ are known
- Unknown parameters: $\alpha_1, \alpha_2 \implies \mu_1, \mu_2$ are unknown

First compute the eigenvalues

$$\mu_1 = \ln(\lambda_1)/\Delta \quad (5)$$

$$\mu_2 = \ln(\lambda_2)/\Delta \quad (6)$$

Next, compute the unknown SDE parameters

$$\alpha_1 = \mu_1 + \mu_2 \quad (7)$$

$$\alpha_0 = -\mu_1\mu_2 \quad (8)$$

2.2 Second Order SDE to ARMA(2, 1)

- Known parameters: $\alpha_0, \alpha_1, \sigma \implies \mu_1, \mu_2$ are known
- Unknown parameters: $\phi_1, \phi_2, \theta_1, \sigma_a \implies \lambda_1, \lambda_2$ are unknown

First, compute roots of AR(2) characteristic equation

$$\lambda_1 = e^{\mu_1\Delta} \quad (9)$$

$$\lambda_2 = e^{\mu_2\Delta} \quad (10)$$

Next, compute

$$\phi_1 = \lambda_1 + \lambda_2 \quad (11)$$

$$\phi_2 = -\lambda_1\lambda_2 \quad (12)$$

Next define $P = \frac{-\mu_1(1 + \lambda_1^2)(1 - \lambda_2^2) + \mu_2(1 + \lambda_2^2)(1 - \lambda_1^2)}{\mu_1\lambda_1(1 - \lambda_2^2) - \mu_2\lambda_2(1 - \lambda_1^2)}$. Then θ_1 is the root of the equation

$$\theta_1^2 + 2P\theta_1 + 1 = 0 \quad \text{choose the root such that } |\theta_1| < 1 \quad (13)$$

Also, we have that

$$\sigma_a^2 = \frac{\sigma^2}{2\mu_1(\mu_1^2 - \mu_2^2)} \frac{(\lambda_1 - \lambda_2)^2}{(\lambda_1 - \theta_1)} \left(\frac{\lambda_1 - \theta_1}{1 - \lambda_1^2} - \frac{\lambda_2 - \theta_1}{1 - \lambda_1\lambda_2} \right)^{-1} \quad (14)$$