Financial Time Series – (S)ARIMA processes and unit root tests

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(S)ARIMA processes

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$$X_t = m_t + s_t + Y_t.$$

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Definition

Let X be a stochastic process and let $d, D \in \mathbb{N}_0$. Then X is a SARIMA $(p, d, q) \times (P, D, Q)_s$ process if the process Y defined by $Y_t := \nabla^d \nabla_s X_t$ is a causal ARMA process defined by

$$\phi(B)\Phi(B^s)Y_t = \theta(B)\Theta(B^s)Z_t,$$

where $\phi(z) = 1 - \phi_1 z - \ldots - \phi_p z^p$, $\Phi(z) = 1 - \Phi_1 z - \ldots - \Phi_P z^P$, $\theta(z) = 1 + \theta_1 z^1 + \ldots + \theta_q z^q$ and $\Theta(z) = 1 + \Theta_1 z^1 + \ldots + \Theta_Q z^Q$ for $p, q, P, Q \in \mathbb{N}_0$.

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Y in the definition is a causal ARMA(p + sP, q + sQ) process.

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- $\phi^*(B)X_t := \phi(B)(1-B)^d X_t = \theta(B)Z_t$ where $Z \sim WN(0, \sigma^2)$ and ϕ and θ are polynomials of degree p and q
- $\phi(z) \neq 0$ for $|z| \leq 1$ while ϕ^* has a zero of degree d at z=1 \rightarrow a unit root

- When to difference a time series?
- Apply (1-B) until sample ACF is no longer "slowly decreasing".
- More systematic: *Unit root tests* (Dickey and Fuller). Restrict to ARIMA(p, d, 0) vs. AR(p).
- $H_0 = \{$ there is a unit root $\}$, $H_1 = \{$ there is no unit root $\}$
- To start: let (X_1, \ldots, X_n) be observations of

$$X_t - \mu = \phi_1(X_{t-1} - \mu) + Z_t,$$

where $Z \sim WN(0, \sigma^2)$

- When $\phi_1 < 1$, this gives an AR(1) process with mean μ so set $H_0 = \{\phi_1 = 1\}$ and $H_1 = \{\phi_1 < 1\}$
- Rewrite to

$$\nabla X_t = X_t - X_{t-1} = \phi_0^* + \phi_1^* X_{t-1} + Z_t$$

where $\phi_0^* := \mu(1 - \phi_1)$, and $\phi_1^* := \phi_1 - 1$

• Let $\hat{\phi}_1^*$ be OLS estimator of ϕ_1^* by regressing ∇X_t on 1 and X_{t-1} , i.e.,

$$(\hat{\phi}_0^*, \hat{\phi}_1^*) = \operatorname*{arg\,min}_{(\phi_0^*, \phi_1^*)} \sum_{t=2}^n (\nabla X_t - \phi_0^* - \phi_1^* X_{t-1})^2.$$

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• Then

$$\widehat{SE}(\hat{\phi}_1^*) = S\left(\sum_{t=2}^n (X_{t-1} - \bar{X}_{n-1})^2\right)^{-1/2},$$

where

$$S^{2} := (n-3)^{-1} \sum_{t=2}^{n} (\nabla X_{t} - \hat{\phi}_{0}^{*} - \hat{\phi}_{1}^{*} X_{t-1})^{2}$$

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• Dickey and Fuller derived the limit distribution for $n \to +\infty$ under $H_0: \phi_1 = 1$:

$$\hat{\tau}_{\mu} := \frac{\phi_1^*}{\widehat{\operatorname{SE}}(\hat{\phi}_1^*)}$$

• The 0.01, 0.05, and 0.10 quantiles are -3.43, -2.86, and -2.57.

• Extend to AR(p) case:

$$X_t - \mu = \sum_{j=1}^p \phi_j (X_{t-j} - \mu) + Z_t,$$

• Rewrite to

$$\nabla X_t = \phi_0^* + \phi_1^* X_{t-1} + \sum_{j=2}^p \phi_j^* \nabla X_{t+1-j} + Z_t;$$

where

$$\begin{split} \phi_0^* &:= \mu \left(1 - \sum_{i=1}^p \phi_i \right), \\ \phi_1^* &:= \sum_{i=1}^p \phi_i - 1, \\ \phi_j^* &:= - \sum_{i=j}^p \phi_i \end{split}$$

for
$$j = 2, ..., p$$
.

Method (Dickey–Fuller test)

Estimate ϕ_1^* as the coefficient of X_{t-1} in the OLS regression of ∇X_t onto 1, X_{t-1} , ∇X_{t-1} , ..., ∇X_{t-1+p} . For large *n* the *t*-ratio

$$\hat{\tau}_{\mu} := \frac{\hat{\phi}_1^*}{\widehat{\operatorname{SE}}(\hat{\phi}_1^*)},$$

where $\widehat{\rm SE}(\hat{\phi}_1^*)$ is the estimated standard error of $\hat{\phi}_1^*$, has the same limit distribution as for the AR(1) process with $0.01,\,0.05,\,{\rm and}\,0.10$ quantiles $-3.43,\,-2.86,\,{\rm and}\,-2.57,$ respectively. Test the null hypothesis $H_0:\phi_1^*=0$ and reject according to the chosen level. If a root is detected, repeat the procedure with the differenced process until rejection to determine d.