Time Series Analysis - STAT 478 - Final Exam - Part 1 Q5

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Part 1: Problem 5

Suppose that $\{Y_t\}$ is a seasonal ARIMA process in the form

$$Y_t = (1 - \theta B)(1 - \Theta B^4)e_t$$

where $\{e_t\}$ is a zero mean white noise process with variance $\sigma^2.$

Part (a)

Derive expressions for $E(Y_t)$ and $Var(Y_t)$.

Note that

$$\begin{split} Y_t &= (1 - \theta \, \mathbf{B})(1 - \Theta \, \mathbf{B}^4) e_t \\ &= (1 - \Theta \, \mathbf{B}^4 - \theta \, \mathbf{B} + \theta \Theta \, \mathbf{B}^5) e_t \\ &= e_t - \theta e_{t-1} - \Theta e_{t-4} + \theta \Theta e_{t-5}. \end{split}$$

Then,

$$\begin{split} \mathbf{E}(Y_t) &= \mathbf{E}(e_t) - \Theta \, \mathbf{E}(e_{t-4}) - \theta \, \mathbf{E}(e_{t-1}) + \theta \Theta \, \mathbf{E}(e_{t-5}) \\ &- 0 \end{split}$$

and

$$\begin{split} \operatorname{Var}(Y_t) &= \operatorname{Var}(e_t) + \Theta^2 \operatorname{Var}(e_{t-4}) + \theta^2 \operatorname{Var}(e_{t-1}) + \theta^2 \Theta^2 \operatorname{Var}(e_{t-5}) \\ &= \sigma^2 + \Theta^2 \sigma^2 + \theta^2 \sigma^2 + \theta^2 \Theta^2 \sigma^2 \\ &= \sigma^2 (1 + \Theta^2 + \theta^2 + \theta^2 \Theta^2) \\ &= \sigma^2 (1 + \Theta^2) (1 + \theta^2). \end{split}$$

Part (b)

Derive the autocovariance function, that is, calculate $\mathrm{Cov}(Y_t,Y_{t-k})$ for $k=1,2,\dots$

The autocovariance function of $\{Y_t\}$ is given by

$$\gamma_k = \begin{cases} \sigma^2(1+\Theta^2)(1+\theta^2) & k = 0 \\ -\theta\sigma^2(1+\Theta^2) & k = 1 \\ 0 & k = 2 \\ \theta\Theta\sigma^2 & k = 3 \\ -\Theta\sigma^2(1+\theta\Theta) & k = 4 \\ \theta\Theta\sigma^2 & k = 5 \\ 0 & k > 5. \end{cases}$$

Proof. We do a case analysis on values of k = 1, 2, ..., but first note that by symmetry

$$Cov(Y_{t-k}, Y_t) = Cov(Y_t, Y_{t-k})$$

and so we only consider non-negative values of k.

1. Let k = 1, then

$$\begin{split} \operatorname{Cov}(Y_t, Y_{t-1}) &= \operatorname{Cov}(e_t - \theta e_{t-1} - \Theta e_{t-4} + \theta \Theta e_{t-5}, e_{t-1} - \theta e_{t-2} - \Theta e_{t-5} + \theta \Theta e_{t-6}) \\ &= \operatorname{Cov}(-\theta e_{t-1}, e_{t-1}) + \operatorname{Cov}(\theta \Theta e_{t-5}, -\Theta e_{t-5}) \\ &= -\theta \operatorname{Cov}(e_{t-1}, e_{t-1}) - \theta \Theta^2 \operatorname{Cov}(e_{t-5}, e_{t-5}) \\ &= -\theta \sigma^2 - \theta \Theta^2 \sigma^2 \\ &= -\theta \sigma^2 (1 + \Theta^2). \end{split}$$

2. Let k=2, then

$$\begin{aligned} \text{Cov}(Y_t, Y_{t-2}) &= \text{Cov}(e_t - \theta e_{t-1} - \Theta e_{t-4} + \theta \Theta e_{t-5}, e_{t-2} - \theta e_{t-3} - \Theta e_{t-6} + \theta \Theta e_{t-7}) \\ &= 0. \end{aligned}$$

3. Let k=3, then

$$\begin{split} \operatorname{Cov}(Y_t, Y_{t-3}) &= \operatorname{Cov}(e_t - \theta e_{t-1} - \Theta e_{t-4} + \theta \Theta e_{t-5}, e_{t-3} - \theta e_{t-4} - \Theta e_{t-7} + \theta \Theta e_{t-8}) \\ &= \operatorname{Cov}(-\Theta e_{t-4}, -\theta e_{t-4}) \\ &= \theta \Theta \operatorname{Cov}(e_{t-4}, e_{t-4}) \\ &= \theta \Theta \sigma^2. \end{split}$$

4. Let k = 4, then

$$\begin{split} \operatorname{Cov}(Y_t,Y_{t-4}) &= \operatorname{Cov}(e_t - \theta e_{t-1} - \Theta e_{t-4} + \theta \Theta e_{t-5}, e_{t-4} - \theta e_{t-5} - \Theta e_{t-8} + \theta \Theta e_{t-9}) \\ &= \operatorname{Cov}(-\Theta e_{t-4}, e_{t-4}) + \operatorname{Cov}(\theta \Theta e_{t-5}, -\Theta e_{t-5}) \\ &= -\Theta \operatorname{Cov}(e_{t-4}, e_{t-4}) - \theta \Theta^2 \operatorname{Cov}(e_{t-5}, e_{t-5}) \\ &= -\Theta \sigma^2 - \theta \Theta^2 \sigma^2 \\ &= \sigma^2 (-\Theta - \theta \Theta^2) \\ &= -\Theta \sigma^2 (1 + \theta \Theta). \end{split}$$

5. Let k = 5, then

$$\begin{split} \operatorname{Cov}(Y_t, Y_{t-5}) &= \operatorname{Cov}(e_t - \theta e_{t-1} - \Theta e_{t-4} + \theta \Theta e_{t-5}, e_{t-5} - \theta e_{t-6} - \Theta e_{t-9} + \theta \Theta e_{t-10}) \\ &= \operatorname{Cov}(\theta \Theta e_{t-1}, e_{t-5}) \\ &= \theta \Theta \operatorname{Cov}(e_{t-1}, e_{t-5}) \\ &= \theta \Theta \sigma^2. \end{split}$$

6. Let k > 5, then $Cov(Y_t, Y_{t-k}), k > 5$, is zero.

We see that the autocovariance function is independent of time t and only a function of lag k. We collect the cases and form the indicated piecewise function.

Part (c)

Characterize this models as SARIMA $(p,d,q) \times (P,D,Q)_s$, that is, identify p,d,q,P,D,Q,s.

 $\{Y_t\}$ is a zero mean process of type SARIMA $(0,1,1)\times(0,1,1)_4$, i.e., $p=0,\ d=1,\ q=1,\ P=0,\ D=1,\ Q=1,\ {\rm and}\ s=4.$