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 SMIT S-136 Time Series Analysis with Python
 Part I of Assignment 6

Let x_t be a causal AR(2) process:

$$x_t = \phi_1 x_{t-1} + \phi_2 x_{t-2} + w_t, \text{ where } w_t \stackrel{\text{i.i.d.}}{\sim} \mathcal{N}(0, \sigma_w^2).$$

Assume we know ϕ_1 , ϕ_2 , and σ_w . Suppose we observe x_1, x_2 , and x_3 , i.e. $n = 3$.

- (a) Find the minimum mean square error predictor x_4^3 (the superscript "3" here indicates that the predictor is based on x_1, x_2, x_3) of x_4 using the conditional expectation:

$$x_4^3 = E[x_4 | x_1, x_2, x_3].$$

- (b) Find the linear minimum mean square error predictor x_4^2 of x_4 . You need to justify your answer here.
 (c) For the predictors you obtained in (a) and (b), find the one-step-ahead prediction error P_4^3 .
 (d) In both cases (a) and (b), construct the 95% prediction interval for x_4^3 , that is, an interval (L, R) , such that

$$P(L \leq x_4 \leq R | x_1, x_2, x_3) = 0.95$$

SOLUTION:

$$\begin{aligned} \text{a) } x_4^3 &= E[x_4 | x_1, x_2, x_3] = E[\phi_1 x_3 + \phi_2 x_2 + w_4 | x_1, x_2, x_3] \\ &= \underline{\phi_1 x_3 + \phi_2 x_2} \quad \text{as } x_t \text{ is a causal AR(2) process} \end{aligned}$$

$$\text{b) } x_4^3 = \underline{\phi_1 x_3 + \phi_2 x_2}$$

\Rightarrow As $E[x_4 | x_1, x_2, x_3] = E[\phi_1 x_3 + \phi_2 x_2 + w_4 | x_1, x_2, x_3]$ is already a linear prediction of x_4 , the predictors in a) and b) are the same

$$\text{c) } P_4^3 = E[(x_4 - x_4^3)^2] = E[(\phi_1 x_3 + \phi_2 x_2 + w_4 - \phi_1 x_3 - \phi_2 x_2)^2] = E[w_4^2] = \underline{\sigma_w^2}$$

\Rightarrow As AR(2) is a causal process and x_1, x_2, x_3 , so all x_t with $t \leq n-t-4$, is observable, the prediction error is σ_w^2

$$\text{d) Interval bounds: } x_{n+1}^{\pm} \pm 1.96 \cdot \sqrt{P_{n+1}^{\pm}} = x_4^3 \pm 1.96 \cdot \sqrt{P_4^3}$$

$$\Rightarrow \text{Positive bound: } \phi_1 x_3 + \phi_2 x_2 + 1.96 \cdot \sqrt{\sigma_w^2} = \phi_1 x_3 + \phi_2 x_2 + 1.96 \sigma_w$$

$$\Rightarrow \text{Negative bound: } \phi_1 x_3 + \phi_2 x_2 - 1.96 \cdot \sqrt{\sigma_w^2} = \phi_1 x_3 + \phi_2 x_2 - 1.96 \sigma_w$$

$$\Rightarrow \underline{[\phi_1 x_3 + \phi_2 x_2 + 1.96 \sigma_w, \phi_1 x_3 + \phi_2 x_2 - 1.96 \sigma_w]}$$