

# Seminar 8 Solutions

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## Disclaimer

Full solutions are available on my.wbs. All exercises are examinable material, not just the ones we covered in the seminars.

# Roadmap

# Question 1: Forecasting with ARIMA(1,2)

Write down a set of equations to produce one-step, two-step, and three-step ahead forecasts for  $y_t$ , given that it follows an ARIMA(1,2) process:

$$y_t = \underbrace{\phi y_{t-1}}_{\text{AR}(1)} + \varepsilon_t + \underbrace{\theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2}}_{\text{MA}(2)}, \quad \varepsilon_t \sim N(0, \sigma_t^2).$$

$$y_c = \phi y_{t-1} + \varepsilon_t + \theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2} \quad \varepsilon_t \sim N(0, \sigma^2)$$

## Roadmap

$t+1$

$$y_{t+1} = \phi y_t + \varepsilon_{t+1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1} \leftarrow$$

$t+2$

$$y_{t+2} = \phi y_{t+1} + \varepsilon_{t+2} + \theta_1 \varepsilon_{t+1} + \theta_2 \varepsilon_t \leftarrow$$

$t+3$

$$y_{t+3} = \phi y_{t+2} + \varepsilon_{t+3} + \theta_1 \varepsilon_{t+2} + \theta_2 \varepsilon_{t+1}$$

## FORECAST

$$\underline{E[y_{t+1} | \mathcal{I}_t]} = E_t[y_{t+1}]$$

$$\begin{aligned} &= E_t[\phi y_t + \varepsilon_{t+1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1}] \\ &= \phi E_t[y_t] + \underline{E_t[\varepsilon_{t+1}]} + \theta_1 E_t[\varepsilon_t] + \theta_2 E_t[\varepsilon_{t-1}] \\ &= \phi y_t + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1} \end{aligned}$$

$$\begin{aligned} E_t[y_{t+2}] &= E_t[\phi y_{t+1} + \varepsilon_{t+2} + \theta_1 \varepsilon_{t+1} + \theta_2 \varepsilon_t] \\ &= \phi \underline{E_t[y_{t+1}]} + \theta_2 \varepsilon_t \end{aligned}$$

$$\begin{aligned} &= \phi (\phi y_t + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1}) + \theta_2 \varepsilon_t \\ &= \phi^2 y_t + \phi \theta_1 \varepsilon_t + \phi \theta_2 \varepsilon_{t-1} + \theta_2 \varepsilon_t \end{aligned}$$

$$= \phi^2 y_t + \varepsilon_t (\phi \theta_1 + \theta_2) + \phi \theta_2 \varepsilon_{t-1}$$

$$E_c[y_{t+3}] = E_c[\phi y_{t+2} + \varepsilon_{t+3} + \theta_1 \varepsilon_{t+2} + \theta_2 \varepsilon_{t+1}]$$
$$= \phi E_c[y_{t+2}]$$

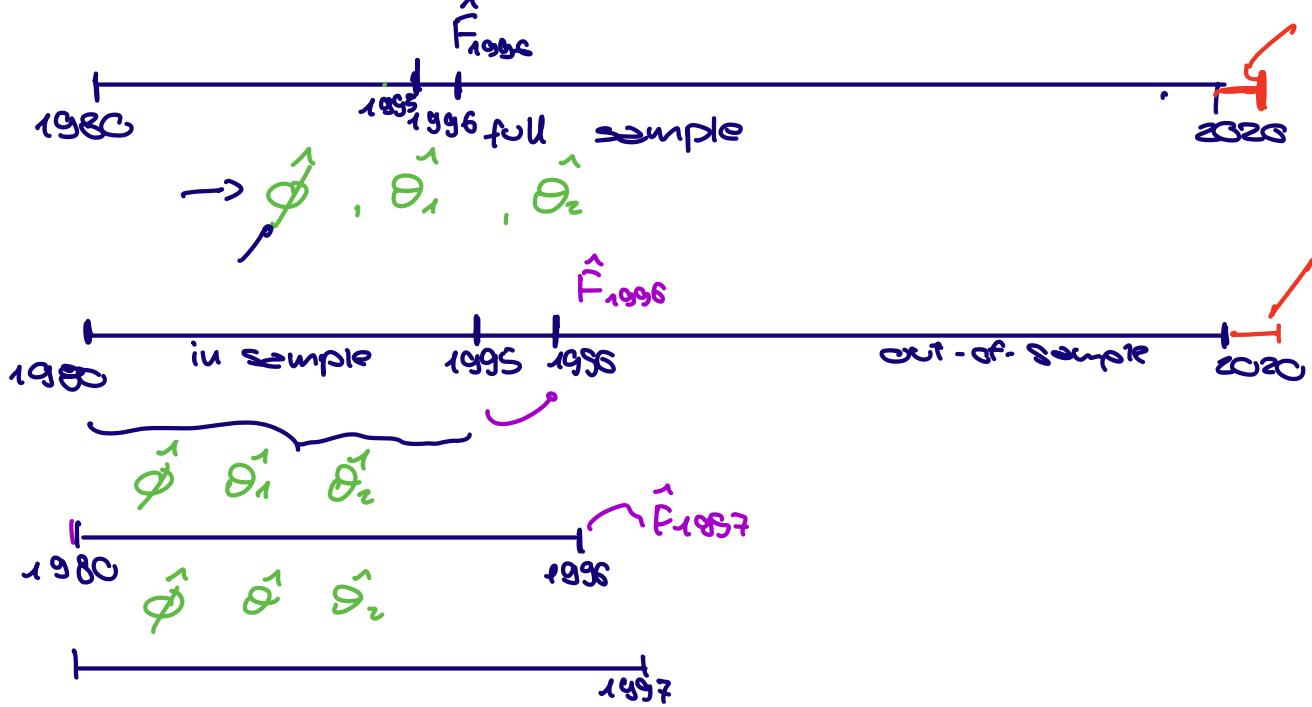
# Question 2: Pseudo Out-of-Sample Forecasting

Outline the stages of a **pseudo out-of-sample** forecasting evaluation, comparing a "benchmark" AR(1) with the ARIMA(1,2) model.

Hints:

- (i) Recursive/repeated estimation with an expanding sample.
- (ii) Choice of  $h$ -steps forecasting at each point.
- (iii) Forecast error comparison via root mean square forecast error (RMSE) and statistical testing.

ARIMA(1,2)  $y_t = \phi y_{t-1} + \varepsilon_t + \theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2}$   $\varepsilon_t \sim N(0, \sigma^2)$



ARIMA(1,2)

AR(1)

$$\begin{array}{c} \hat{F}_{1996} \quad F_{1996} \quad \hat{\epsilon}_{1996} = F_{1996} - \hat{F}_{1996} \\ \vdots \quad \vdots \\ \hat{F}_{2020} \quad \hat{\epsilon}_{2020} \end{array}$$

$$MSE = \frac{1}{T} \sum \hat{\epsilon}_t^2$$

$$RMSE_{ARIMA} = \sqrt{MSE}$$

$$RMSE_{AR} = \sqrt{\frac{MSE}{T}}$$

$$\bar{T} = \frac{RMSE_{ARIMA}}{RMSE_{AR}}$$

# Roadmap

## Question 3: Forecast Uncertainty in MA(2)

would you evaluate forecast uncertainty of 1-step and 2-step ahead forecasts?

Assume the forecasts come from an MA(2) process.

**Hint:** Consider the variances of the forecasts at each step.

$$Y_t = \varepsilon_t + \theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2} \quad \varepsilon_t \sim N(0, \sigma^2_\varepsilon)$$

$t+1$

$$Y_{t+1} = \varepsilon_{t+1} + \theta_1 \varepsilon_t + \theta_2 \varepsilon_{t-1} \quad A$$

$$E_c[Y_{t+1}] = \theta_1 E_c \varepsilon_t + \theta_2 E_c \varepsilon_{t-1} \quad B$$

$t+2$

$$Y_{t+2} = \varepsilon_{t+2} + \theta_1 \varepsilon_{t+1} + \theta_2 \varepsilon_t \quad C$$

$$E_c[Y_{t+2}] = \theta_2 E_c \varepsilon_t$$

$t+3$

$$Y_{t+3} = \varepsilon_{t+3} + \theta_1 \varepsilon_{t+2} + \theta_2 \varepsilon_{t+1}$$

$$E_c[Y_{t+3}] = 0$$

$$E_c[Y_{t+h}] = 0 \quad h > 2$$

## FORECAST ERROR

$$\epsilon_{t+1,c} = Y_{t+1} - E_c[Y_{t+1}] = \varepsilon_{t+1}$$

$$\epsilon_{t+2,c} = Y_{t+2} - E_c[Y_{t+2}] = \varepsilon_{t+2} + \theta_1 \varepsilon_{t+1}$$

$$\epsilon_{t+3,c} = Y_{t+3} - E_c[Y_{t+3}] = \varepsilon_{t+3} + \theta_1 \varepsilon_{t+2} + \theta_2 \varepsilon_{t+1}$$

## VARIANCE FC

$$\text{var}(\epsilon_{t+1,c}) = \text{var}(\varepsilon_{t+1}) = \sigma^2_\varepsilon$$

$$\begin{aligned} \text{var}(\epsilon_{t+2,c}) &= \text{var}(\varepsilon_{t+2} + \theta_1 \varepsilon_{t+1}) \\ &= \sigma^2_\varepsilon + \theta_1^2 \sigma^2_\varepsilon + 2 \text{cov}(\varepsilon_{t+2}, \theta_1 \varepsilon_{t+1}) \\ &= \sigma^2_\varepsilon (1 + \theta_1^2) \end{aligned}$$

$$\begin{aligned} \text{var}(\epsilon_{t+3,c}) &= \text{var}(\varepsilon_{t+3} + \theta_1 \varepsilon_{t+2} + \theta_2 \varepsilon_{t+1}) \\ &= \sigma^2_\varepsilon + \theta_1^2 \sigma^2_\varepsilon + \theta_2^2 \sigma^2_\varepsilon \end{aligned}$$

$$\stackrel{1}{=} \sigma_{\varepsilon}^2 (\ell + \partial_1^2 + \partial_2^2) \quad \text{at}$$