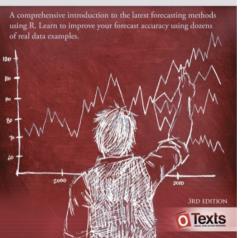
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FORECASTING PRINCIPLES AND PRACTICE



9. ARIMA models

9.8 Forecasting
OTexts.org/fpp3/

- Rearrange ARIMA equation so y_t is on LHS.
- Rewrite equation by replacing t by T + h.
- On RHS, replace future observations by their forecasts, future errors by zero, and past errors by corresponding residuals.

Start with h = 1. Repeat for h = 2, 3, ...

$$(1-\hat{\phi}_1B-\hat{\phi}_2B^2-\hat{\phi}_3B^3)(1-B)y_t=(1+\hat{\theta}_1B)\varepsilon_t,$$

ARIMA(3,1,1) forecasts: Step 1

$$(1 - \hat{\phi}_1 B - \hat{\phi}_2 B^2 - \hat{\phi}_3 B^3)(1 - B)y_t = (1 + \hat{\theta}_1 B)\varepsilon_t,$$

$$\left[1 - \hat{\phi}_{1}B - \hat{\phi}_{2}B^{2} - \hat{\phi}_{3}B^{3} - B + \hat{\phi}_{1}B^{2} + \hat{\phi}_{2}B^{3} + \hat{\phi}_{3}B^{4}\right]y_{t} = (1 + \hat{\theta}_{1}B)\varepsilon_{t}$$

3

ARIMA(3,1,1) forecasts: Step 1

$$(1 - \hat{\phi}_1 B - \hat{\phi}_2 B^2 - \hat{\phi}_3 B^3)(1 - B)y_t = (1 + \hat{\theta}_1 B)\varepsilon_t$$

$$\left[1 - \hat{\phi}_{1}B - \hat{\phi}_{2}B^{2} - \hat{\phi}_{3}B^{3} - B + \hat{\phi}_{1}B^{2} + \hat{\phi}_{2}B^{3} + \hat{\phi}_{3}B^{4}\right]y_{t} = (1 + \hat{\theta}_{1}B)\varepsilon_{t}$$

$$\left[1 - (1 + \hat{\phi}_1)B + (\hat{\phi}_1 - \hat{\phi}_2)B^2 + (\hat{\phi}_2 - \hat{\phi}_3)B^3 + \hat{\phi}_3B^4\right]y_t = (1 + \hat{\theta}_1B)\varepsilon_t$$

3

ARIMA(3,1,1) forecasts: Step 1

$$(1 - \hat{\phi}_1 B - \hat{\phi}_2 B^2 - \hat{\phi}_3 B^3)(1 - B)y_t = (1 + \hat{\theta}_1 B)\varepsilon_t$$

$$\left[1 - \hat{\phi}_{1}B - \hat{\phi}_{2}B^{2} - \hat{\phi}_{3}B^{3} - B + \hat{\phi}_{1}B^{2} + \hat{\phi}_{2}B^{3} + \hat{\phi}_{3}B^{4}\right]y_{t} = (1 + \hat{\theta}_{1}B)\varepsilon_{t}$$

$$\left[1 - (1 + \hat{\phi}_1)B + (\hat{\phi}_1 - \hat{\phi}_2)B^2 + (\hat{\phi}_2 - \hat{\phi}_3)B^3 + \hat{\phi}_3B^4\right]y_t = (1 + \hat{\theta}_1B)\varepsilon_t$$

$$y_{t} - (1 + \hat{\phi}_{1})y_{t-1} + (\hat{\phi}_{1} - \hat{\phi}_{2})y_{t-2} + (\hat{\phi}_{2} - \hat{\phi}_{3})y_{t-3} + \hat{\phi}_{3}y_{t-4} = \varepsilon_{t} + \hat{\theta}_{1}\varepsilon_{t-1}$$

3

$$(1 - \hat{\phi}_1 B - \hat{\phi}_2 B^2 - \hat{\phi}_3 B^3)(1 - B)y_t = (1 + \hat{\theta}_1 B)\varepsilon_t$$

$$\left[1 - \hat{\phi}_{1}B - \hat{\phi}_{2}B^{2} - \hat{\phi}_{3}B^{3} - B + \hat{\phi}_{1}B^{2} + \hat{\phi}_{2}B^{3} + \hat{\phi}_{3}B^{4}\right]y_{t} = (1 + \hat{\theta}_{1}B)\varepsilon_{t}$$

$$\left[1-(1+\hat{\phi}_1)B+(\hat{\phi}_1-\hat{\phi}_2)B^2+(\hat{\phi}_2-\hat{\phi}_3)B^3+\hat{\phi}_3B^4\right]\mathsf{y}_t=(1+\hat{\theta}_1B)\varepsilon_t$$

$$y_{t} - (1 + \hat{\phi}_{1})y_{t-1} + (\hat{\phi}_{1} - \hat{\phi}_{2})y_{t-2} + (\hat{\phi}_{2} - \hat{\phi}_{3})y_{t-3} + \hat{\phi}_{3}y_{t-4} = \varepsilon_{t} + \hat{\theta}_{1}\varepsilon_{t-1}$$

$$\mathsf{y}_t = (1 + \hat{\phi}_1) \mathsf{y}_{t-1} - (\hat{\phi}_1 - \hat{\phi}_2) \mathsf{y}_{t-2} - (\hat{\phi}_2 - \hat{\phi}_3) \mathsf{y}_{t-3} - \hat{\phi}_3 \mathsf{y}_{t-4} + \varepsilon_t + \hat{\theta}_1 \varepsilon_{t-1}$$

Point forecasts (h=1)

$$y_{t} = (1 + \hat{\phi}_{1})y_{t-1} - (\hat{\phi}_{1} - \hat{\phi}_{2})y_{t-2} - (\hat{\phi}_{2} - \hat{\phi}_{3})y_{t-3} - \hat{\phi}_{3}y_{t-4} + \varepsilon_{t} + \hat{\theta}_{1}\varepsilon_{t-1}$$

Point forecasts (h=1)

$$y_{t} = (1 + \hat{\phi}_{1})y_{t-1} - (\hat{\phi}_{1} - \hat{\phi}_{2})y_{t-2} - (\hat{\phi}_{2} - \hat{\phi}_{3})y_{t-3} - \hat{\phi}_{3}y_{t-4} + \varepsilon_{t} + \hat{\theta}_{1}\varepsilon_{t-1}$$

$$y_{T+1} = (1 + \hat{\phi}_1)y_T - (\hat{\phi}_1 - \hat{\phi}_2)y_{T-1} - (\hat{\phi}_2 - \hat{\phi}_3)y_{T-2} - \hat{\phi}_3y_{T-3} + \varepsilon_{T+1} + \hat{\theta}_1\varepsilon_T$$

Point forecasts (h=1)

$$y_{t} = (1 + \hat{\phi}_{1})y_{t-1} - (\hat{\phi}_{1} - \hat{\phi}_{2})y_{t-2} - (\hat{\phi}_{2} - \hat{\phi}_{3})y_{t-3} - \hat{\phi}_{3}y_{t-4} + \varepsilon_{t} + \hat{\theta}_{1}\varepsilon_{t-1}$$

ARIMA(3,1,1) forecasts: Step 2

$$y_{T+1} = (1 + \hat{\phi}_1)y_T - (\hat{\phi}_1 - \hat{\phi}_2)y_{T-1} - (\hat{\phi}_2 - \hat{\phi}_3)y_{T-2} - \hat{\phi}_3y_{T-3} + \varepsilon_{T+1} + \hat{\theta}_1\varepsilon_T$$

$$\hat{y}_{T+1|T} = (1 + \hat{\phi}_1)y_T - (\hat{\phi}_1 - \hat{\phi}_2)y_{T-1} - (\hat{\phi}_2 - \hat{\phi}_3)y_{T-2} - \hat{\phi}_3y_{T-3} + \hat{\theta}_1e_T$$

Point forecasts (h=2)

$$y_{t} = (1 + \hat{\phi}_{1})y_{t-1} - (\hat{\phi}_{1} - \hat{\phi}_{2})y_{t-2} - (\hat{\phi}_{2} - \hat{\phi}_{3})y_{t-3} - \hat{\phi}_{3}y_{t-4} + \varepsilon_{t} + \hat{\theta}_{1}\varepsilon_{t-1}$$

Point forecasts (h=2)

$$y_{t} = (1 + \hat{\phi}_{1})y_{t-1} - (\hat{\phi}_{1} - \hat{\phi}_{2})y_{t-2} - (\hat{\phi}_{2} - \hat{\phi}_{3})y_{t-3} - \hat{\phi}_{3}y_{t-4} + \varepsilon_{t} + \hat{\theta}_{1}\varepsilon_{t-1}$$

$$\mathbf{y}_{T+2} = (\mathbf{1} + \hat{\phi}_1)\mathbf{y}_{T+1} - (\hat{\phi}_1 - \hat{\phi}_2)\mathbf{y}_T - (\hat{\phi}_2 - \hat{\phi}_3)\mathbf{y}_{T-1} - \hat{\phi}_3\mathbf{y}_{T-2} + \varepsilon_{T+2} + \hat{\theta}_1\varepsilon_{T+1}.$$

Point forecasts (h=2)

$$y_{t} = (1 + \hat{\phi}_{1})y_{t-1} - (\hat{\phi}_{1} - \hat{\phi}_{2})y_{t-2} - (\hat{\phi}_{2} - \hat{\phi}_{3})y_{t-3} - \hat{\phi}_{3}y_{t-4} + \varepsilon_{t} + \hat{\theta}_{1}\varepsilon_{t-1}$$

ARIMA(3,1,1) forecasts: Step 2

$$y_{T+2} = (1+\hat{\phi}_1)y_{T+1} - (\hat{\phi}_1 - \hat{\phi}_2)y_T - (\hat{\phi}_2 - \hat{\phi}_3)y_{T-1} - \hat{\phi}_3y_{T-2} + \varepsilon_{T+2} + \hat{\theta}_1\varepsilon_{T+1}.$$

$$\hat{y}_{T+2|T} = (1 + \hat{\phi}_1)\hat{y}_{T+1|T} - (\hat{\phi}_1 - \hat{\phi}_2)y_T - (\hat{\phi}_2 - \hat{\phi}_3)y_{T-1} - \hat{\phi}_3y_{T-2}.$$

Assuming $\varepsilon_t \sim N(0, \sigma^2)$

95% prediction interval

$$\hat{y}_{T+h|T} \pm 1.96 \sqrt{v_{T+h|T}}$$

where $v_{T+h|T}$ is estimated forecast variance.

Assuming $\varepsilon_t \sim N(0, \sigma^2)$

95% prediction interval

$$\hat{\mathbf{y}}_{T+h|T} \pm 1.96\sqrt{\mathbf{v}_{T+h|T}}$$

where $v_{T+h|T}$ is estimated forecast variance.

 $\mathbf{v}_{T+1|T} = \hat{\sigma}^2$ for all ARIMA models

Assuming $\varepsilon_t \sim N(0, \sigma^2)$

95% prediction interval

$$\hat{y}_{T+h|T} \pm 1.96 \sqrt{v_{T+h|T}}$$

where $v_{T+h|T}$ is estimated forecast variance.

- $\mathbf{v}_{T+1|T} = \hat{\sigma}^2$ for all ARIMA models
- Multi-step prediction intervals for ARIMA(0,0,q):

$$y_{t} = \varepsilon_{t} + \sum_{i=1}^{q} \hat{\theta}_{i} \varepsilon_{t-i}.$$

$$v_{T|T+h} = \hat{\sigma}^{2} \left[1 + \sum_{i=1}^{h-1} \hat{\theta}_{i}^{2} \right], \quad \text{for } h = 2, 3, \dots.$$

Assuming $\varepsilon_t \sim N(0, \sigma^2)$

95% prediction interval

$$\hat{\mathbf{y}}_{\mathsf{T}+h|\mathsf{T}} \pm 1.96\sqrt{\mathsf{v}_{\mathsf{T}+h|\mathsf{T}}}$$

where $v_{T+h|T}$ is estimated forecast variance.

- $\mathbf{v}_{T+1|T} = \hat{\sigma}^2$ for all ARIMA models
- Multi-step prediction intervals for ARIMA(0,0,q):

$$y_t = \varepsilon_t + \sum_{i=1}^q \hat{\theta}_i \varepsilon_{t-i}.$$

$$v_{T|T+h} = \hat{\sigma}^2 \left[1 + \sum_{i=1}^{h-1} \hat{\theta}_i^2 \right], \quad \text{for } h = 2, 3, \dots$$

Other models beyond scope of this book.

- Prediction intervals increase in size with forecast horizon.
- Prediction intervals can be difficult to calculate by hand
- Calculations assume residuals are uncorrelated and normally distributed.
- Prediction intervals tend to be too narrow.
 - the uncertainty in the parameter estimates has not been accounted for.
 - model uncertainty has not been accounted for.
 - the ARIMA model assumes historical patterns will not change during the forecast period.
 - ▶ the ARIMA model assumes uncorrelated future errors.