

ECON 4115/5115

Chapter 5. The Forecaster's Toolbox



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➤ Forecasting workflow:

- Data preparation
- Plot the data
- Define the forecasting model
- Estimate the model
- Check the model's in-sample performance
- Produce forecasts

➤ Some simple forecasting methods:

- Average method
- Naïve method
- Seasonal naïve method

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➤ Fitted values and residuals (in-sample):

- Fitted values: $\hat{y}_{t|t-1}$
- Residuals: $e_t = y_t - \hat{y}_{t|t-1}$

➤ Residual diagnostics

- Two most important features: residuals should be mean zero and white noise

➤ Prediction intervals

- Based on probability distributions
- One-step and multi-step prediction intervals: $\hat{y}_{T+h|T} \pm c \cdot \hat{\sigma}_h$

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➤ Forecasting using transformations:

- Box-Cox transformation can be used to make variation and/or seasonality constant.

- Box-Cox transformation: $w_t = \begin{cases} \log(y_t) & \text{if } \lambda = 0 \\ (y_t^\lambda - 1)/\lambda & \text{otherwise} \end{cases}$

- Reverse Box-Cox transformation: $y_t = \begin{cases} \exp(w_t) & \text{if } \lambda = 0 \\ (\lambda w_t + 1)^{1/\lambda} & \text{otherwise} \end{cases}$

- The *fable* package will automatically produce confidence intervals.

- The back-transformed forecast confidence interval may be asymmetric.

➤ Forecasting using decompositions

- The *decomposition_model()* function can be a useful forecasting tool.

- The seasonal, trend and remainder components can be forecasted separately.

- The forecasts are then added (or multiplied) together to produce the final forecast.



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➤ Evaluating forecasting accuracy

- The size of the residuals are not always a great indicator of forecasting accuracy.
- Over-fitting can reduce in-sample residuals, but won't always improve forecasting.
- It's better to check the out-of-sample forecasting accuracy.
- Textbook terminology:
 - “training” data = “in-sample” data
 - “test” data = “out-of-sample” data
- The *filter()* and *slice()* functions can help split the sample.
- Residuals vs. Forecast Errors
 - Residuals: $e_t = y_t - \hat{y}_{t|t-1}$
 - Forecast errors: $e_{T+h} = y_{T+h} - \hat{y}_{T+h|T}$



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- Scale-dependent accuracy measures:

- Mean absolute error (MAE): $\text{mean}(|e_t|)$
- Root mean squared error (RMSE): $\sqrt{\text{mean}(e_t^2)}$

- Scale-independent accuracy measures:

- Mean absolute percentage error (MAPE): $\text{mean}(|100e_t/y_t|)$
- Symmetric MAPE (sMAPE)
- Mean absolute scaled errors (MASE), where

$$\text{MASE} = \text{mean}(|q_j|), q_j = \frac{e_j}{\frac{1}{T-m} \sum_{t=m+1}^T |y_t - y_{t-m}|}$$

- The *accuracy()* function will calculate the RMSE, MAE, MAPE & MASE.