

Boston Marathon

1. (a) The data `boston_marathon` is in the `fpp3` package. Plot the winning times in the Men's open division. Note that the winning times are stored as a variable of type `Time`. That won't play well with our model later, so convert them to numeric.
- (b) Fit a simple exponential smoothing model (additive error, no trend) to the winning times. What value of the smoothing parameter α was chosen? Make residual plots and comment on the fit of the model.
- (c) Plot the data with the fitted values.
- (d) Make a 10-year forecast of winning times and plot it with the data.

Mexico GDP

2. (a) The data `global_economy` is in the `fpp3` package. Plot the gross domestic product (GDP) of Mexico.
- (b) Fit two ETS models to Mexico's GDP: One with additive error and additive trend (Holt's linear method), one with additive error and damped trend. Make 10-year forecasts for each and plot along with the original data.
- (c) Fit the same two models but use the log of GDP. Plot the forecasts and compare.

Cement

3. (a) The data `aus_production` is in the `fpp3` package. Plot the Australian production of Cement. This is a quarterly series.
 - (b) Auto-fit an ETS model to cement production. What model was chosen? What was the level smoothing parameter α ?
 - (c) Plot the data and a 5-year forecast.
4. Cross validation
 - (a) Create a stretched data window with initial size of 5 years. This should result in 199 series.
 - (b) Fit three models to the 199 training series: 1) Holt-Winters multiplicative, which has multiplicative error and season components, and additive trend. 2) Holt-Winters multiplicative with damped trend. 3) Seasonal naive.
 - (c) Create a 1-year forecast for all models on all series and check the accuracy against the original data. Which model performed the best for this 1-year forecast?