

Eurostat Nowcasting: Methodological Note

Haseeb Mahmud

This is the abstract. It consists of two paragraphs.

Introduction

The September submission is based on the combination/ensemble of three statistical forecasting methods, namely, Autoregressive Integrated Moving Average (ARIMA), Error-Trend-Seasonality (ETS) and Theta model.

Statistical Methods

Autoregressive Integrated Moving Average

Autoregressive integrated moving average models (ARIMA) has three components, namely,

- The autoregressive component p ,
- The number of nonseasonal differences needed for stationarity d
- Moving average or lagged forecasted error q

To begin with, the time series is required to be stationary. One way to make the series stationary is to compute the differences between consecutive observations. The procedure is called *differencing*. This can be written as,

$$y'_t = y_t - y_{t-1} \quad (1)$$

If the series is differenced with one lag, it is called *first difference*, or *first-order differencing* i.e. $d = 1$. In case the series is still non-stationary, further differencing is needed. For example, the second-order differencing looks like this,

$$y''_t = y'_t - y'_{t-1} \quad (2)$$

Differencing the data twice means, $d = 2$.

The next component in ARIMA model is called *autoregressive component*. An autoregressive model of order p can be written as,

$$y_t = c + \phi_1 y_{t-1} + \phi_2 y_{t-2} + \dots + \phi_p y_{t-p} + \epsilon_t \quad (3)$$

where, ϵ_t is white noise. Here we are predicting the series using its past values. We can refer the form Equation 1 as $AR(p)$.

Instead of using past values to predict the future values, it is possible to use past forecast errors,

$$y_t = c + \epsilon_t + \theta_1 \epsilon_{t-1} + \theta_2 \epsilon_{t-2} + \dots + \theta_q \epsilon_{t-q} \quad (4)$$

where ϵ_t is white noise. The Equation 4 is called $MA(q)$ model or moving average model of order q .

If we combine differencing (d) with autoregressive (p) and moving average (q) component, we get the full Autoregressive Integrated Moving Average (ARIMA) model, which can be expressed as,

$$y'_t = c + \phi_1 y_{t-1} + \phi_2 y_{t-2} + \dots + \phi_p y_{t-p} + \theta_1 \epsilon_{t-1} + \theta_2 \epsilon_{t-2} + \dots + \theta_q \epsilon_{t-q} + \epsilon_t \quad (5)$$