

# Box-Jenkins approach I

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# Overview

1. Steps of Box-Jenkins approach
2. Stationarity of time series:
  - 2.1 Determination of difference filter  $d$
  - 2.2 Unit root test by Dickey-Fuller
3. Empirical example: Dow Jones

# The Box-Jenkins Methodology

## Overview

The Box-Jenkins methodology is a systematic, iterative framework for building and refining ARIMA models for time series analysis. It consists of four steps:

1. **Identification**
2. **Estimation and Selection**
3. **Model Diagnosis**
4. **Prediction**

## Step 1: Identification

**Goal:** Identify the structure of the ARIMA model  $(p, d, q)$ .

- ▶ **Stationarity:** Verify whether the time series is stationary:
  - ▶ Visual inspection (trends, seasonality)
  - ▶ Statistical tests (e.g., Augmented Dickey-Fuller test)
- ▶ **Differencing ( $d$ ):** Apply differencing if the series is non-stationary:

$$Y'_t = Y_t - Y_{t-1} \quad (\text{first-order differencing})$$

- ▶ **Order Selection:** Use:
  - ▶ **ACF (Autocorrelation Function)** to analyze lag correlations.
  - ▶ **PACF (Partial Autocorrelation Function)** to determine direct lag effects.
- ▶ Patterns guide  $p$  (AR order) and  $q$  (MA order).

## Step 2: Estimation and Selection

**Goal:** Estimate model parameters and select the best-fitting model.

► **Parameter Estimation:**

► **Model Selection:**

► Compare models using:

$$\text{AIC} = -2 \ln(L) + 2k \quad \text{and} \quad \text{BIC} = -2 \ln(L) + k \ln(n)$$

## Step 3: Model Diagnosis

**Goal:** Verify the adequacy of the selected model.

- ▶ **Residual Analysis:**
- ▶ **Statistical Tests:**
- ▶ Refine the model if residuals fail these tests.

## Step 4: Prediction

**Goal:** Use the validated model for forecasting future values.

► **Forecasting:**

$$\hat{Y}_t = \phi_1 Y_{t-1} + \cdots + \theta_1 \epsilon_{t-1} + c$$

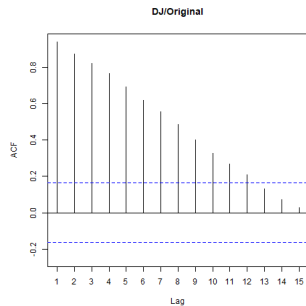
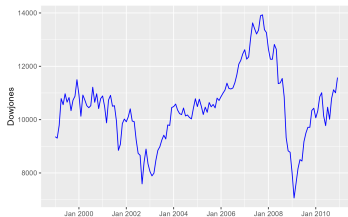
► **Error Metrics:**

- Mean Absolute Error (MAE)
- Root Mean Squared Error (RMSE)
- Mean Absolute Percentage Error (MAPE)

► **Validation:**

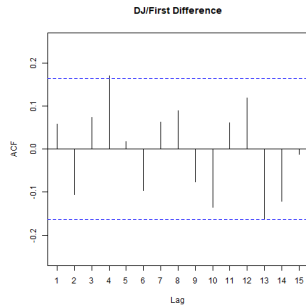
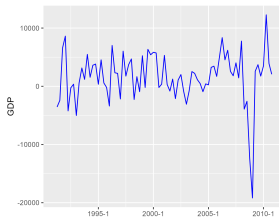
- Compare forecasts to actual values (in-sample or out-of-sample).
- Plot forecasts and prediction intervals for visualization.

# Time series plot Dow Jones / Original / Monthly 1999-2010





# Time series plot Dow Jones / First Difference/ Monthly 1999-2010



## Unit Root Test by Dickey and Fuller I

- ▶ **Unit Root Tests:** These are a set of statistical testing procedures that examine whether the time series data generating process contains a **unit root**, which is indicative of a random walk component (non-stationary behavior).
- ▶ The **Dickey-Fuller (DF) Test:** A prominent test within this group, designed to check for stationarity by testing the presence of a unit root.
- ▶ The **Simple DF Test:**
  - ▶ Null hypothesis ( $H_0$ ): The time series is **not** stationary, implying the presence of a unit root.
  - ▶ Alternative hypothesis ( $H_a$ ): The time series is stationary.

## Unit Root Test by Dickey and Fuller II

Dickey and Fuller propose three estimated equations, where parameters can be determined using **ordinary least squares (OLS)** methods of linear regression:

(i)  $X_t = b_1 X_{t-1} + \varepsilon_t$  (**Test for Random Walk**)

- ▶ This is the simplest test, where the presence of a unit root implies the series follows a pure random walk.

(ii)  $X_t = b_0 + b_1 X_{t-1} + \varepsilon_t$  (**Test for Random Walk with Drift**)

- ▶ Adds a constant ( $b_0$ ) to account for a non-zero mean in the process.

(iii)  $X_t = b_0 + b_1 X_{t-1} + b_2 t + \varepsilon_t$  (**Test for Random Walk with Drift and Deterministic Trend**)

- ▶ Incorporates a deterministic trend ( $b_2 t$ ) to account for a systematic change in the series over time.

## Unit Root Test by Dickey and Fuller III

For case (i), it applies:

$$\Delta X_t = X_t - X_{t-1} = (b_1 - 1)X_{t-1} + \varepsilon_t = \alpha_1 X_{t-1} + \varepsilon_t$$

In general: With  $\alpha_1 := b_1 - 1$ , the three equations can be expressed as:

- (i)  $\Delta X_t = \alpha_1 X_{t-1} + \varepsilon_t$
- (ii)  $\Delta X_t = \alpha_0 + \alpha_1 X_{t-1} + \varepsilon_t$
- (iii)  $\Delta X_t = \alpha_0 + \alpha_1 X_{t-1} + \alpha_2 t + \varepsilon_t$

If the null hypothesis  $H_0 : \alpha_1 = 0$  is **true**, the parameters are **not**  $t$ -distributed. They follow a specific distribution established by Dickey and Fuller.

## Dow Jones – Original

	RW	RW with Drift	RW with Drift/Trend
Constant		611.987* (314.117)	606.858* (315.375)
lag(Dowjones)	0.001 (0.004)	-0.057* (0.029)	-0.059* (0.030)
Trend			0.340 (1.043)
Num.Obs.	143	143	143
R2	0.000	0.029	0.029
R2 Adj.	-0.007	0.022	0.016
AIC	2159.5	2157.2	2159.1
BIC	2165.4	2166.1	2170.9
RMSE	453.85	447.10	446.89

## Dow Jones – First differences

	RW	RW with Drift	RW with Drift/Trend
Constant		15.313 (38.793)	16.492 (87.206)
lag(Dowjones/FD)	−0.941*** (0.092)	−0.941*** (0.093)	−0.941*** (0.093)
Trend			−0.016 (1.073)
Num.Obs.	142	142	142
R2	0.468	0.468	0.468
R2 Adj.	0.464	0.464	0.460
AIC	2144.9	2146.8	2148.8
BIC	2150.8	2155.6	2160.6
RMSE	454.66	454.41	454.41