

STATISTICAL LABORATORY



Applied Mathematics for Economics and Management
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<https://doity.com.br/estatistica-aplicada-a-nutricao>



<https://basiccode.com.br/produto/informatica-basica/>

PROGRAM



1. Fundamental
Concepts of
Statistics



2. Exploratory
Data Analysis



3. Organizing and
Summarizing Data



4. Association and
Relationships
Between Variables



5. Index Numbers



6. Time Series
Analysis

LECTURE 9: LINK INDICES AND CHAIN INDICES

LINK INDEX (YEAR-TO-YEAR INDEX)

Purpose

- To compare **each period with the one immediately before it**.
- This index measures **short-term changes** and avoids having one fixed base year.

Formula:

$$I_{t|t-1} = \frac{p_t}{p_{t-1}} \times 100$$

(or the same using quantities)

Where:

- p_t = price in period t
- p_{t-1} = price in the previous period

Example:

Suppose the price of a product was:

- Year 1: 4.00
- Year 2: 5.00

$$I_{2|1} = \frac{5.00}{4.00} \times 100 = 125$$

→ Prices increased by 25% from Year 1 to Year 2.

CHAIN INDEX (CHAINED INDEX)

Purpose

- To measure the **change from a base year (0) to any later year (t)** by multiplying the **sequence of link indices**.
- Useful when **changing product baskets** over time.

Main Formula:

$$I_{t|0}^{chain} = I_{t|t-1} \times I_{t-1|t-2} \times \dots \times I_{1|0}$$

$$I_{t|0}^{chain} = I_{1|0} \times \frac{I_{2|1}}{100} \times \dots \times \frac{I_{t|t-1}}{100}$$

Alternative Notation for Chain Indices

$$I_{1|0}^{chain} = I_{1|0}$$

$$I_{2|0}^{chain} = I_{2|1} \times I_{1|0}$$

$$I_{3|0}^{chain} = I_{3|2} \times I_{2|1} \times I_{1|0}$$

$\vdots$

$$I_{t|0}^{chain} = I_{t|t-1} \times I_{t-1|t-2} \times \dots \times I_{1|0}$$

Note:

The link (moving-base) indices are not expressed as percentages, but as relative decimals.

Example (continuing previous data):

If:

- $I_{1|0} = 110$
- $I_{2|1} = 125$

Then:

Note:

The link (moving-base) indices are expressed as percentages.

$$I_{2|0}^{chain} = 110 \times \frac{125}{100} = 137.5$$

→ From the base year (0) to Year 2, prices increased by 37.5%.

FIXED-BASE AND MOVING-BASE INDICES

Fixed-base Index

- Compares **each period to a single base period** (usually Year 0).
- Used to measure **long-term trends** relative to a constant reference.

$$I_{t|0}^{fixed} = \frac{x_t}{x_0} \times 100$$

Moving-base Index

- Compares **each period to a changing base**.
- Useful for analysing **short-term fluctuations**.

$$I_{t|t-1}^{moving} = \frac{x_t}{x_{t-1}} \times 100$$

Types of Moving-base Indices

Type	Definition	Purpose
Link Index (<i>Índice Elos</i>)	The base is the immediately previous period ($t - 1$).	Captures period-to-period variation .
Homologous Index (<i>Índice Homólogo</i>)	The base is the same period in the previous year ($t - 12$).	Used to control for seasonal patterns .

CHAINED INDEX: FIXED-BASE OR MOVING-BASE?

In **Silvestre (2007)**, the distinction is made as follows:

- **Chained indices** are obtained from **link relatives** (i.e., **moving-base indices**, where each period is compared to the previous one).
- Then, these link indices are **multiplied in sequence** to show the accumulated change since a fixed initial period.
- Therefore, **even though the calculation uses moving-base indices**, the final result is expressed **relative to a fixed initial period**. So, a **chained index** is considered a **fixed-base index**, because it shows the evolution from the initial period t_0 .

Index type	Comparison	Note
Fixed base	Always with t_0	e.g., Laspeyres, Paasche
Moving base (link relatives)	Each period with $t - 1$	e.g., link index
Chained index	Calculated from link relatives	Final result relative to $t_0 \rightarrow$ considered fixed-base

LECTURE 9: BASE CHANGE AND RECONCILIATION

CHANGING THE BASE OF AN INDEX

$$i_{t|0} = \frac{x_t}{x_0} = \frac{x_t}{x_0} \cdot \frac{x_b}{x_b} = \frac{x_t}{x_b} \cdot \frac{x_b}{x_0} = i_{t|b} \cdot i_{b|0}$$

Explanation (Notes):

- $i_{t|0}$ → index of period t with original base 0
- $i_{t|b}$ → index of period t with new base b
- $i_{b|0}$ → index of period b with original base 0
- This formula shows that the index with the original base can be expressed as the product of the index with the new base and the index of the new base relative to the original base.


$$i_{t|0} = \frac{i_{t|b}}{i_{b|0}} \cdot 100$$

Usage:

- Allows rescaling of index series without recalculating all values from scratch.

CHANGING THE BASE OF AN INDEX: EXAMPLE

Example: Suppose we have a price series:

Period (t)	Price x_t	Index $x_{t 0} = (x_t / x_0) \times 100$
0	100	100
1	120	120
2	150	150

We want to **change the base** from period 0 to period 1 ($b = 1$).

Compute $i_{t|b}$ using the formula:

$$i_{t|b} = \frac{i_{t|0}}{i_{b|0}} \cdot 100$$

- $i_{1|1} = \frac{120}{120} \cdot 100 = 100$
- $i_{2|1} = \frac{150}{120} \cdot 100 = 125$

Interpretation:

- Now, period 1 becomes the base (100).
- Period 2 shows a 25% increase relative to period 1.

CHANGING THE BASE OF AN INDEX: EXAMPLE

Example: Suppose we have a price series:

Period (t)	Price x_t	Index $x_{t 0} = (x_t / x_0) \times 100$ (Base 0)	Index $x_{t 1} = (x_t / x_1) \times 100$ (Base 1)
0	100	100	83.33
1	120	120	100
2	150	150	125

- $i_{t|0}$ = Index with original base (period 0 = 100)
- $i_{t|1}$ = Index with new base (period 1 = 100)
- Period 0 relative to new base 1: $i_{0|1} = \frac{i_{0|0}}{i_{1|0}} \cdot 100 = 83.33$
- Period 2 relative to new base 1: $i_{2|1} = \frac{i_{2|0}}{i_{1|0}} \cdot 100 = 125$

Interpretation:

- Now all periods are measured relative to period 1. Period 0 is lower than 100, indicating it was below the new base.

RECONCILIATION OF INDICES: BASE UNIFORMIZATION

Base uniformization consists in expressing all index series on the **same reference base**, so that they can be compared.

This may occur in two situations:

- 1. The indices already share the same base**, but we want to change to a *new* common base.
- 2. The indices have different bases**, so the first step is to convert them to a *common* base.

RECONCILIATION OF INDICES: BASE UNIFORMIZATION

We consider two index series:

- Index A defined for periods $0, 1, \dots, b$, with base 0
- Index B defined for periods $b, b + 1, \dots, T$, with base b

The goal is to obtain **one single index series** over the whole period.

Index A	$I_{0 0}^A$	$I_{1 0}^A$	$I_{2 0}^A$	...	$I_{b 0}^A$				
Index B					$I_{b b}^B$	$I_{b+1 b}^B$	$I_{b+2 b}^B$	...	$I_{T b}^B$

Situation 1 — Final Series Expressed in Base b

$$I_{t|b}^C = \begin{cases} \frac{I_{t|0}^A}{I_{b|0}^A} \times 100, & t = 0, 1, \dots, b \\ I_{t|b}^B, & t = b + 1, \dots, T \end{cases}$$

- Interpretation:

Convert **Index A** to base b; keep **Index B** as it already uses base b.

Important Note on “×100” and “÷100”

If indices are recorded as **index numbers**, where the base = 100,
→ **Keep the “×100” and “÷100”** in the formulas.

If indices are recorded as **simple ratios** (base = 1, not 100),
→ **Remove the “×100” and “÷100”** from both formulas.

RECONCILIATION OF INDICES: BASE UNIFORMIZATION

We consider two index series:

- Index A defined for periods $0, 1, \dots, b$, with **base 0**
- Index B defined for periods $b, b + 1, \dots, T$, with **base b**

The goal is to obtain **one single index series** over the whole period.

Index A	$I_{0 0}^A$	$I_{1 0}^A$	$I_{2 0}^A$	...	$I_{b 0}^A$				
Index B					$I_{b b}^B$	$I_{b+1 b}^B$	$I_{b+2 b}^B$	...	$I_{T b}^B$

Situation 2 — Final Series Expressed in Base 0

$$I_{t|0}^C = \begin{cases} I_{t|0}^A, & t = 0, 1, \dots, b \\ I_{t|b}^B \cdot \frac{I_{b|0}^A}{100}, & t = b + 1, \dots, T \end{cases}$$

- Interpretation:

Keep Index A as it is; convert Index B from base b to base 0.

Important Note on “×100” and “÷100”

If indices are recorded as **index numbers**, where the base = 100,
→ **Keep the “×100” and “÷100”** in the formulas.

If indices are recorded as **simple ratios** (base = 1, not 100),
→ **Remove the “×100” and “÷100”** from both formulas.

WHY DIVISION IN CASE 1 AND MULTIPLICATION IN CASE 2?

Case 1 — Converting from Base 0 to Base b (Division)

$$I_{t|b}^A = \frac{I_{t|0}^A}{I_{b|0}^A} \times 100$$

We **divide** because we need to **remove** the cumulative change from period 0 to period b.

Interpretation:

"To express period t relative to b, we cancel the growth up to b."

→ Changing to a *later* base (0 → b) requires dividing.

Case 2 — Converting from Base b to Base 0 (Multiplication)

$$I_{t|0}^B = I_{t|b}^B \times \frac{I_{b|0}^A}{100}$$

We **multiply** because we need to **add** the cumulative change from period 0 to period b.

Interpretation:

"To express period t relative to 0, we combine the change from 0→b and b→t."

→ Changing to an *earlier* base (b → 0) requires multiplying.

WHY DIVISION IN CASE 1 AND MULTIPLICATION IN CASE 2?

Direction of Base Change	Operation	Why	
From an older base to a more recent base	Divide	Removes the cumulative variation up to b	
From a more recent base to an older base	Multiply	Adds the cumulative variation up to b	

RECONCILIATION OF INDICES: BASE UNIFORMIZATION - EXAMPLE

Suppose we have two index series:

- Index A (base year = 0):

$$I_{0|0}^A = 100, \quad I_{1|0}^A = 110, \quad I_{2|0}^A = 120, \quad I_{3|0}^A = 150$$

- Index B (base year = 3):

$$I_{3|3}^B = 100, \quad I_{4|3}^B = 105, \quad I_{5|3}^B = 115$$

We want a single reconciled series over $t = 0, \dots, 5$.

RECONCILIATION OF INDICES: BASE UNIFORMIZATION - EXAMPLE

Case A — Final series expressed with base = 3

Formula

$$I_{t|3}^C = \begin{cases} \frac{I_{t|0}^A}{I_{3|0}^A} \times 100, & t = 0, 1, 2, 3 \\ I_{t|3}^B, & t = 4, 5 \end{cases}$$

To change to a later base (older → later): divide by the index at the new base (and ×100 if indices use base-100).

Reason: converting from an *older* base (0) to a *later* base (3) requires **dividing** by the index at the new base (and re-scaling to 100).

Reason: converting from an *older* base (0) to a *later* base (3) requires **dividing** by the index at the new base (and re-scaling to 100).

Calculation (since $I_{3|0}^A = 150$)

- $I_{0|3}^C = 100/150 \times 100 = 66.67$
- $I_{1|3}^C = 110/150 \times 100 = 73.33$
- $I_{2|3}^C = 120/150 \times 100 = 80.00$
- $I_{3|3}^C = 150/150 \times 100 = 100.00$
- $I_{4|3}^C = 105$ (from B)
- $I_{5|3}^C = 115$ (from B)

Year(t)	0	1	2	3	4	5
$I_{t 3}^C$	66.67	73.33	80	100	105	115

RECONCILIATION OF INDICES: BASE UNIFORMIZATION - EXAMPLE

Case B — Final series expressed with base = 0

Formula

$$I_{t|0}^C = \begin{cases} I_{t|0}^A, & t = 0, 1, 2, 3 \\ I_{t|b}^B \times \frac{I_{b|0}^A}{100}, & t = 4, 5 \end{cases}$$

where $b = 3$.

Reason: converting from a *later* base (3) back to an *earlier* base (0) requires **multiplying** by the base index $I_{3|0}^A$ (and adjusting by 100).

To change to an earlier base (later → older): multiply the later-base index by the index of the reference period expressed in the older base (and ÷ 100 if indices use base-100).

Calculation (since $I_{3|0}^A = 150$)

- For $t = 0, 1, 2, 3$ keep A: 100, 110, 120, 150
- $I_{4|0}^C = 105 \times 150/100 = 157.50$
- $I_{5|0}^C = 115 \times 150/100 = 172.50$

Year(t)	0	1	2	3	4	5
$I_{t 3}^C$	100	110	120	150	157.50	172.50

LECTURE 9: TIMES SERIES

TIMES SERIES: INTRODUCTION

Data observed over time

- measurements or values recorded at successive time points.

Observations are collected in time order

- the sequence matters; order affects analysis.

Useful for studying trends, cycles, and temporal patterns

- e.g., stock prices, temperature, monthly sales.

Graphical representation

- often visualized in a **time plot (chronogram)** to detect patterns.

Different from cross-sectional data

- cross-sectional data are collected **at a single point in time** for different individuals, firms, or regions, while chronological data track **one variable over time**.

Note:

- **Time is the essential factor** in chronological data, unlike cross-sectional data where time does not vary.

OBJECTIVES OF STUDYING TIME SERIES

► Description

- To summarize and illustrate the behavior of a variable over time.

► Explanation

- To identify the causes behind observed patterns and fluctuations.

► Forecasting

- To predict future values based on past trends.

► Control

- To monitor and manage processes by detecting deviations from expected behavior.

DEFINITION OF A TIME SERIES

- Let x be a variable observed over time.
- Suppose we collect n observations:

$$x_1, x_2, x_3, \dots, x_n$$

- Abbreviated notation: $\{x_t\}_{t=1}^n$
- Graphical representation: displayed in a time plot (chronogram)

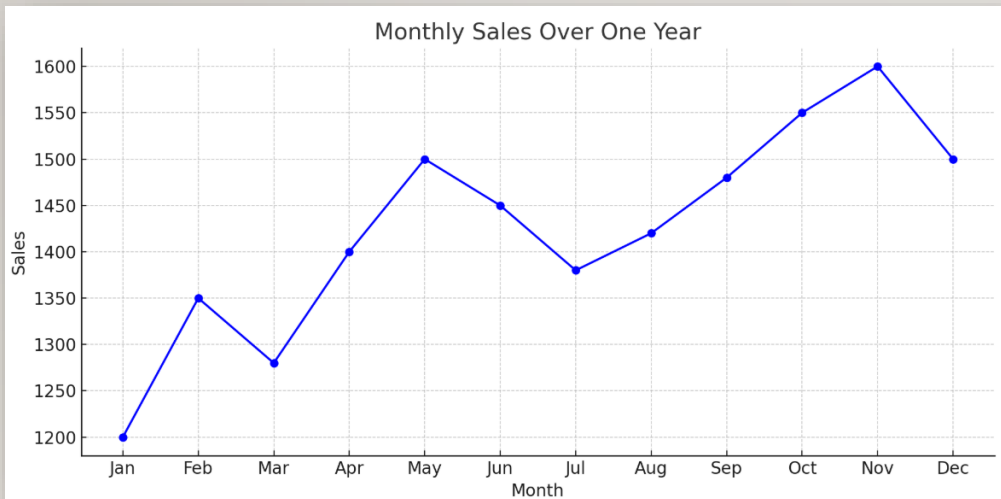
TIME SERIES: EXAMPLE

Example:

- Monthly sales of a company over one year:

$$x_1 = 1200, x_2 = 1350, x_3 = 1280, \dots, x_{12} = 1500$$

Chronogram



Interpretation:

- Sales **increased** from January to February, then **decreased** from February to March.
- After that, sales **rose steadily** until May, followed by a **decline** until June.
- From June onwards, sales **increased again**, reaching the **highest value** in November.
- Finally, sales **declined** in December.

Summary: The monthly sales show a **fluctuating trend** with **two main growth periods** and a **peak in November**, indicating possible **seasonal effects** or **promotional impacts**.

COMPONENTS AND MODELS OF A TIME SERIES

- Additive Model:

$$x_t = T_t + s_t + c_t + e_t$$

- Multiplicative Model:

$$x_t = T_t \cdot s_t \cdot c_t \cdot e_t$$

- Mixed Models (Silvestre):

1. Type 1:

$$x_t = (T_t + c_t) \cdot s_t + e_t$$

2. Type 2:

$$x_t = T_t \cdot c_t \cdot s_t + e_t$$

Components:

- Trend (T_t)
- Seasonality (s_t)
- Cyclical or Oscillatory Movements (c_t)
- Residual or Irregular Component (e_t)

Notes:

•Trend (T_t):

The long-term movement or general direction of the time series over a period of time. Shows whether the series is increasing, decreasing, or stable.

•Seasonality (s_t):

Regular, repeating patterns within a fixed period, such as months, quarters, or days. Often linked to calendar or weather effects.

•Cyclical or Oscillatory Movements (c_t):

Medium- to long-term fluctuations that occur irregularly, often related to economic or business cycles.

•Residual or Irregular Component (e_t):

Random or unpredictable variations that are not explained by trend, seasonality, or cyclical movements.

COMPONENTS AND MODELS OF A TIME SERIES

Components:

- Trend (T_t)
- Seasonality (s_t)
- Cyclical or Oscillatory Movements (c_t)
- Residual or Irregular Component (e_t)

- **Additive Model:**

Assumes that components add together to form the series: $x_t = T_t + s_t + c_t + e_t$

- **Multiplicative Model:**

Assumes that components multiply together: $x_t = T_t \cdot s_t \cdot c_t \cdot e_t$

- **Mixed Models (Silvestre):**

1. **Type 1:** $x_t = (T_t + c_t) \cdot s_t + e_t$ → trend and cycle add, then multiplied by seasonality, plus residual
2. **Type 2:** $x_t = T_t \cdot c_t \cdot s_t + e_t$ → trend, cycle, and seasonality multiply, plus residual

COMPONENTS OF A TIME SERIES: TREND

A **time series** is a variable observed over time (monthly, yearly, daily, etc.).

Example: *Monthly sales of a store.*

A time series can be decomposed into **four main components**:

1) Trend (T_t)

The long-term movement of the series.

Key question:

"Is the series **generally** increasing, decreasing, or remaining stable over time?"

- If sales grow year after year → **Upward trend**
- If they decrease → **Downward trend**

Example:

Year 1: 100 → Year 2: 120 → Year 3: 145 → Year 4: 160

Overall direction = **increasing** → upward trend.

COMPONENTS OF A TIME SERIES: SEASONALITY

2) Seasonality (S_t)

A regular and repeating pattern within a fixed period (usually within each year).

Key question:

“Does the series show a pattern that repeats at the **same times** each year?”

Examples:

- Higher sales every **Christmas**
- More restaurant demand **every summer**
- Higher electricity usage **every winter**

Mental image:

Small repeating waves **always occurring in the same months.**

COMPONENTS OF A TIME SERIES: CYCLE OR OSCILLATORY MOVEMENTS

3) Cycle (C_t)

Medium-to-long-term fluctuations that do not follow a fixed calendar pattern.

Key question:

“Are there expansions and contractions that happen, but not at regular intervals?”

Examples:

- Economic expansion → recession → recovery → expansion
- Business cycles that last several years

Difference between Seasonality and Cycle

Seasonality

Repeats on a **fixed schedule** (e.g., every December)

Short period (within the year)

Cycle

No fixed period; timing varies

Long period (several years)

COMPONENTS OF A TIME SERIES: IRREGULAR OR RESIDUAL COMPONENT

4) Irregular or Residual Component (E_t)

The **unpredictable** part of the series; random variation.

Includes:

- Unexpected events (strikes, storms, pandemics)
- Measurement errors
- Accidents and non-systematic shocks

Mental image:

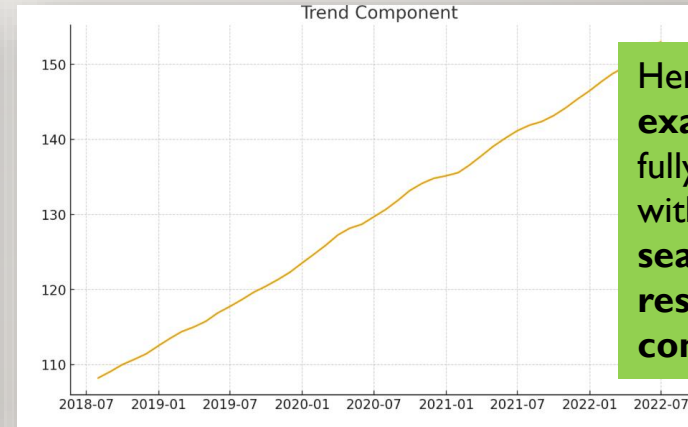
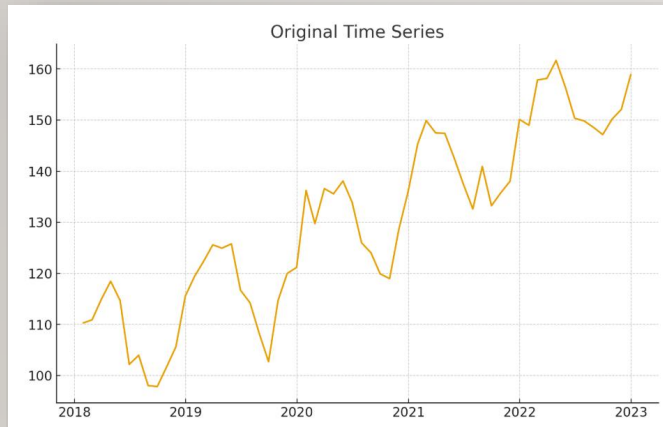
Noise that cannot be explained by Trend, Seasonality, or Cycle.

COMPONENTS OF A TIME SERIES

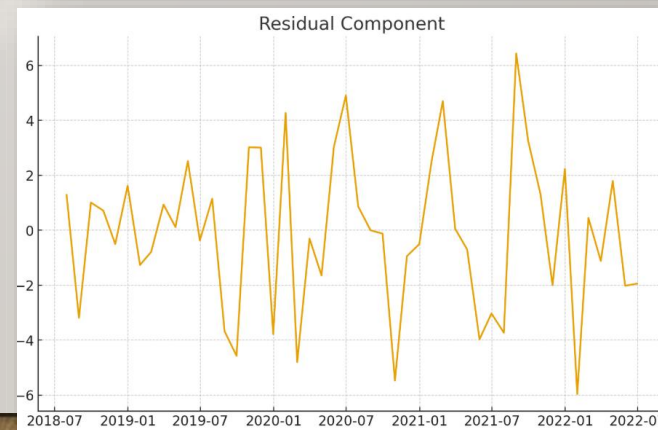
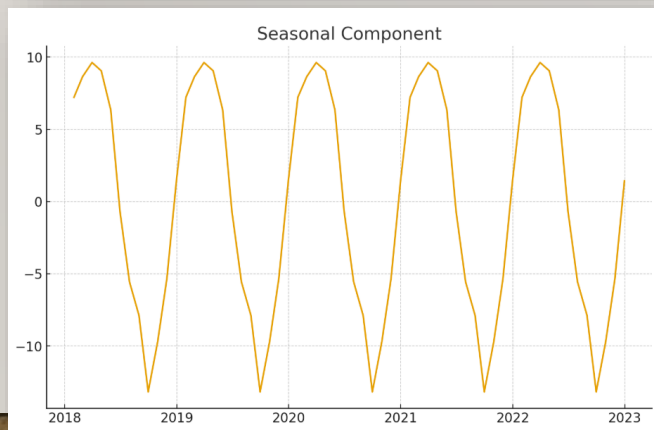
Summary Table

Component	Meaning	Example	
Trend (T_t)	Long-term direction	Sales gradually increase over years	
Seasonality (S_t)	Repeating pattern within each year	Christmas demand spike every year	
Cycle (C_t)	Long-term economic fluctuations	Periods of recession and recovery	
Residual (E_t)	Unpredictable/random variation	Strikes, pandemic, shocks	

COMPONENTS AND MODELS OF A TIME SERIES: EXAMPLE



Here is a **real** example, already fully decomposed with **trend**, **seasonality**, and **residual** components.



COMPONENTS AND MODELS OF A TIME SERIES: EXAMPLE

Original Series

This shows the actual observed values over time.

We can already see:

- A general **upward trend**
- A **repeating seasonal pattern** (waves)

Trend Component (T_t)

A **smooth long-term movement**.

→ Here, the trend steadily increases from 100 to about 160.

Seasonal Component (S_t)

A **repeating pattern** within each year.

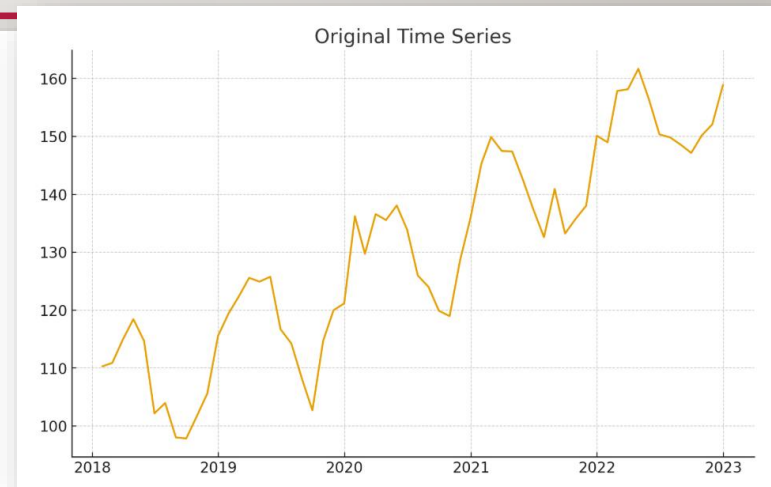
→ Notice that the pattern repeats every **12 months**, always with the same shape

Residual Component (E_t)

This is what remains after removing **Trend** and **Seasonality**:

$$E_t = X_t - T_t - S_t$$

→ It represents random shocks and unpredictable variation.



We generated a monthly time series with:

- an **upward trend**
- a **yearly seasonal pattern**
- some **random noise**

and then applied **seasonal decomposition** (additive model).

Note:

In this series, no cyclical component is considered, since economic cycles require long time spans and do not follow a fixed periodic pattern. Therefore, the decomposition is simplified to: $X_t = T_t + S_t + E_t$

IDENTIFYING THE APPROPRIATE TIME SERIES MODEL

Feature Observed in the Time Plot	Type of Model	Interpretation
The seasonal fluctuations have constant amplitude over time	Additive Model $X_t = T_t + S_t + C_t + E_t$	Seasonal variations do not depend on the level of the series.
The seasonal fluctuations increase or decrease proportionally to the level of the series	Multiplicative Model $X_t = T_t \cdot S_t \cdot C_t \cdot E_t$	Seasonal variations are relative to the level of the series.
The seasonal fluctuations increase slightly, but not fully proportionally to the level	Mixed Model $X_t = (T_t + C_t) \cdot S_t + E_t$	Part of the variation is constant, and part depends on the level of the series.

Key Guiding Rule

- If the **seasonal amplitude** stays constant → *Additive*
- If the **seasonal amplitude** grows with the trend → *Multiplicative*
- If the **amplitude grows slightly but not proportionally** → *Mixed*

THANKS!

Questions?