

# Time Series Fundamentals

## A Guided-Discovery Adventure

Date: \_\_\_\_\_ Name: \_\_\_\_\_ Roll No: \_\_\_\_\_

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### 1. The First Mystery: When Does a Time Series Behave?

Look at this plot (Mystery Plot #1):

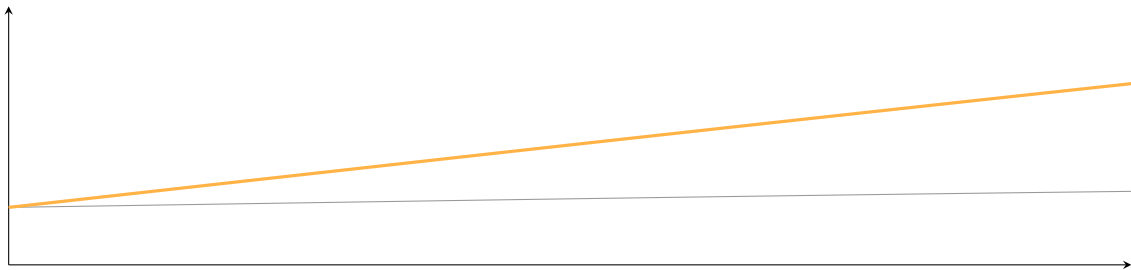


Figure 1: Something is changing... but what?

**Q1.** Stare at the graph. Where does the “average height” of the series seem to drift? Why might this drift make forecasting tricky?

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A model expects the data’s “personality” (mean, variance) to stay stable. When it shifts over time, the model keeps chasing a moving target. This stability is called **stationarity**.

### 2. The Difference Trick: Calming the Chaos

Here’s the same series... and its first difference (Mystery Plot #2):

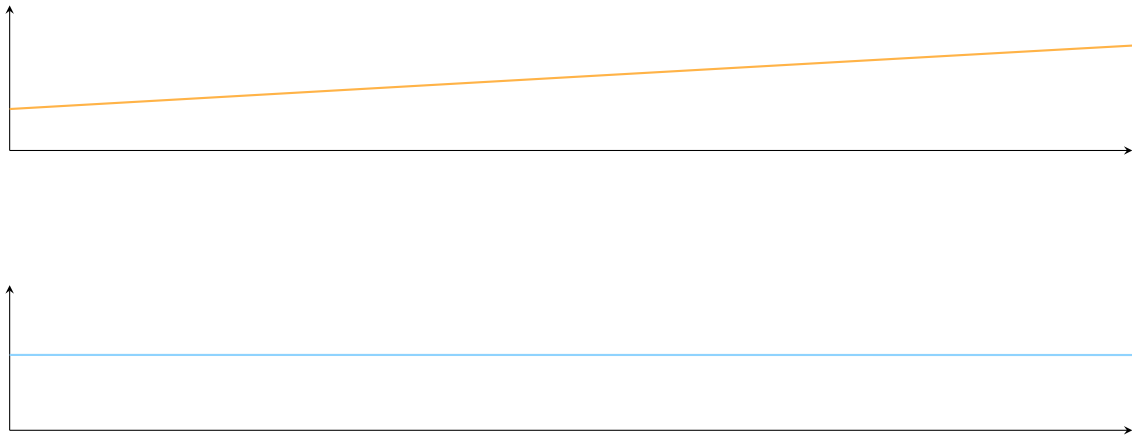


Figure 2: Top: drifting series. Bottom: its daily changes — surprisingly stable!

**Q2.** Why does the bottom plot “calm down”? What part of the original series disappeared after differencing?

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Differencing removes slow movement (trend), leaving behind a stable, wigglier-but-predictable structure.

### 3. Why Logs Matter: When Big Numbers Swing Too Hard

**Mystery Plot #3:** A growing series and its log-transform

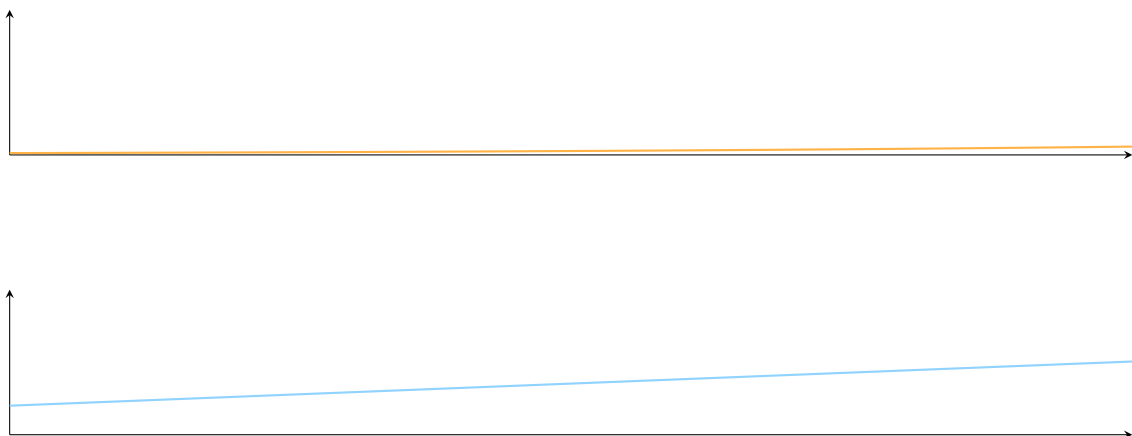


Figure 3: Log-transform “shrinks” big swings into manageable ones.

**Q3.** Why might taking logs help stabilize the “wildness” of the upper plot?

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Logs compress large numbers more than small ones  $\rightarrow$  variance becomes steadier.

## 4. ACF: How Far Back Does Memory Go?

### Mystery Plot #4: The ACF bars

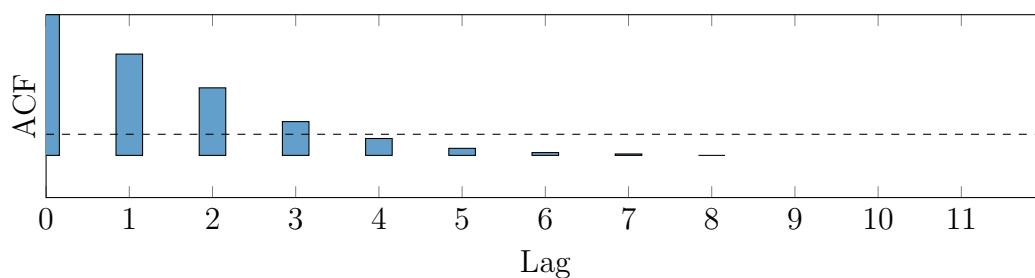


Figure 4: Bars show similarity with the past. How long does memory last here?

**Q4.** Which lag seems most important? Does memory fade quickly or slowly?

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ACF shows *total* correlation with past values, including indirect influence.

## 5. PACF: Direct Influence or Indirect Echo?

### Mystery Plot #5: The PACF

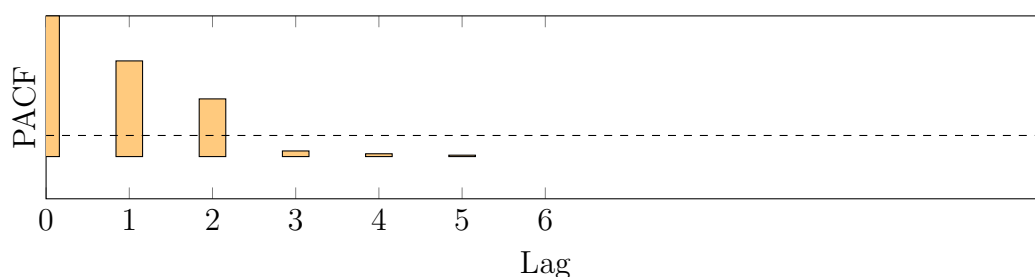


Figure 5: Which lags directly influence the present?

**Q5.** If lag 2 still matters *after removing* lag 1's effect, what does that say about the series structure?

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PACF reveals *direct* effects, helping identify AR order.

## 6. AR or MA? Your Time Series Detective Rulebook

Observation	Likely Model
ACF cuts off after lag $q$	MA( $q$ )
PACF cuts off after lag $p$	AR( $p$ )
Both decay slowly	Non-stationary (difference!)

**Q6.** If ACF spikes at lags 1–3 then vanishes, which model whispers its name?

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**Q7.** If PACF yells only at lag 1 then goes silent, what model might fit?

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## 7. Mini Puzzles: Solve the Mystery!

**Puzzle A.** The series looks like a person climbing stairs while wobbling. Which transformation fixes it?

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**Puzzle B.** ACF shows a perfect spike at lag 3. What is it trying to tell you?

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**Puzzle C.** PACF spikes at lag 2 only. Predict the AR order.

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## 8. Mini Case Study: Food Delivery Orders

A city's daily food orders over 18 months show an upward trend, strong weekly patterns, and high dependence on the previous day.

### Q1. Stationarity check

What problem is present in the data? Name one simple transformation to fix it.

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### Q2. Memory in data

If correlation with past days fades gradually, what does this say about how far back the model should look?

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### Q3. Seasonality link

At which time gaps would you expect repeated correlation peaks, and what part of the data causes them?

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