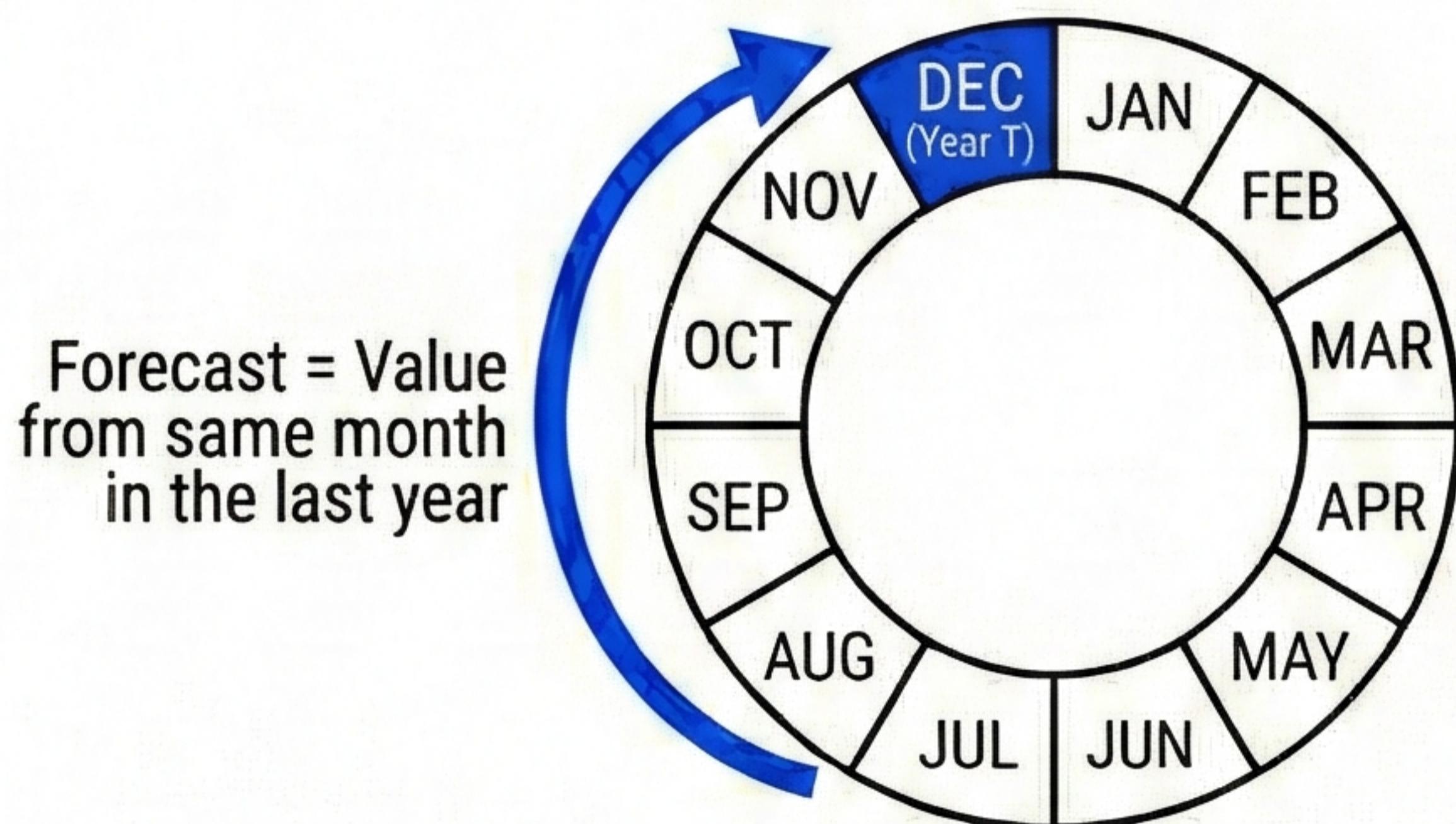


05 EN Predictive Statistics Seasonal Naive model in monthly time series forecasting	2
06 EN Predictive Statistics Pearson correlati- on used for seasonality detection in time series forecasting	3
07 EN Predictive Statistics Coefficient of Det- ermination in time series forecasting	4
08 EN Predictive Statistics Coefficient of Det- ermination for seasonality detection	5

SEASONAL NAIVE MODEL: MONTHLY TIME SERIES FORECASTING

CORE PRINCIPLE: CYCLIC REPETITION

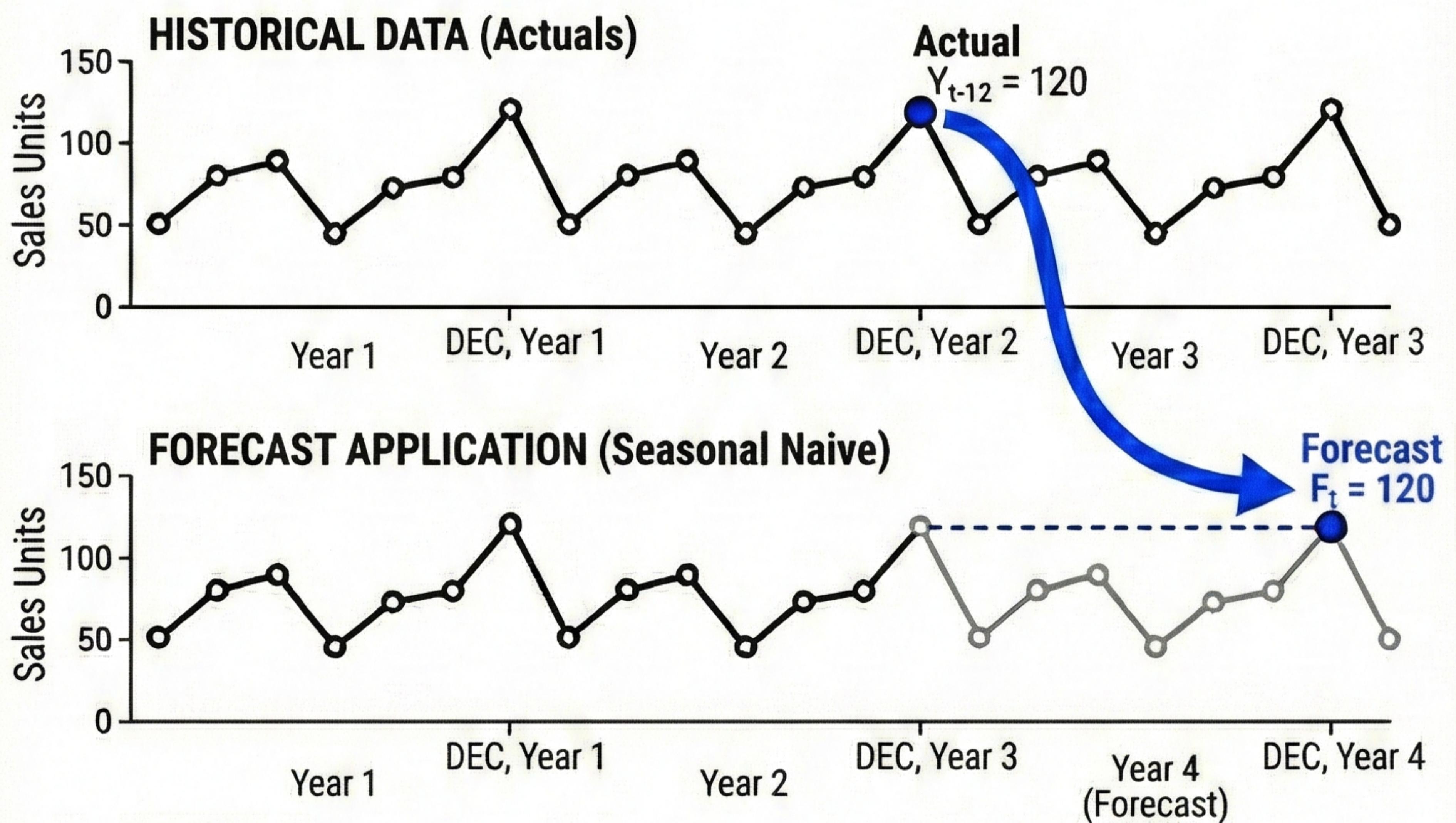


MATHEMATICAL RULE

$$F_t = Y_{t-m}$$

Forecast at time t Actual value at time (t-m) Seasonal period (e.g., m=12 for monthly)

WORKED EXAMPLE: MONTHLY SALES DATA

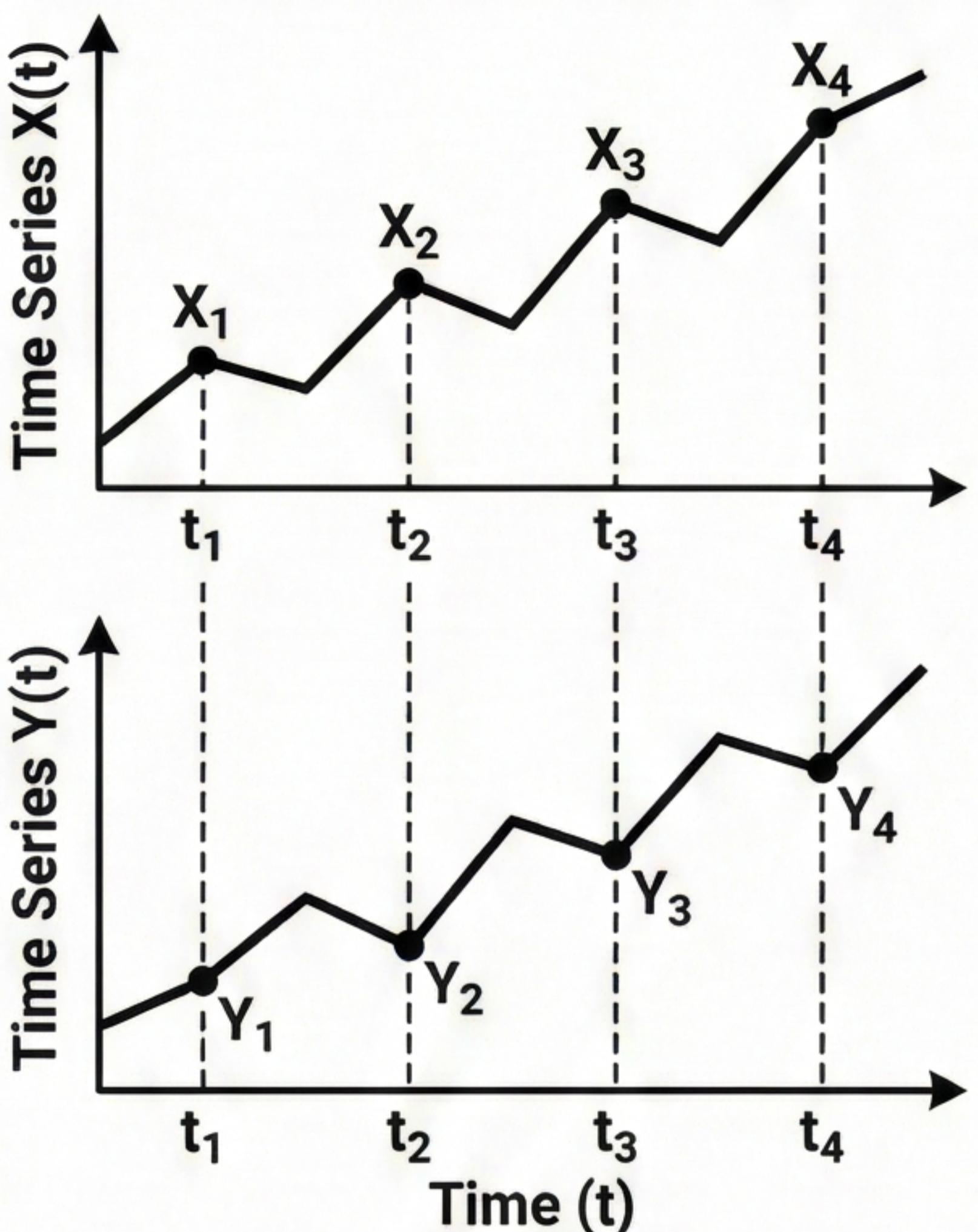


KEY CHARACTERISTICS & LIMITATIONS

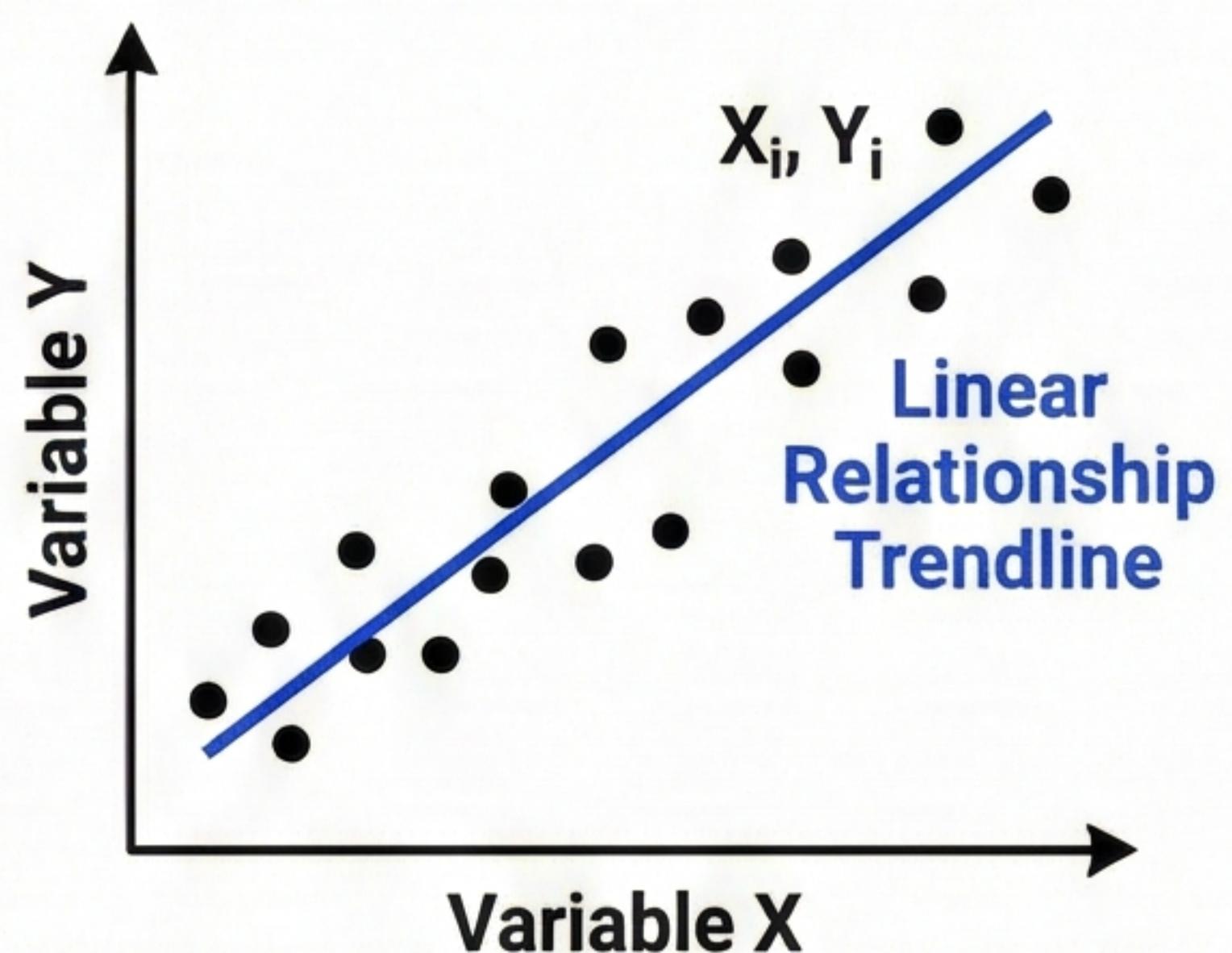
- Simple baseline method
- Does not require parameter estimation
- Assumes perfect seasonal repetition
- Ignores trend changes and irregular fluctuations (Limitation)
- Lagged by one full season (Limitation)

PEARSON CORRELATION FOR TIME SERIES FORECASTING

1. TIME SERIES DATA (INPUT)



2. PEARSON CORRELATION (r)



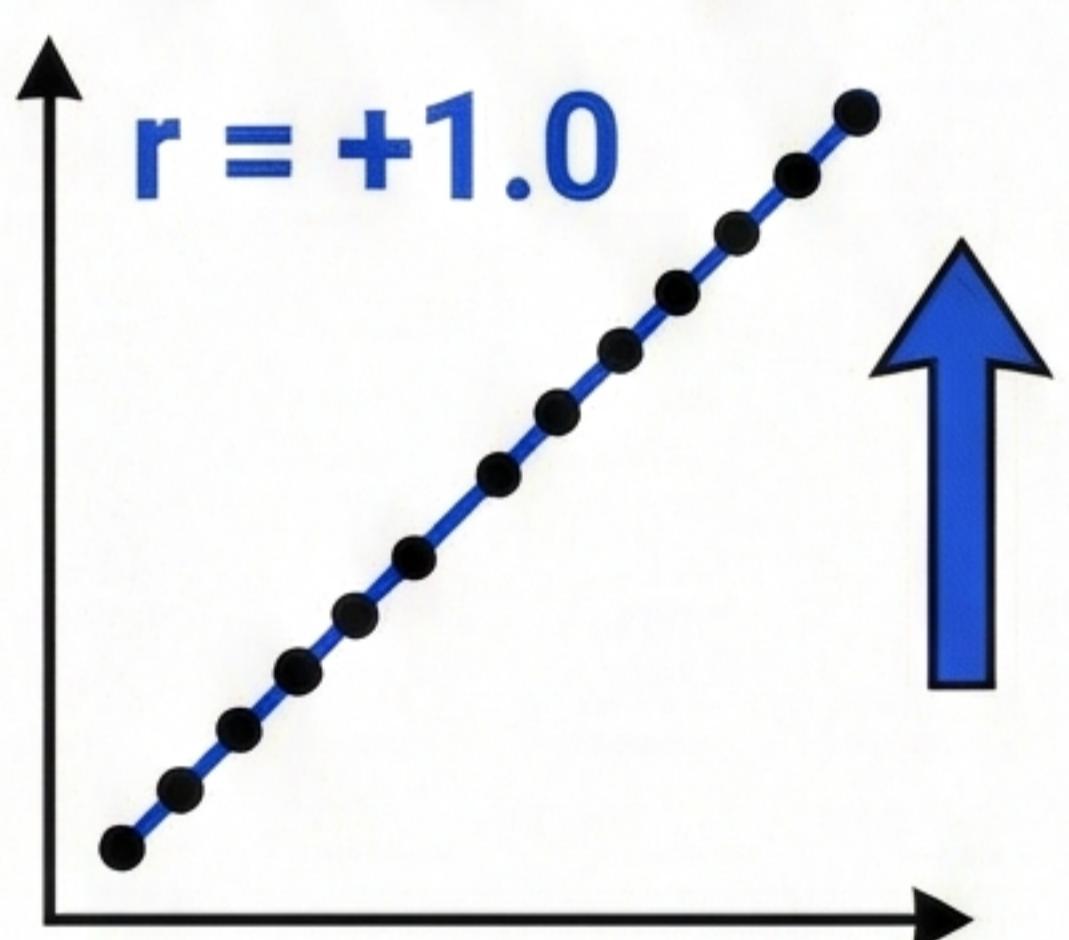
Pearson's r measures the strength and direction of a linear association between two continuous variables.

$$r = \frac{\sum[(X_i - \bar{X})(Y_i - \bar{Y})]}{\sqrt{\sum(X_i - \bar{X})^2} \cdot \sqrt{\sum(Y_i - \bar{Y})^2}}$$

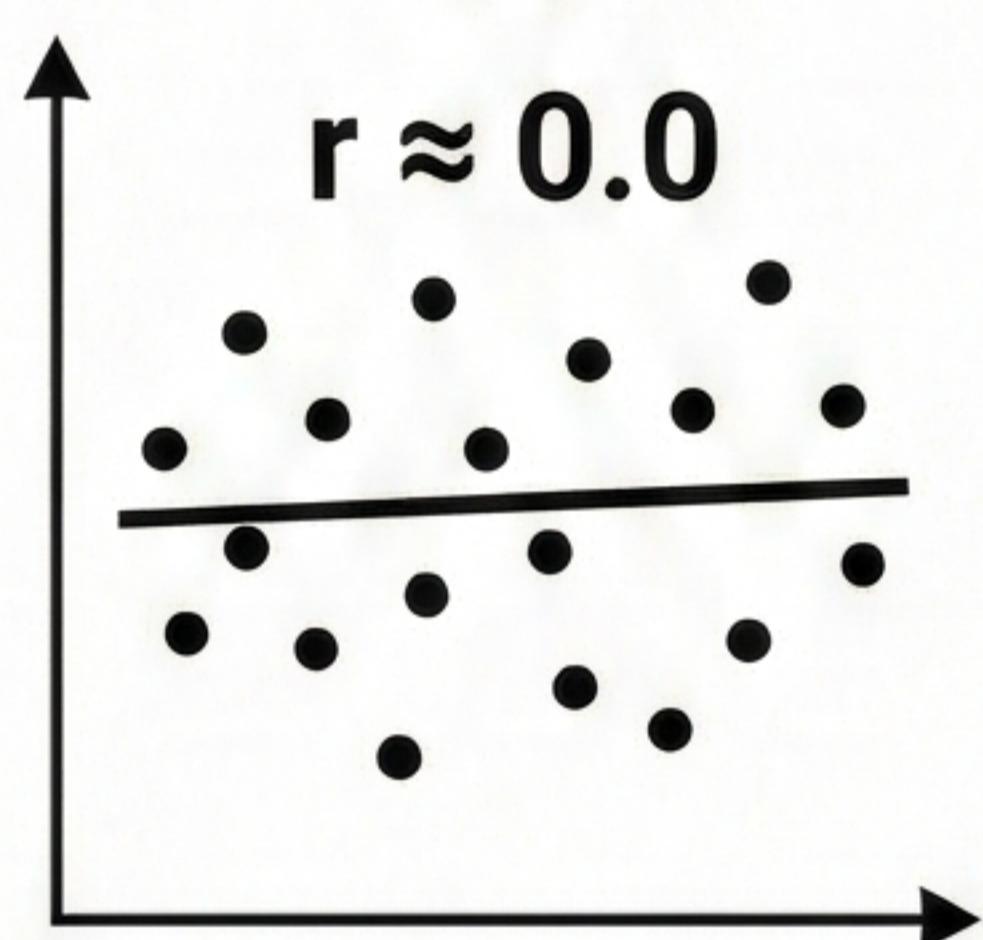
Calculate r

3. CORRELATION STRENGTH & DIRECTION

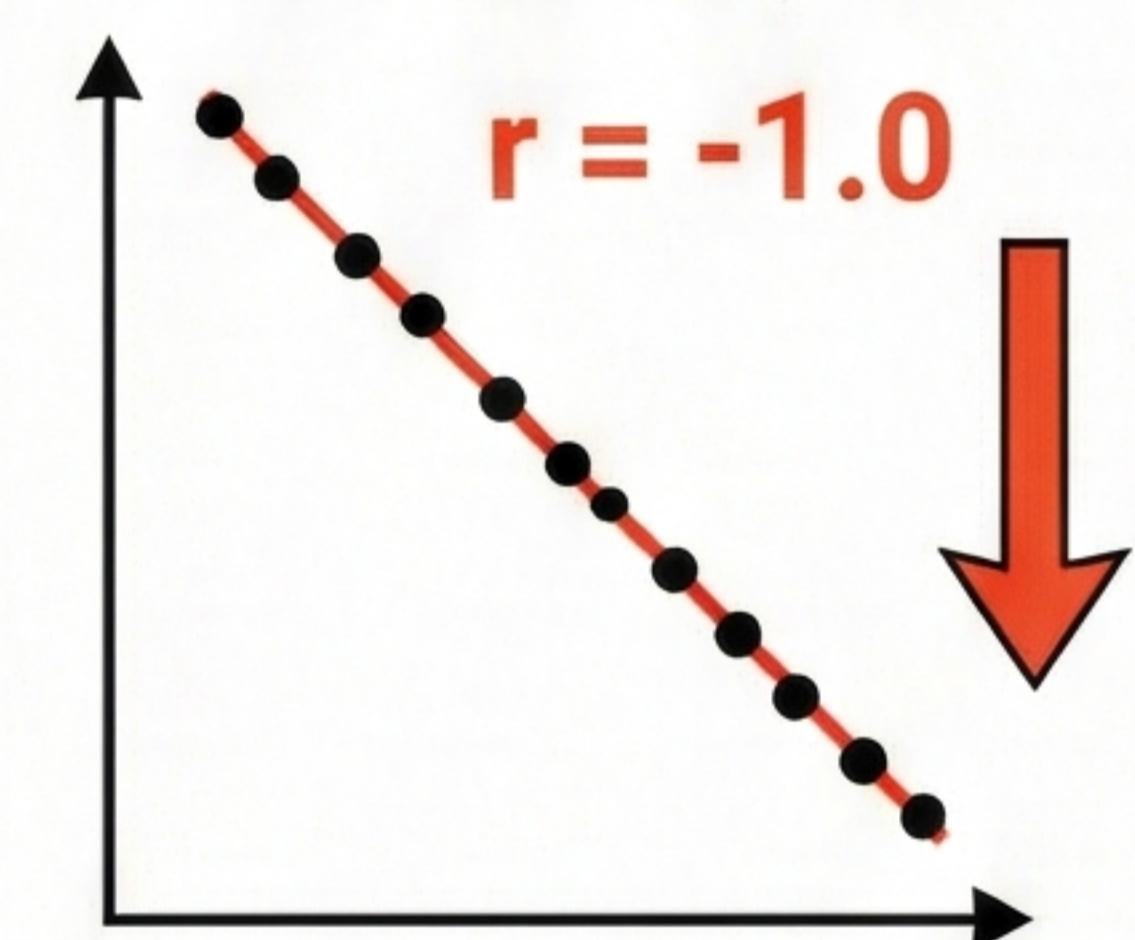
Perfect Positive Correlation (+1)



No Correlation (0)



Perfect Negative Correlation (-1)



Strong Negative

Weak/None

Strong Positive

-1

0

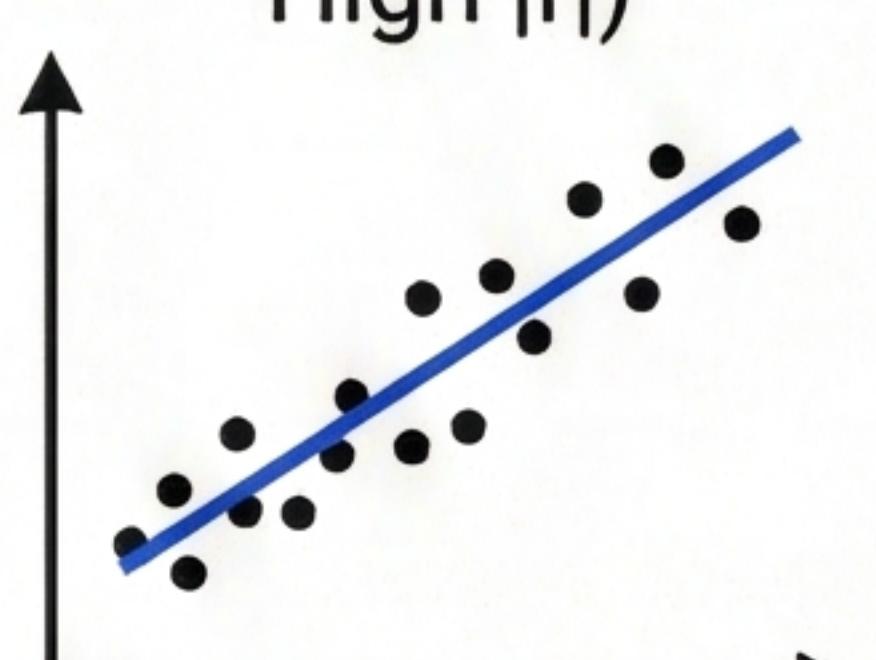
+1



Interpret & Select

4. FORECASTING APPLICATION

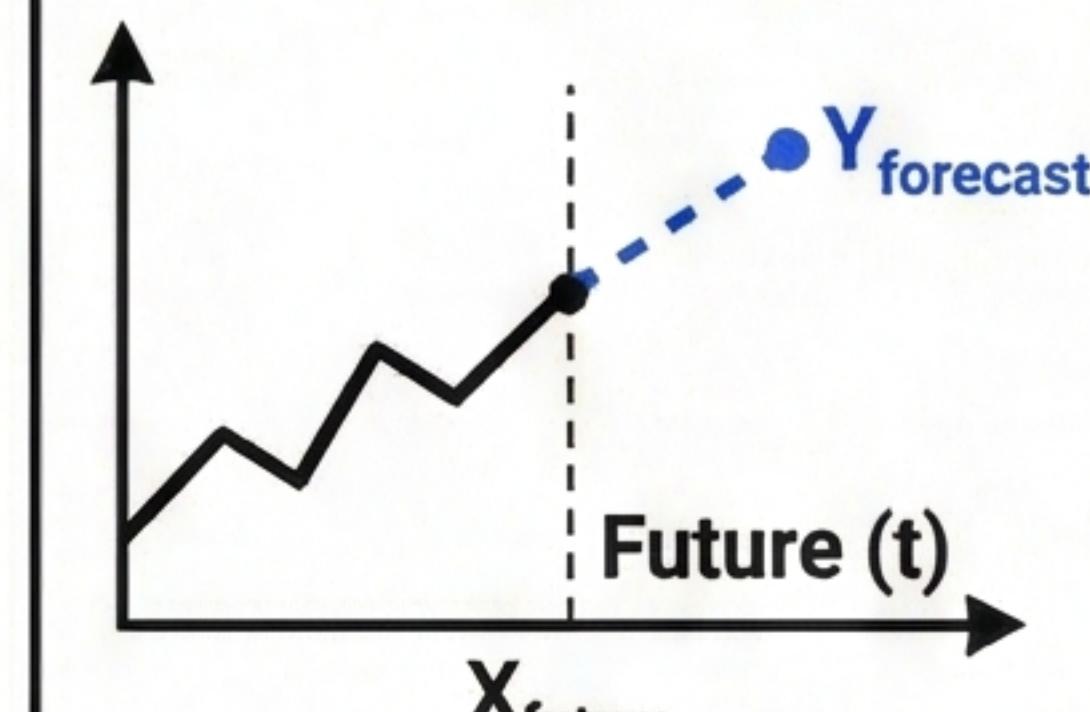
Historical Data
(Strong Correlation,
High $|r|$)



Forecast Future X
(X_{future})

Assumption: Historical linear relationship continues

Predict Future Y
(Y_{forecast})



COEFFICIENT OF DETERMINATION (R^2) IN TIME SERIES FORECASTING

Quantifying the Proportion of Variance Explained by a Forecasting Model

MODULE 1: CORE CONCEPT & DEFINITION

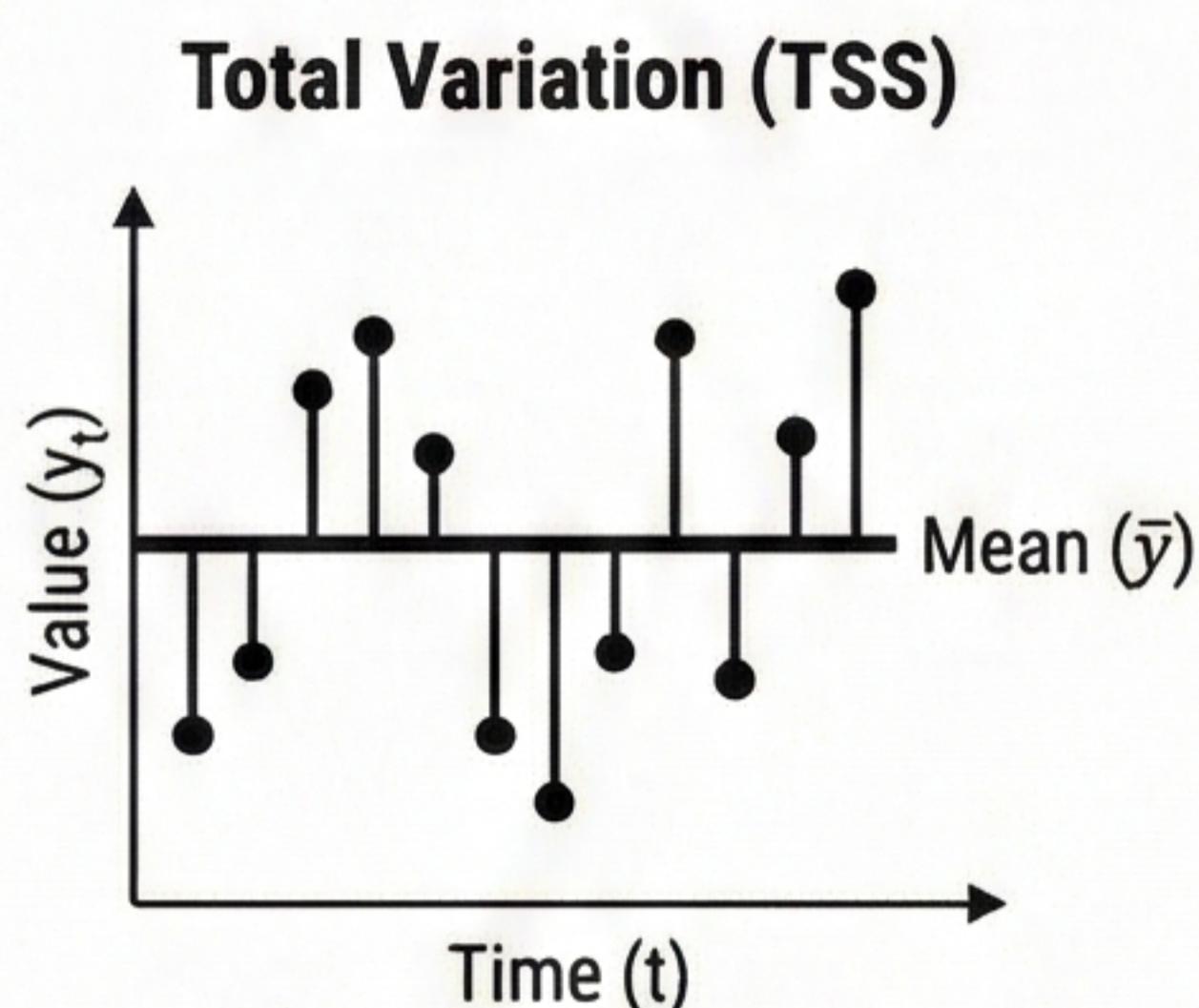
THE METRIC: MODEL FIT & EXPLANATORY POWER



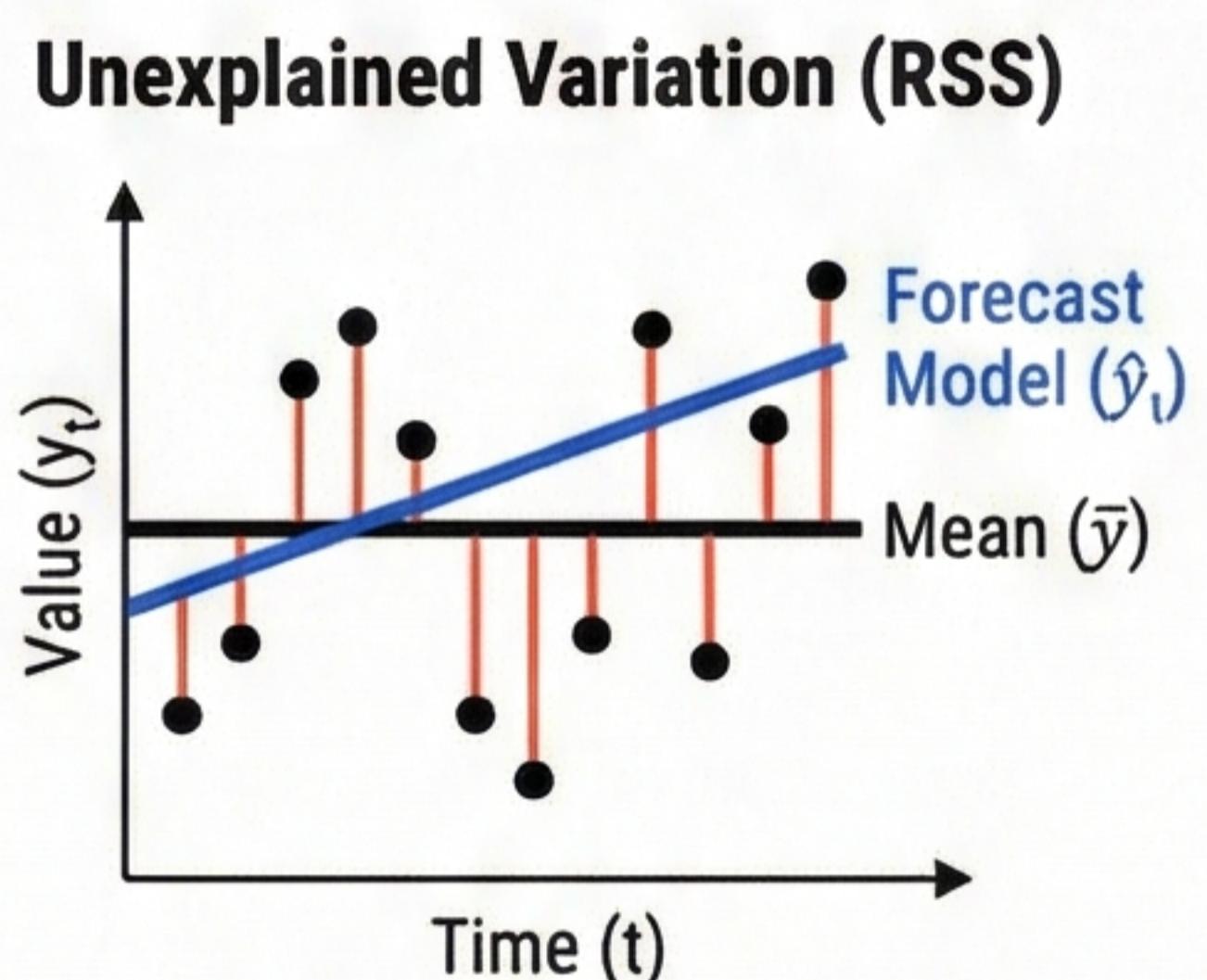
R^2 measures how well the forecasting model's predictions approximate the real data points. It indicates the strength of the relationship between time and the observed variable.

MODULE 2: VISUALIZING EXPLAINED VARIANCE (The "Decomposition")

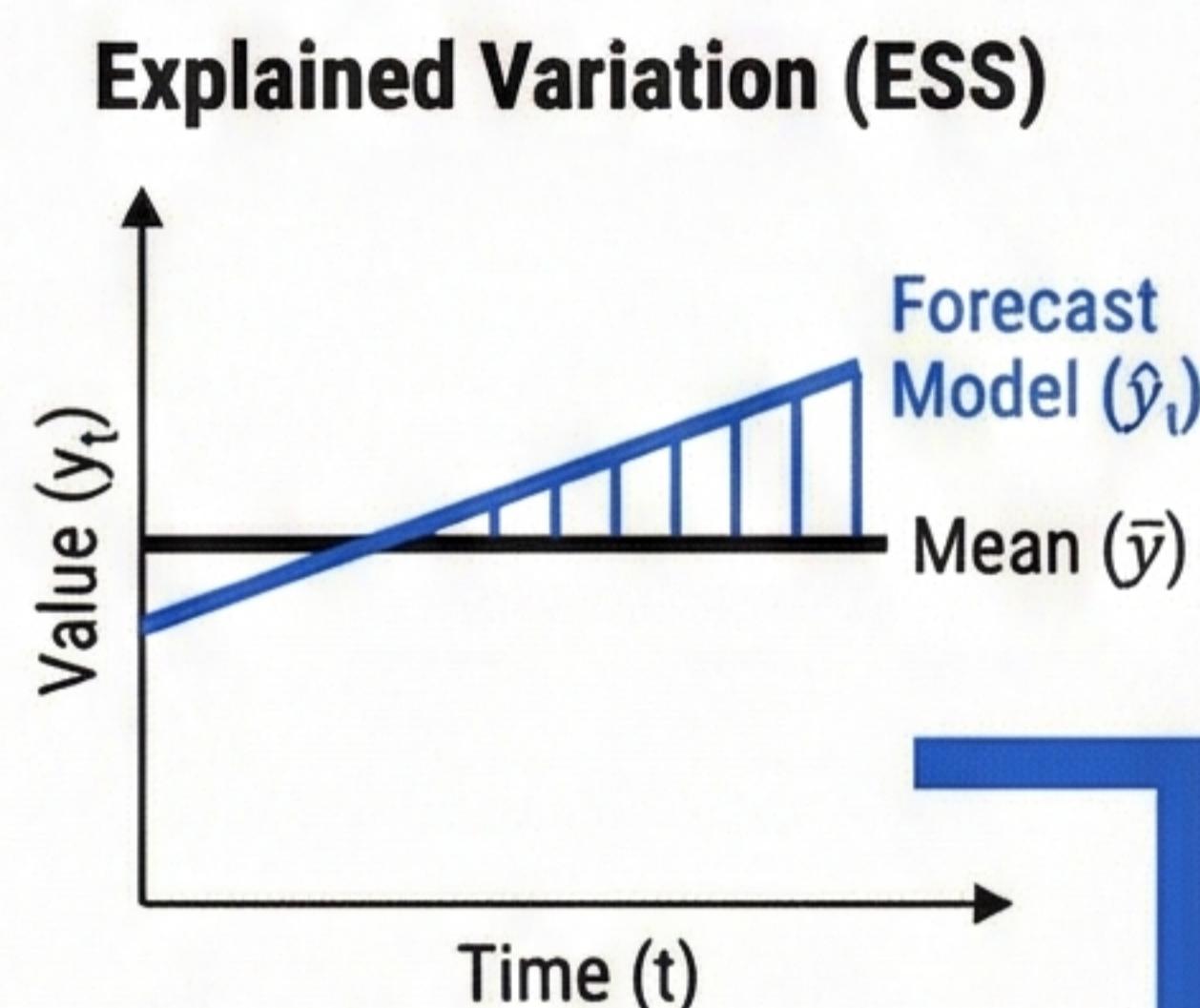
BREAKING DOWN VARIABILITY: TOTAL vs. EXPLAINED



TSS = Total Sum of Squares
(Variation from Mean)



RSS = Residual Sum of Squares
(Prediction Errors)



ESS = Explained Sum of Squares
(Variation captured by Model)

MODULE 3: THE FORMULA (A Deconstructed View)

THE CALCULATION MECHANISM: A RATIO OF VARIANCES

$$R^2 = 1 - \left(\frac{\text{RSS}}{\text{TSS}} \right)$$

$$R^2 = \frac{\text{ESS}}{\text{TSS}}$$

Total Variation (TSS)

$$\sum (y_t - \bar{y})^2$$

The baseline variability in the historical data.

Unexplained Variation (RSS)

$$\sum (y_t - \hat{y}_t)^2$$

The variability NOT captured by the model (Errors).

MODULE 4: INTERPRETATION & LIMITATIONS (Management Insight)

INTERPRETING R^2 : VALUE & CAUTION



PRACTICAL INTERPRETATION (Value)

An R^2 of 0.85 means the forecasting model explains **85%** of the historical variation in the data.

It's a measure of goodness-of-fit on *past* data.

CRITICAL LIMITATIONS (Caution)

R^2 does **NOT** guarantee future forecast accuracy.
A high R^2 can result from **overfitting** the past data, leading to poor future performance.
It also assumes a stationary relationship.



Alert: Always evaluate out-of-sample forecast errors (e.g., MAE, RMSE) alongside R^2 .

R² FOR SEASONALITY DETECTION IN TIME SERIES

Using the Coefficient of Determination to Quantify the Strength of Recurring Patterns.

MODULE 1: CORE CONCEPT & DEFINITION

THE METRIC: SEASONAL VARIANCE EXPLAINED

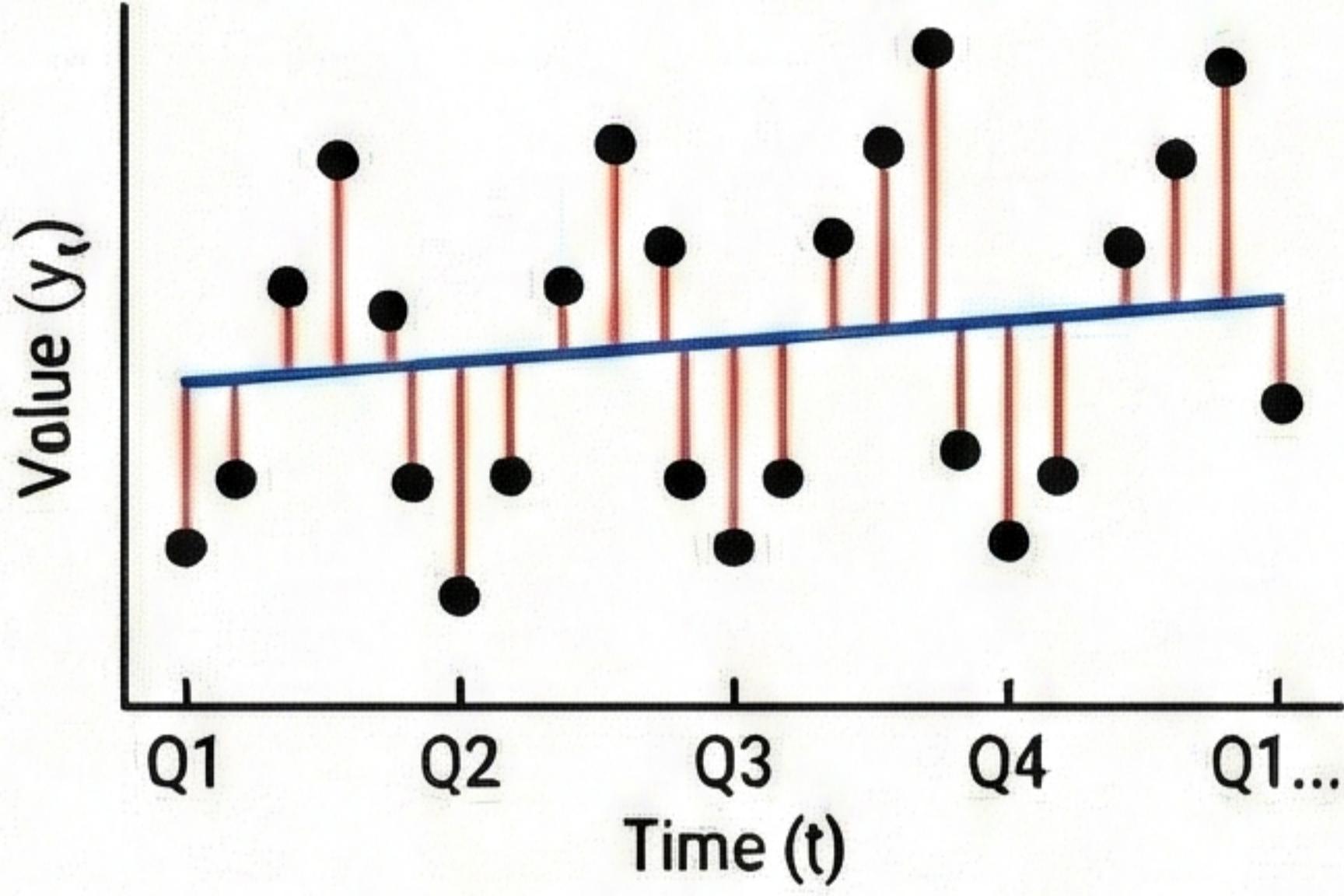
In this context, R² measures the proportion of total variance in the time series that is explained **solely by a seasonal model** (e.g., using seasonal dummy variables). A high R² indicates a strong, consistent seasonal pattern.



MODULE 2: THE MODELING MECHANISM (Visual Comparison)

ISOLATING THE SIGNAL: COMPARING MODELS

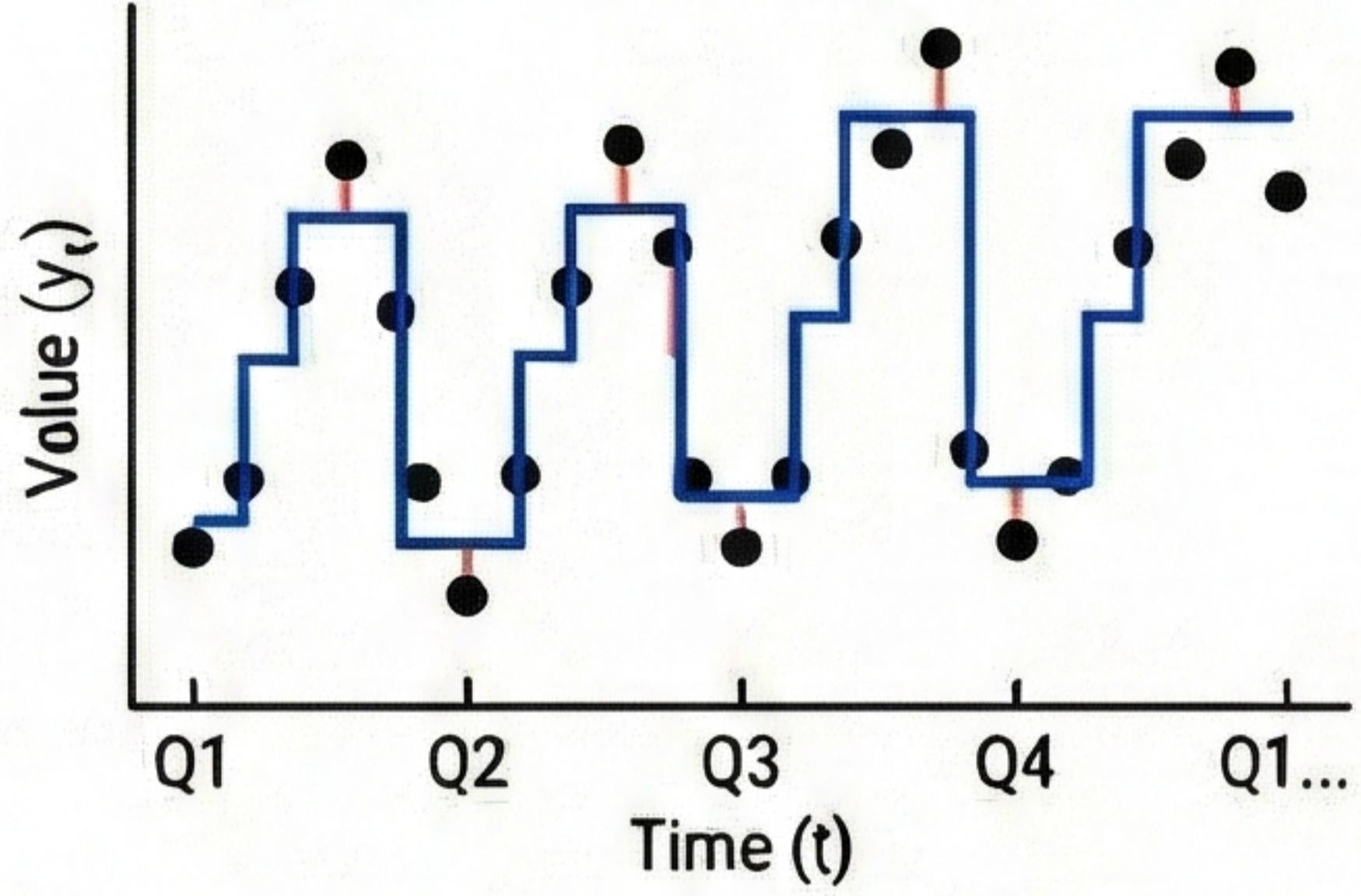
Model A: Trend Only (No Seasonality)



Large Residuals (RSS).

R² is Low (e.g., ~0.10). Weak Signal.

Model B: Seasonal Dummies Only

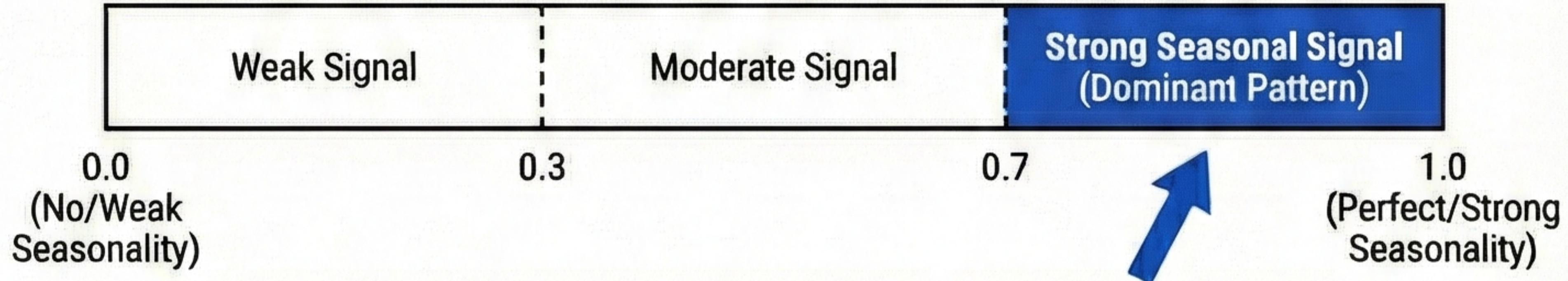


Small Residuals (RSS).

R² is High (e.g., ~0.90). Strong Signal.

MODULE 3: INTERPRETING THE R² SIGNAL (The Scale)

THE DIAGNOSTIC SCALE: STRENGTH OF PATTERN



Use R² from a seasonal-only model as a diagnostic. A high value confirms seasonality is a major driver of past variation, justifying the use of seasonal forecasting methods.

MODULE 4: CRITICAL LIMITATION & MANAGEMENT INSIGHT

MANAGEMENT ALERT: DIAGNOSIS VS. FORECAST



DIAGNOSTIC VALUE (Insight)

High R² confirms that seasonality exists and is consistent in the historical data. It's a green light to proceed with seasonal modeling techniques.



FORECASTING TRAP (Alert)

High R² does **NOT** guarantee accurate future forecasts. The pattern could change or break. It only describes the past fit. Always validate with out-of-sample testing.

Alert: R² measures past fit, not future predictive power.