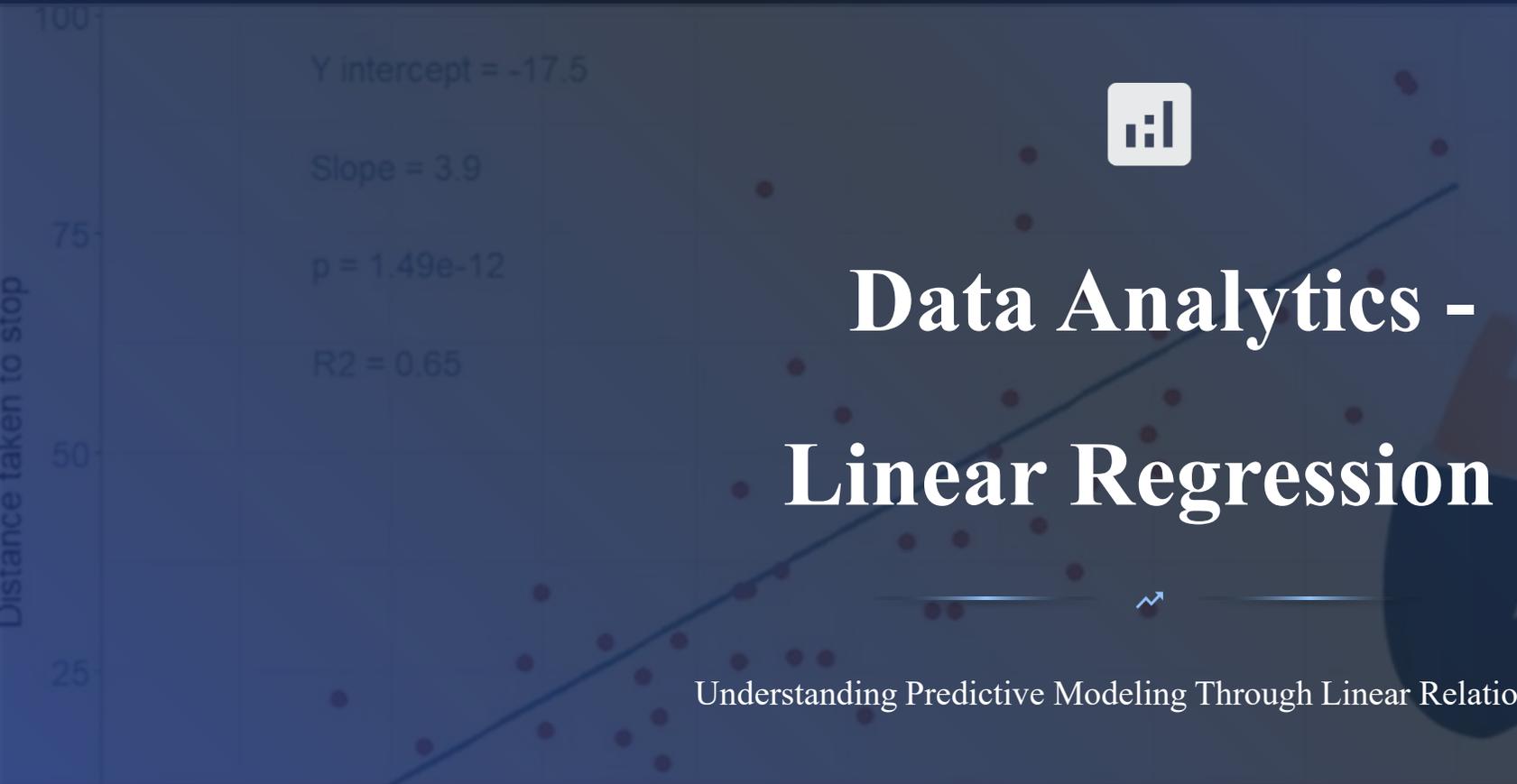


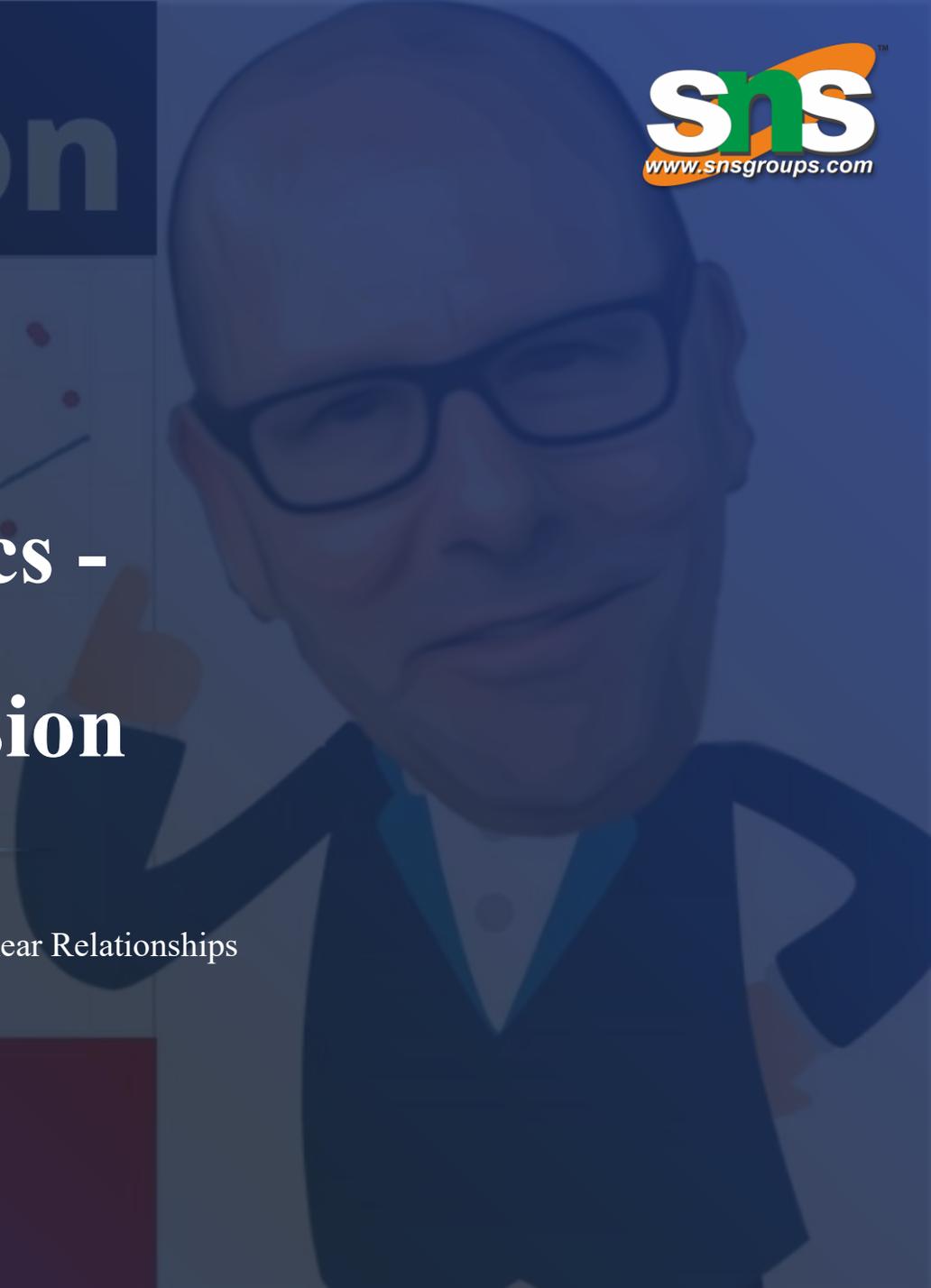
Linear regression



Data Analytics - Linear Regression

Understanding Predictive Modeling Through Linear Relationships

Made easy

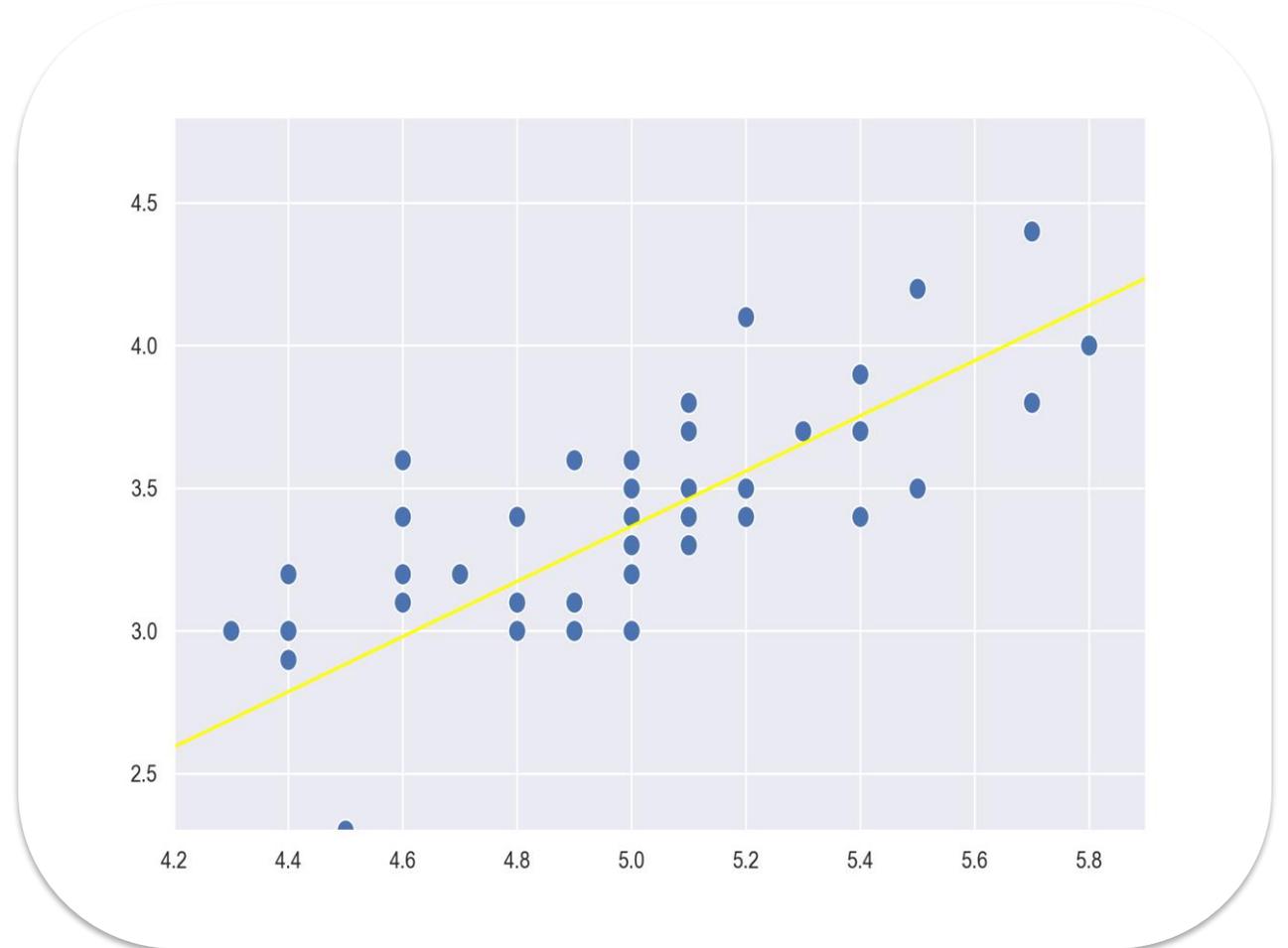


i What is Linear Regression?

- Statistical method to model relationship between variables
- Predict future outcomes based on historical data

📊 Purpose in Data Analytics

- Identify trends in data
- Make accurate predictions
- Support decision-making processes



✓ Simple Linear Regression

- One independent variable (X)
- Equation: $y = mx + b$

✗ Multiple Linear Regression

- Multiple independent variables (X_1, X_2, \dots, X_n)
- Equation: $y = b_0 + b_1X_1 + b_2X_2 + \dots + b_nX_n$

Regressions

Simple
Linear
Regression

$$y = b_0 + b_1 * x_1$$

Multiple
Linear
Regression

Dependent variable (DV) Independent variables (IVs)



$y = b_0 + b_1 * x_1 + b_2 * x_2 + \dots + b_n * x_n$

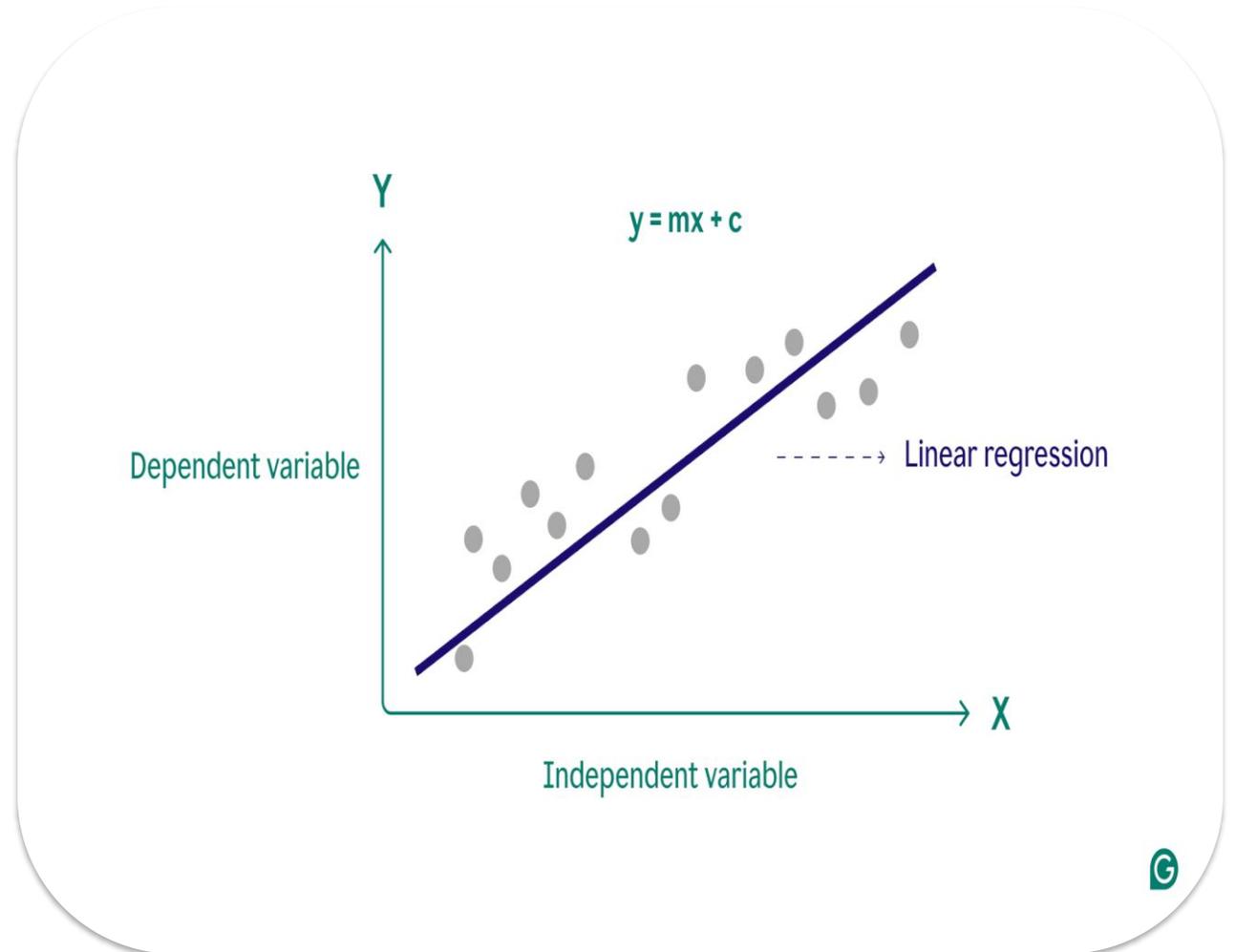
Linear Regression Equation and Components

Main Equation

$$y = mx + b$$

Components

- y** **Dependent Variable**
The value we predict
- m** **Slope**
Rate of change
- x** **Independent Variable**
The input value
- b** **Y-Intercept**
Starting value



Real-world Applications



Business

Sales forecasting, Demand prediction



Finance

Stock price analysis, Risk assessment



Healthcare

Patient outcome prediction, Disease progression



Science

Experimental data analysis, Trend identification

Foundations of Linear Regression

Error Term

The difference between observed and predicted values.

Dependent Variable

The variable being predicted or explained.

Slope

The rate of change of the dependent variable per unit change in the independent variable.

Simple Linear Regression Model

Independent Variable

The variable used to predict the dependent variable.

Intercept

The point where the regression line crosses the y-axis.

Assumptions of Linear Regression

Key Assumptions

1 Linearity

Relationship between variables is linear

2 Independence

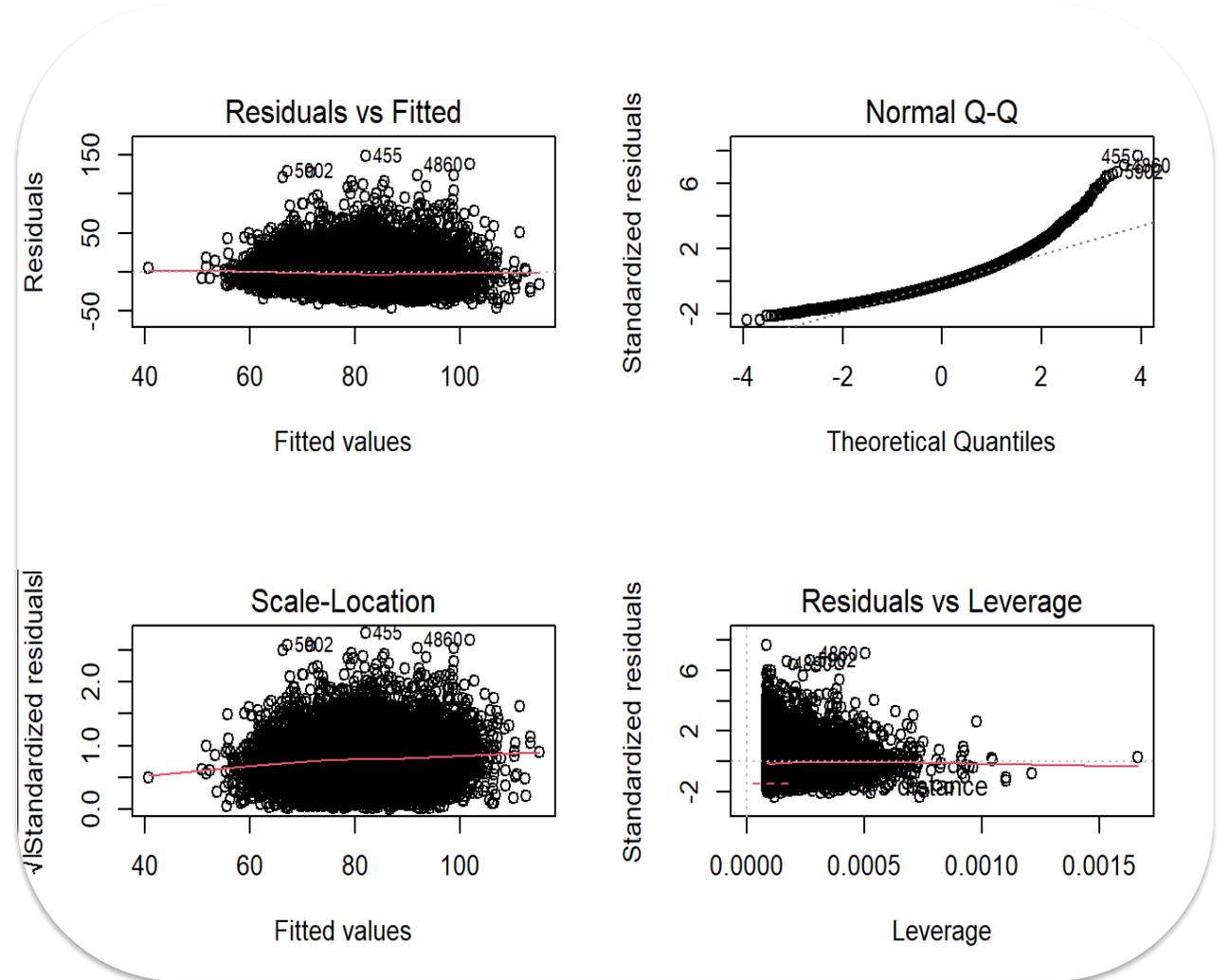
Observations are independent of each other

3 Homoscedasticity

Constant variance of residuals

4 Normality

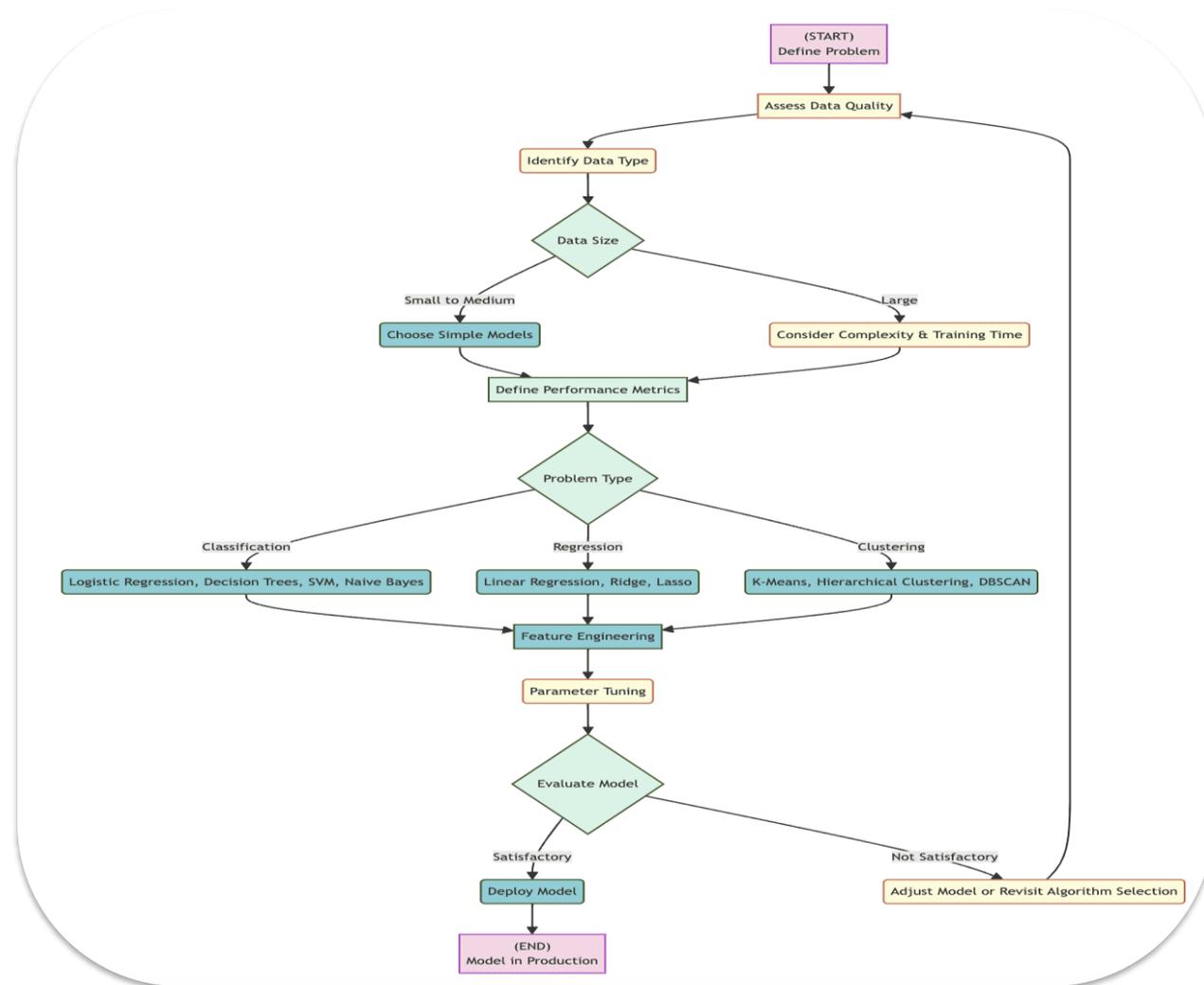
Residuals are normally distributed



Steps in Linear Regression Analysis

Process Flow

- 1 Data Collection**
Gather relevant dataset
- 2 Data Preprocessing**
Clean and prepare data
- 3 Model Training**
Fit the regression model
- 4 Model Testing**
Validate with test data
- 5 Model Evaluation**
Assess model performance
- 6 Prediction**
Make future predictions



Key Metrics



R-squared (R^2)

Measures goodness of fit

- Range: 0 to 1
- Higher is better



Mean Squared Error (MSE)

Average of squared differences

- Lower is better

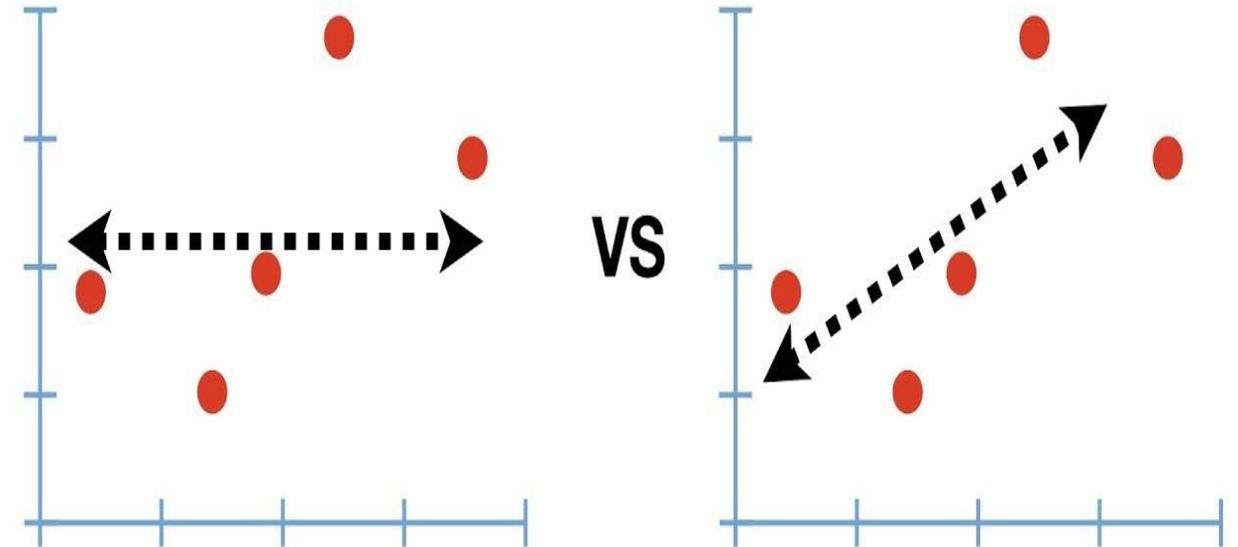


Root Mean Squared Error (RMSE)

Square root of MSE

- Interpretable in target units

R^2 (R squared)....



...Clearly Explained!!!

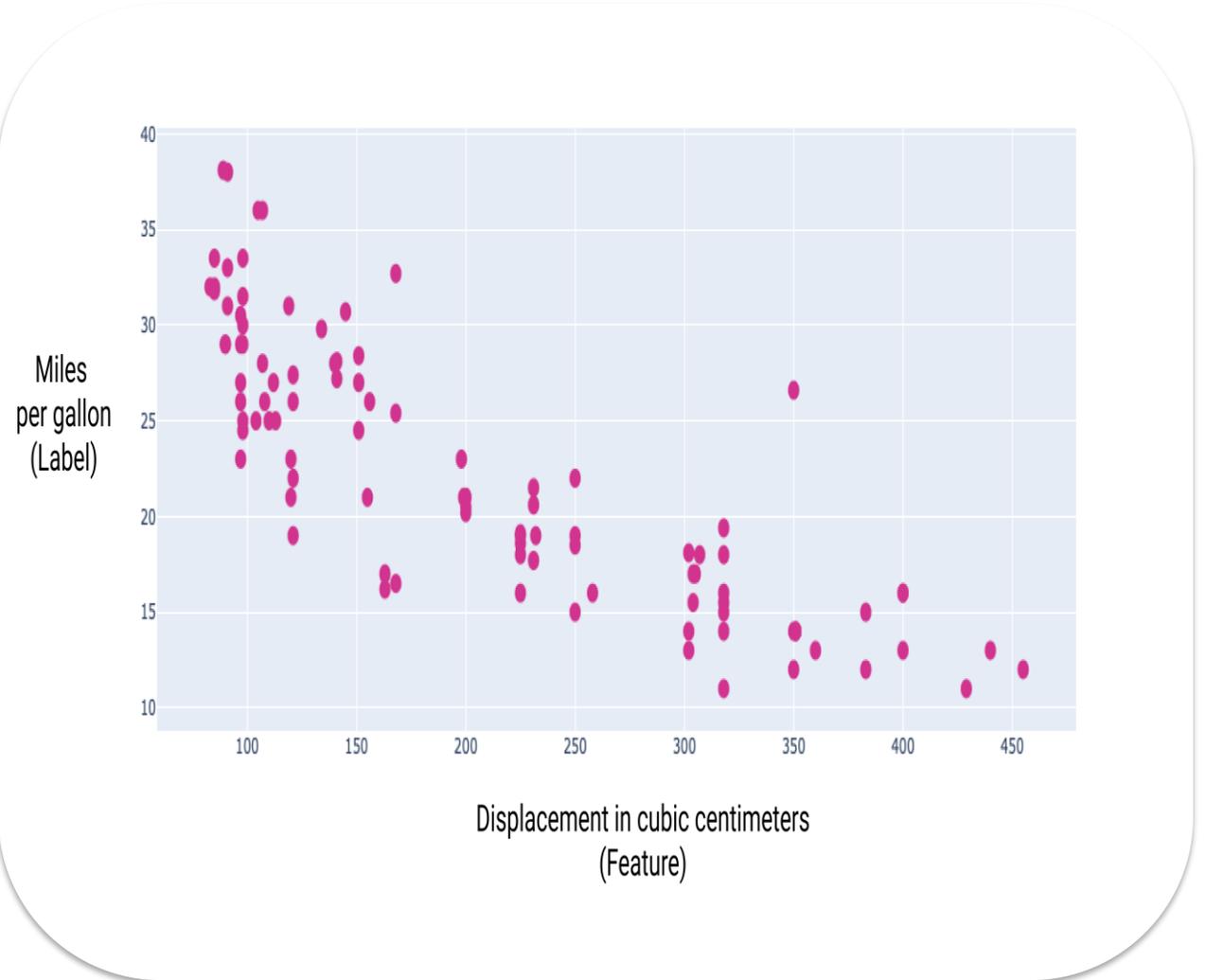
Advantages and Limitations

✓ Advantages

- Simple to understand and interpret
- Efficient for small datasets
- Works well with linear relationships

⚠ Limitations

- Assumes linear relationship
- Sensitive to outliers
- Cannot handle non-linear patterns
- May overfit with many variables



Mind Map with Recap and Summary

Key Takeaways

- ✓ Linear regression models relationships between variables
- ✓ Equation $y = mx + b$ forms the foundation
- ✓ Multiple types: Simple & Multiple Regression
- ✓ Widely applied across business, finance, healthcare & science
- ✓ Four key assumptions must be met for valid results
- ✓ Process: Collection → Training → Evaluation → Prediction
- ✓ R^2 , MSE, RMSE measure model performance
- ✓ Balance advantages with limitations for effective use

