

Implementing Our Regression Solution



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Overview



Pipeline once again

After all: What is Model Training?

Foundational Concepts

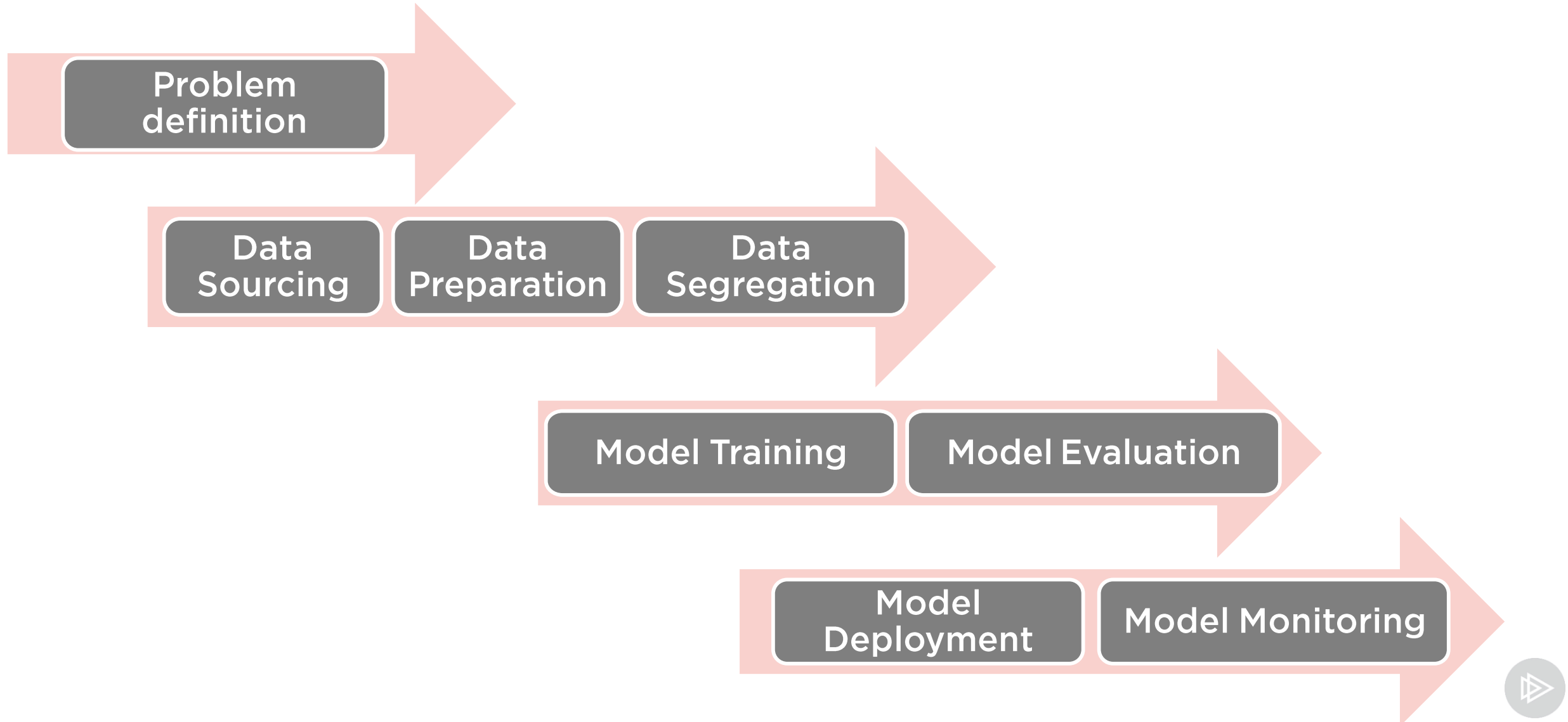
Linear Regression

Model Evaluation

Demo



Model Training and Evaluation



What Is Model Training?



All Machine Learning algorithms use one principle

Three types of Machine Learning algorithms (generally)

- Supervised Learning
- Unsupervised Learning
- Reinforcement Learning

Our focus is Supervised Learning

Types of Supervised Learning Algorithms

Regression

For continuous values

Classification

For discrete (categorical)
values



In the context of Supervised
Learning:

Machine Learning training is
the process of learning an ML
algorithm how to find patterns
in the input data so that they
correspond to the target,
resulting a machine learning
model



Why Should I Care?

Problem

Supervised
Learning

Regression
Problem

Linear
Regression

Ridge
Lasso

Decision
Tree
Regressor

Classification

Supervised

Regression

Unsupervised

Machine Learning
Algorithms

Clustering

Dimension
Reduction

Pattern
Search

Reinforcement



Learning More

How to Think About Machine Learning Algorithms

by Swetha Kolalapudi

If you don't know the question, you probably won't get the answer right. This course is all about asking the right machine learning questions, modeling real-world situations as one of several well understood machine learning problems.

How people feel about something



or



can be measured using a technique known as

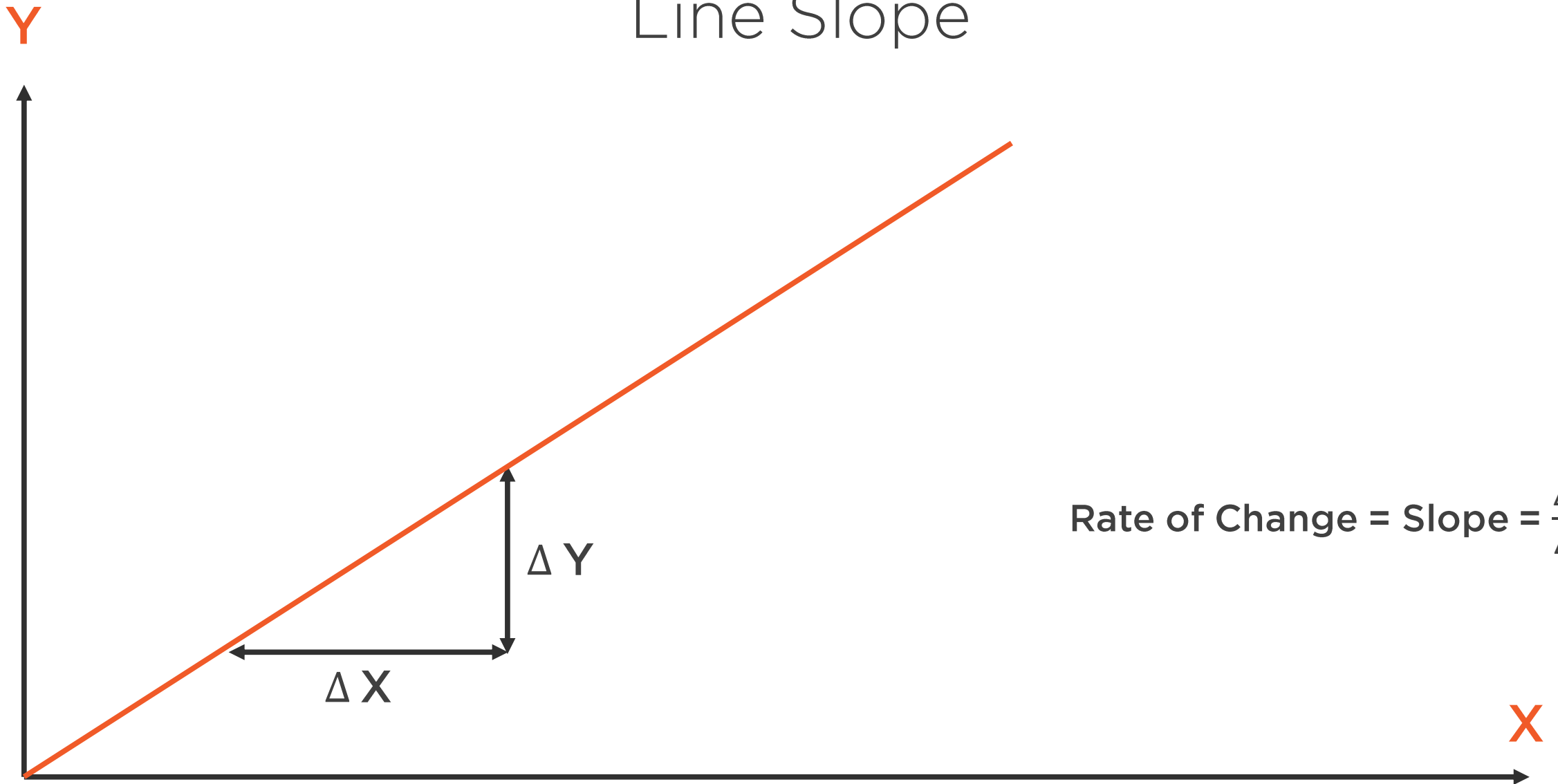
Sentiment Analysis



Foundational Concepts



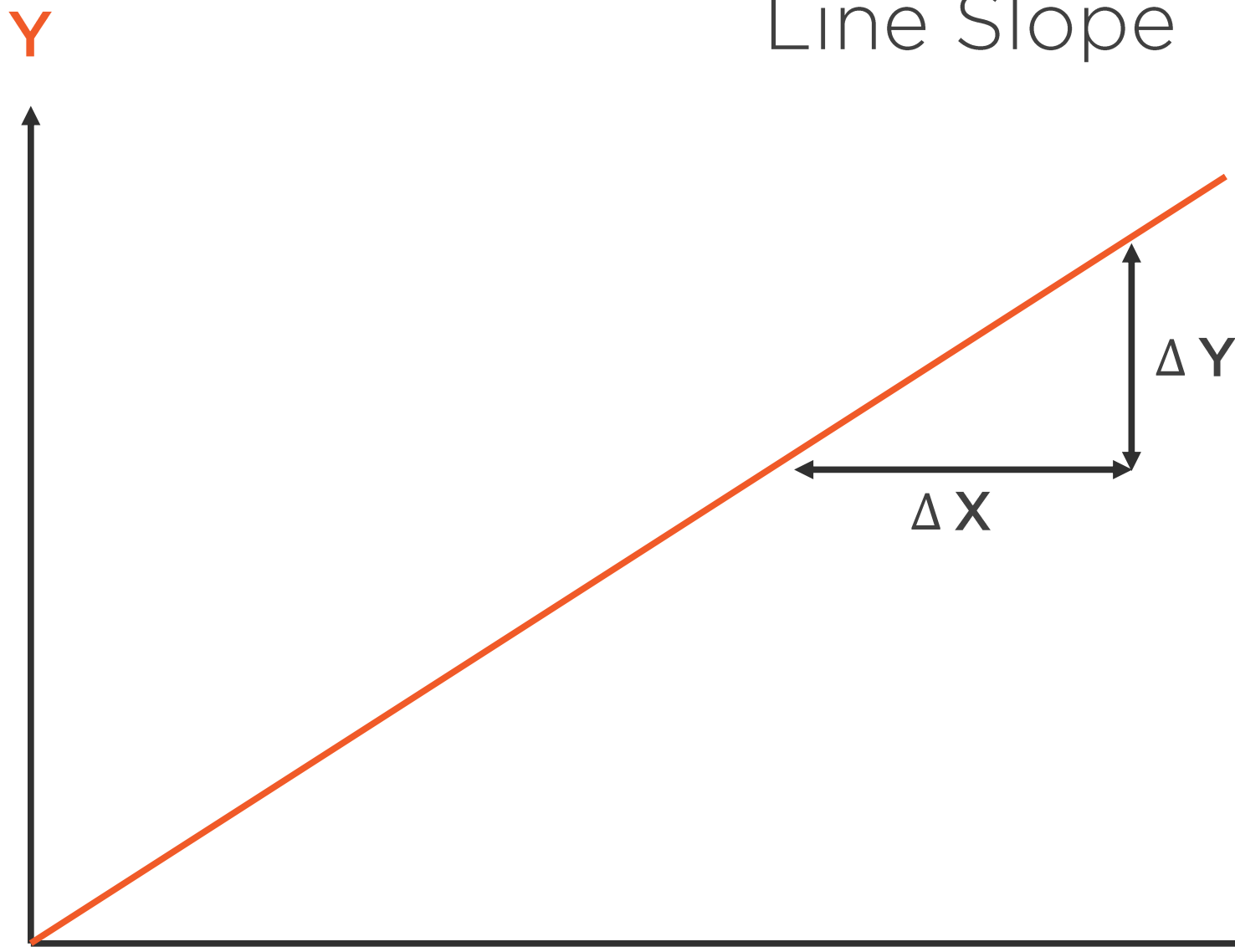
Line Slope



$$\text{Rate of Change} = \text{Slope} = \frac{\Delta y}{\Delta x}$$



Line Slope



$$\text{Rate of Change} = \text{Slope} = \frac{\Delta y}{\Delta x}$$



Important Line Slope Cases



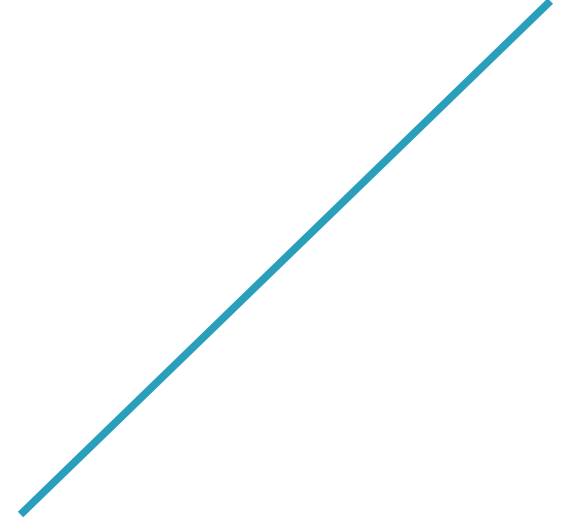
Horizontal Line

Slope = 0



Vertical Line

Slope = Infinity

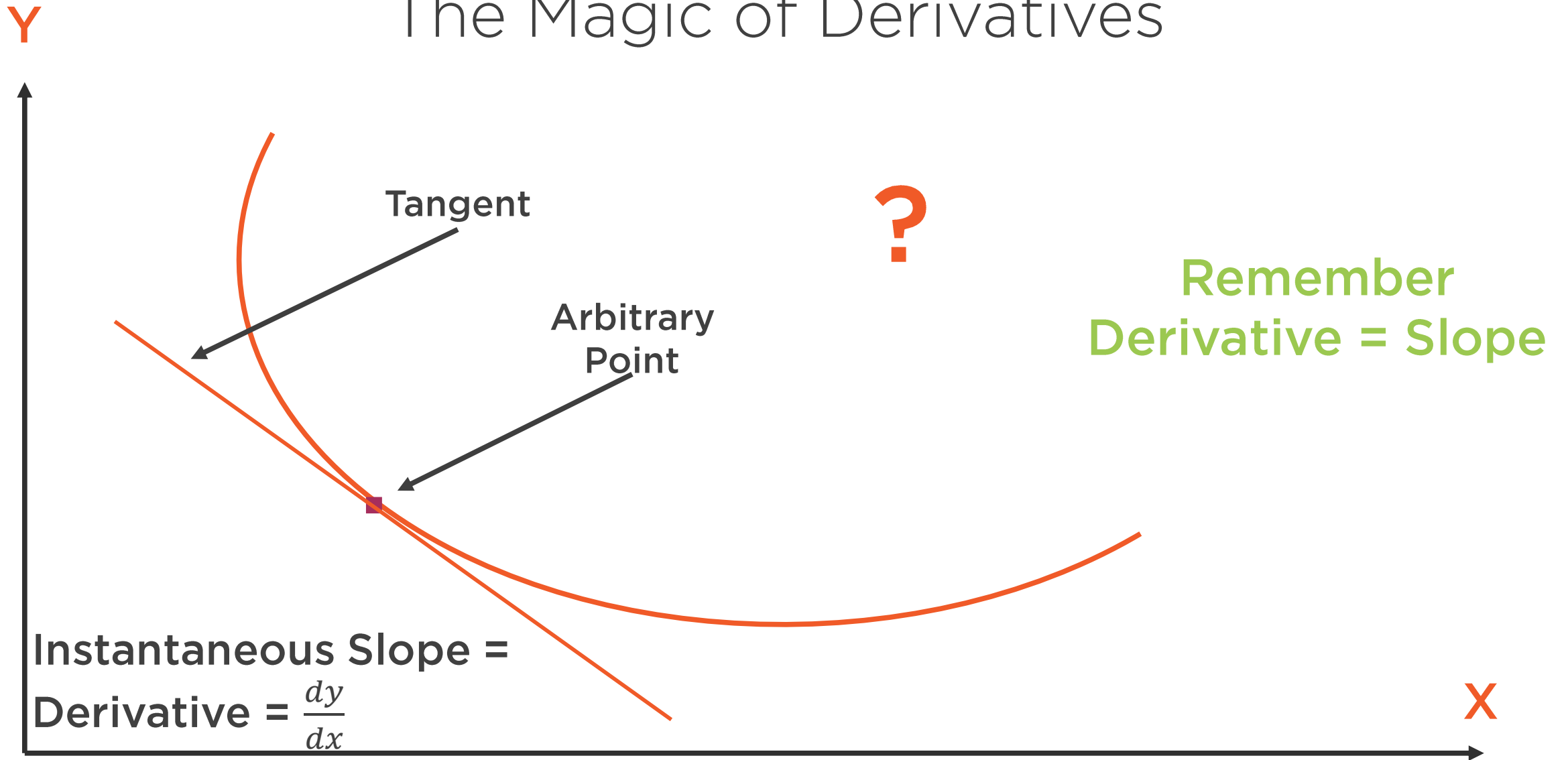


45 Degrees Line

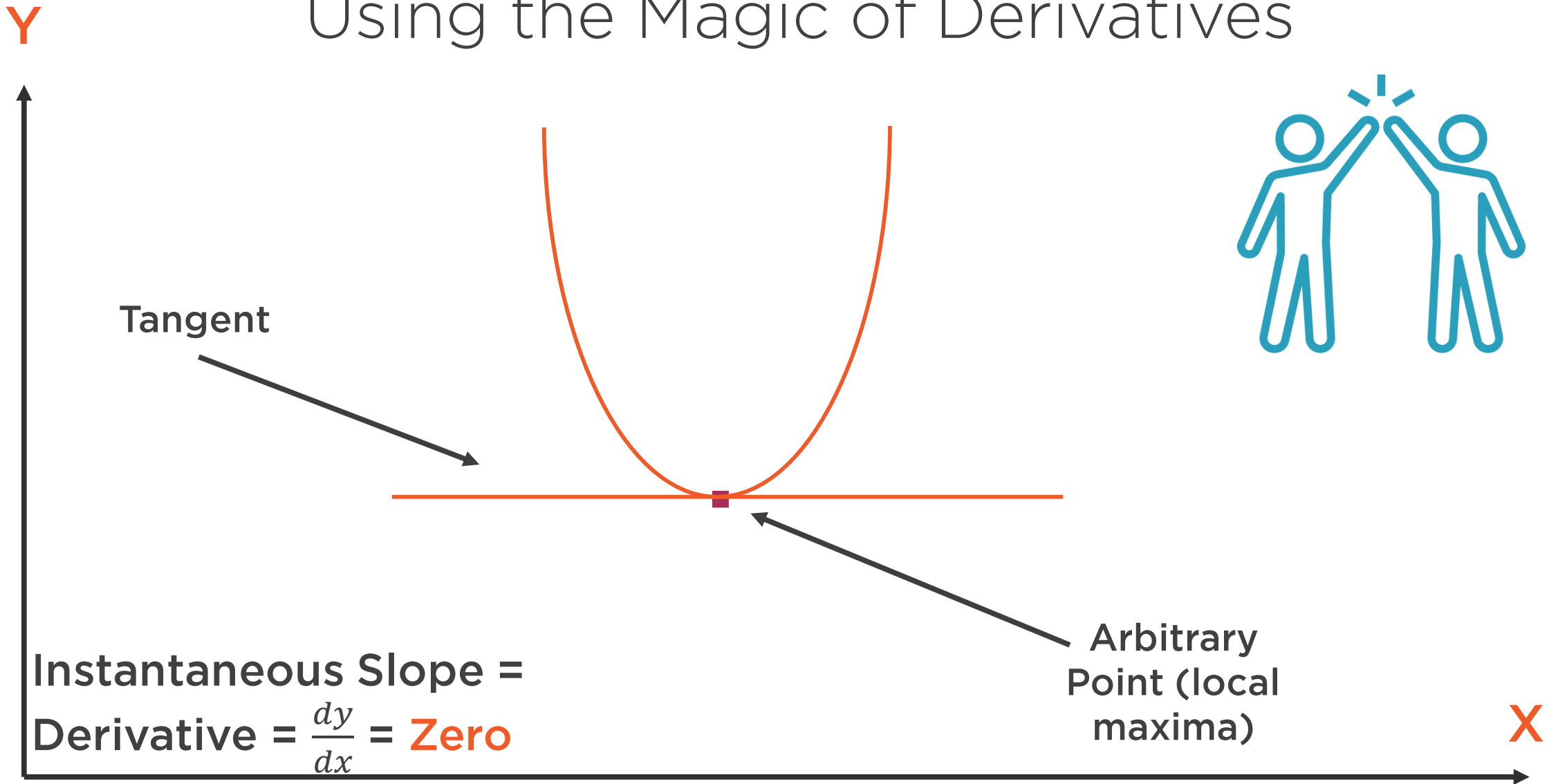
Slope = 1



The Magic of Derivatives



Using the Magic of Derivatives



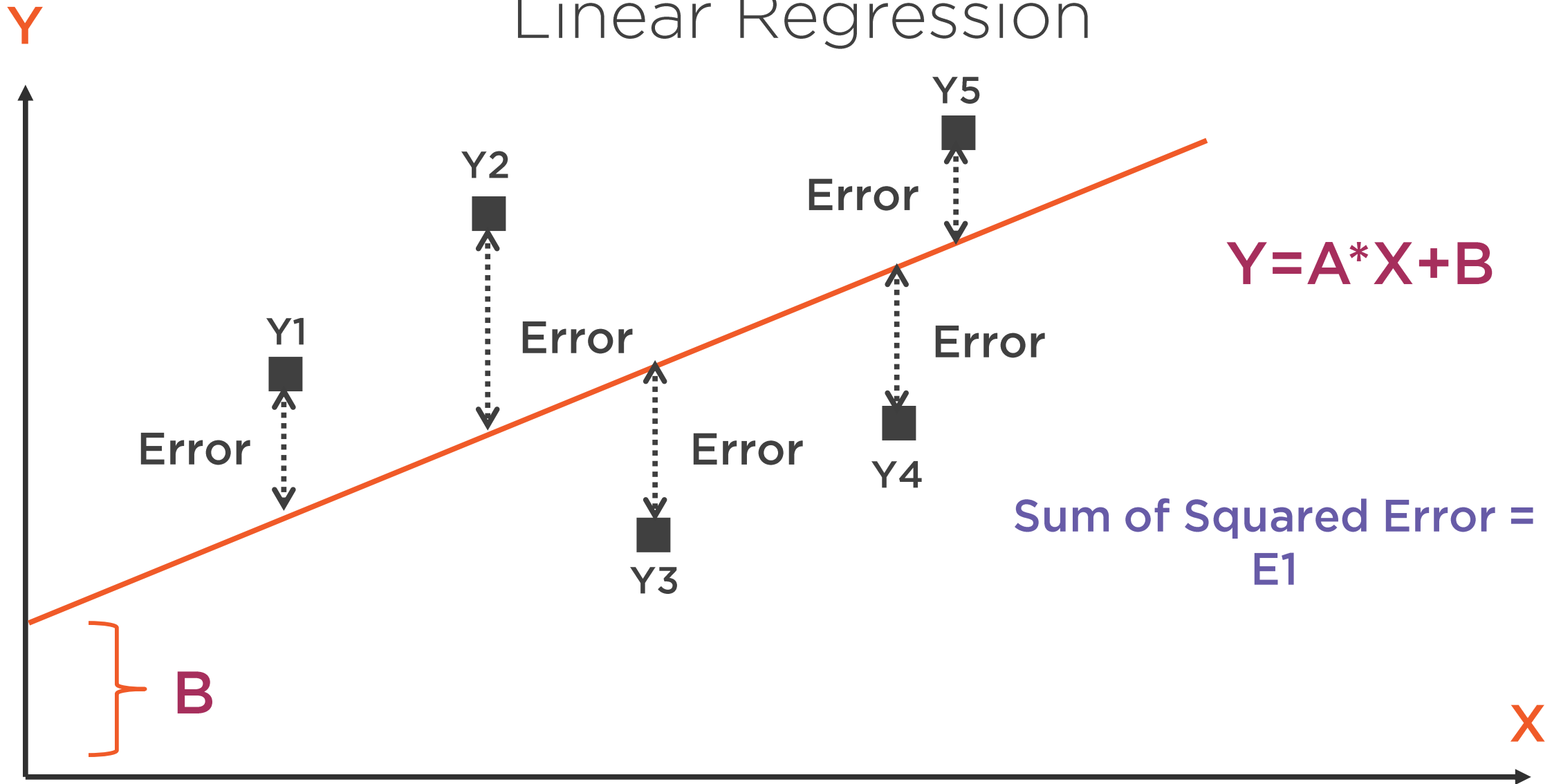
If we want to get the minimum of a function, then we calculate its derivative when it is equal to zero!



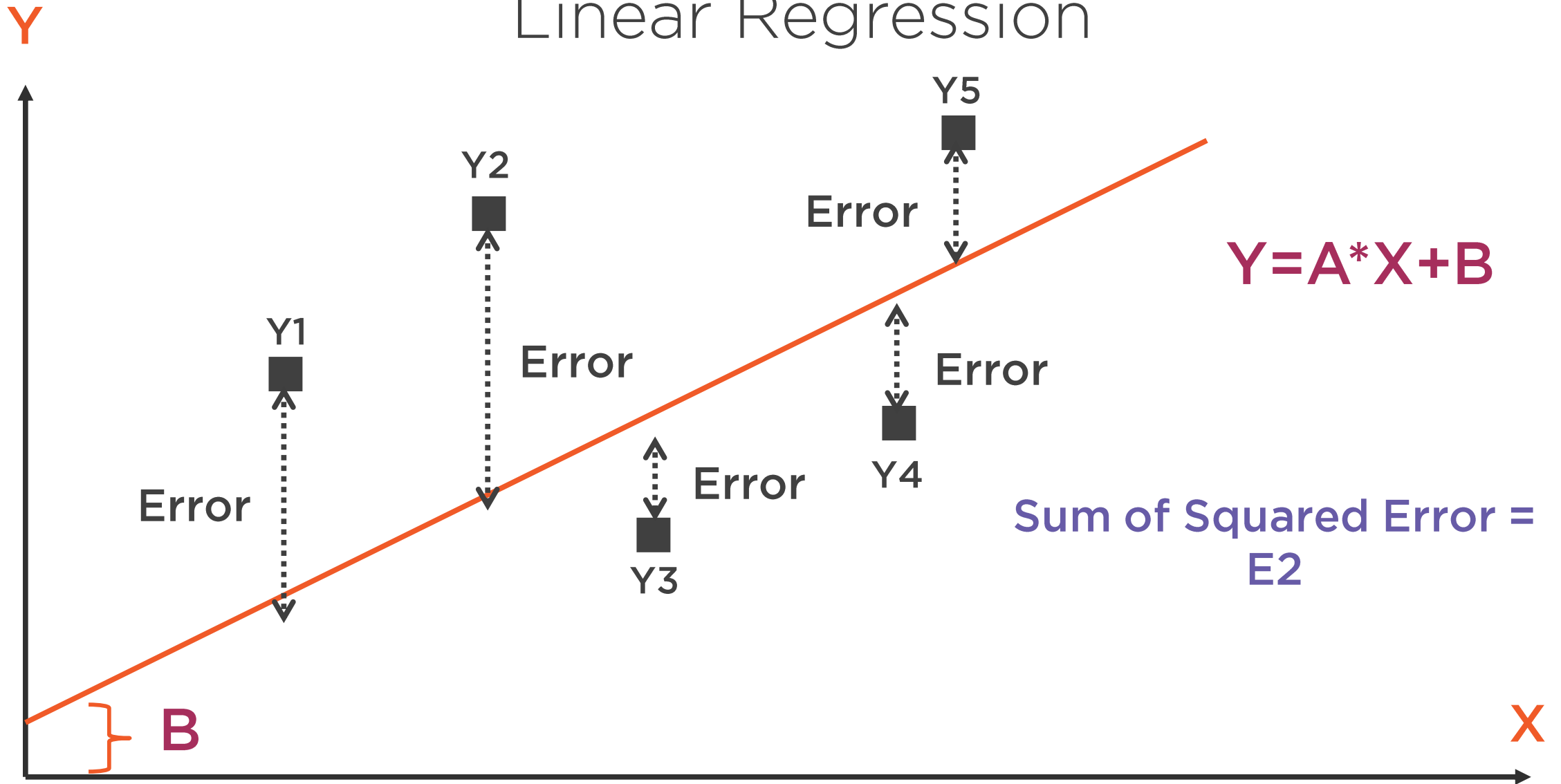
Linear Regression Algorithms



Linear Regression



Linear Regression



How Linear Regression Works?



The of **sum of residuals** is calculated as $\sum(Y_n - (AX_n + B))^2$



The sum of residuals **minimized** by taking the **derivative** at zero with respect to the **slope**



The sum of residuals **minimized** by taking the **derivative** at zero with respect to the **intercept**

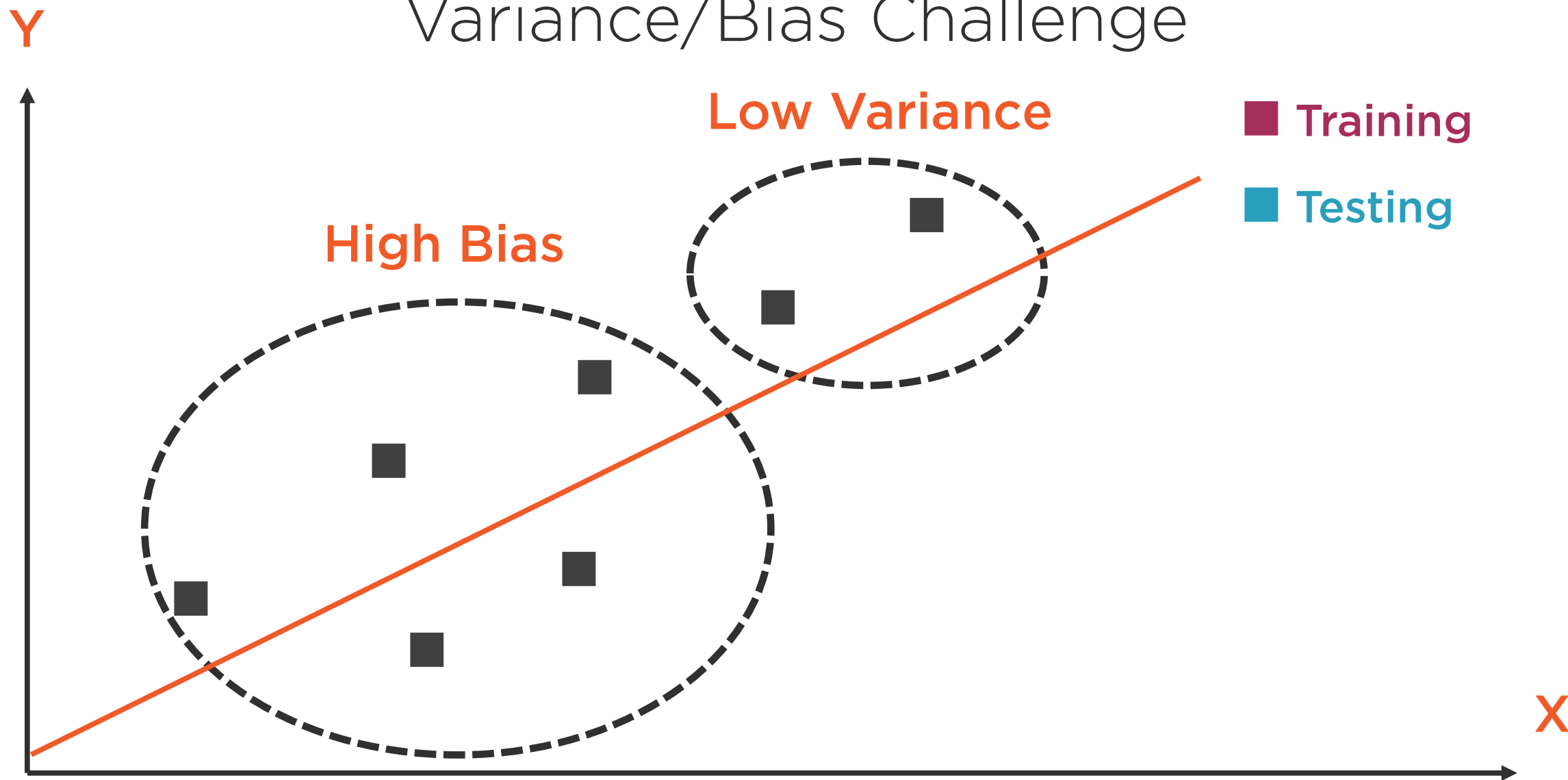


We solve the equations to get the **slope (A)** and **intercept (B)**

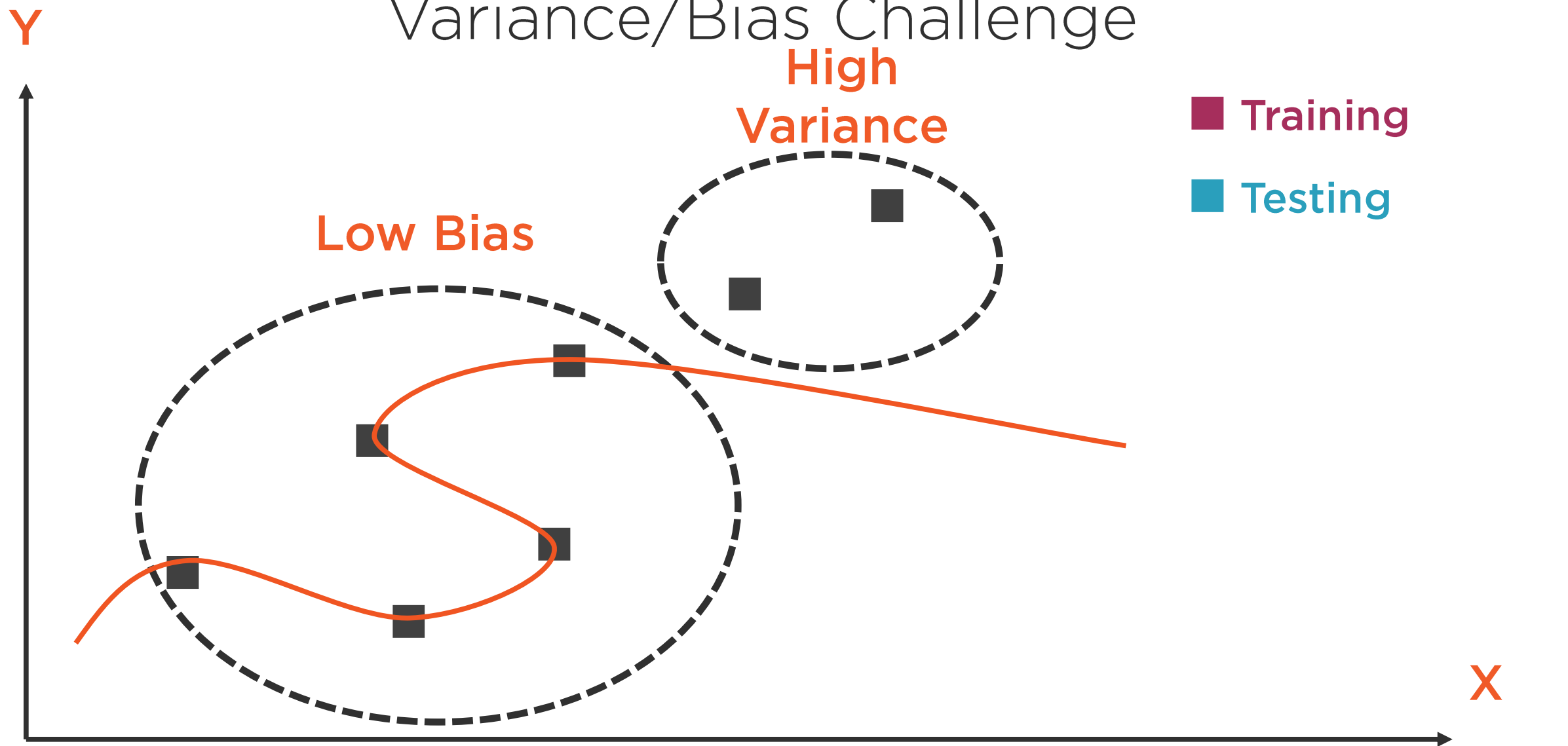


We solve for the **slope (A)** and the **intercept (B)**

Variance/Bias Challenge



Variance/Bias Challenge



Variance/Bias trade-off

Variance

$$Var(x) = \sigma^2 = \frac{\sum(\bar{x} - u)^2}{N}$$

Occurs when the ML algorithm has high ability to fit training data

Error due to fluctuations in the training set

High variance usually indicates overfitting



Bias

Occurs when a ML algorithms has limited ability to learn

Wrong assumption about the problem nature

High bias usually associated with underfitting



Regularization

Is the process of tuning model parameters or complexity so that the model performs better at predicting (generalizing) on out of sample data.



Linear Regression Regularization

Ridge Regression

Be concise and keep
the text to four lines
or fewer

Lasso Regression

Be concise and keep
the text to four lines
or fewer

Elastic Net Regression

Combines techniques
from Ridge and Lasso



Other Regression Algorithms



K-neighbors Regression



Simple algorithm

Relies on distance measurement

Requires standardization

Support Vector Regression (SVR)



Used for both regression and classification (SVM)

Similar to linear regression



Decision Tree Regressor

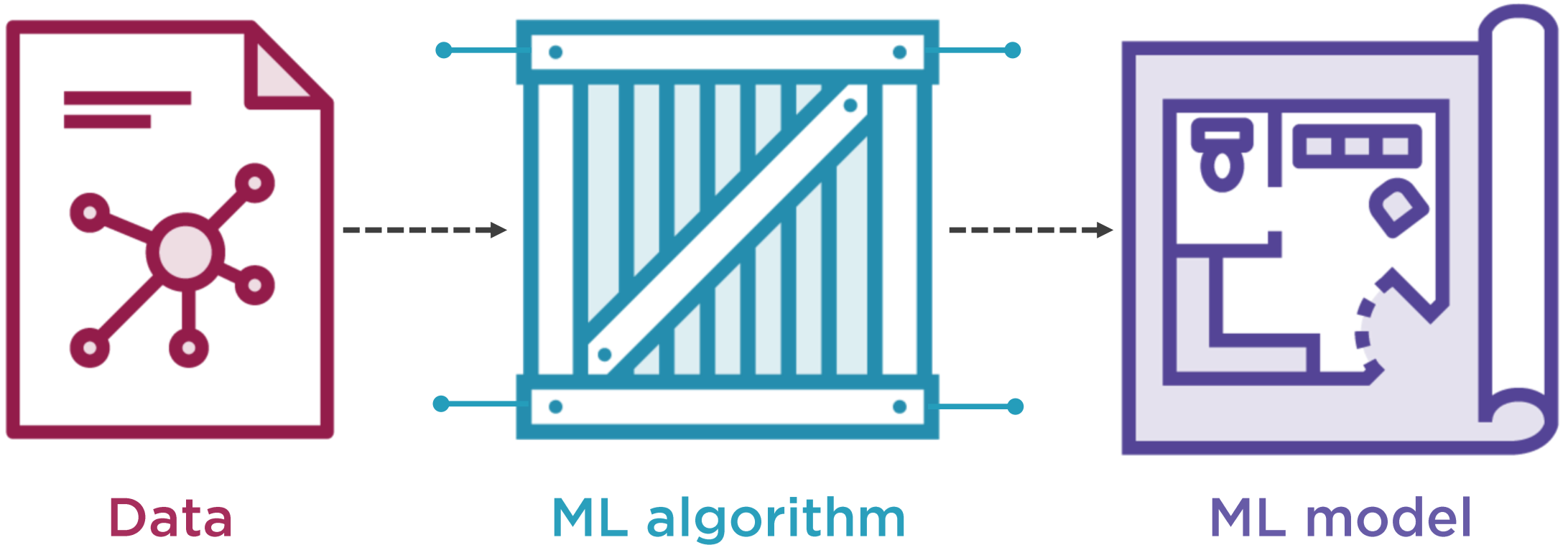


Used for both regression and classification

Reaches the answer by structuring the data in tree leaves



Should I Care?



Model Evaluation



Regression Models Evaluation Metrics

Max Error

Mean Absolute
Error

Mean Squared
Error

R Squared

Others



Max Error

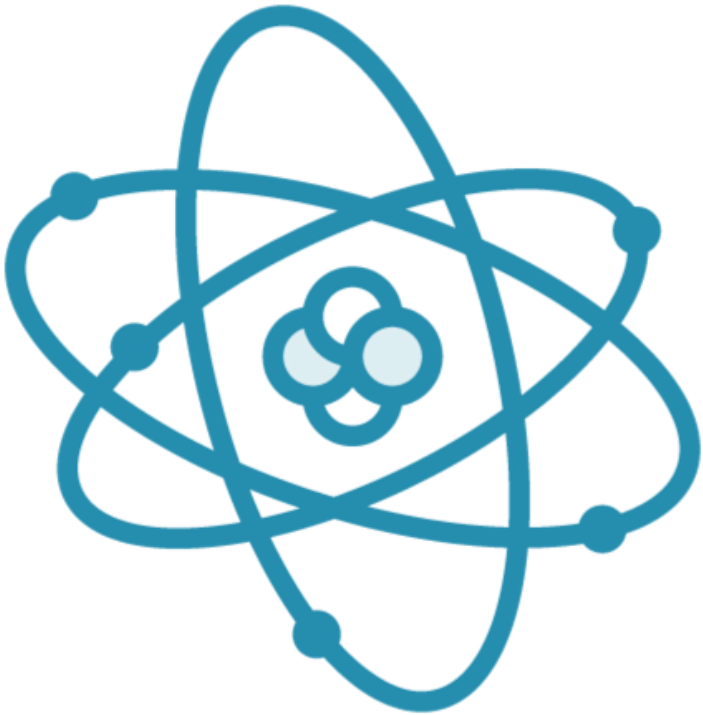


Captures the worst case

How much we can tolerate



Mean Absolute Error



Average of absolute errors



Mean Squared Error



Average of squared errors



MAE vs. MSE

Mean Absolute Error

Removes negative signs by taking **absolute** value

More **robust** to outliers

Mean Squared Error

Removes negative signs by **squaring** the values

Better when we want to **penalize** outliers



R^2 (Coefficient of Determination)



$$R^2 = \frac{Var(mean) - Var(fit)}{Var(mean)} = Correlation^2$$

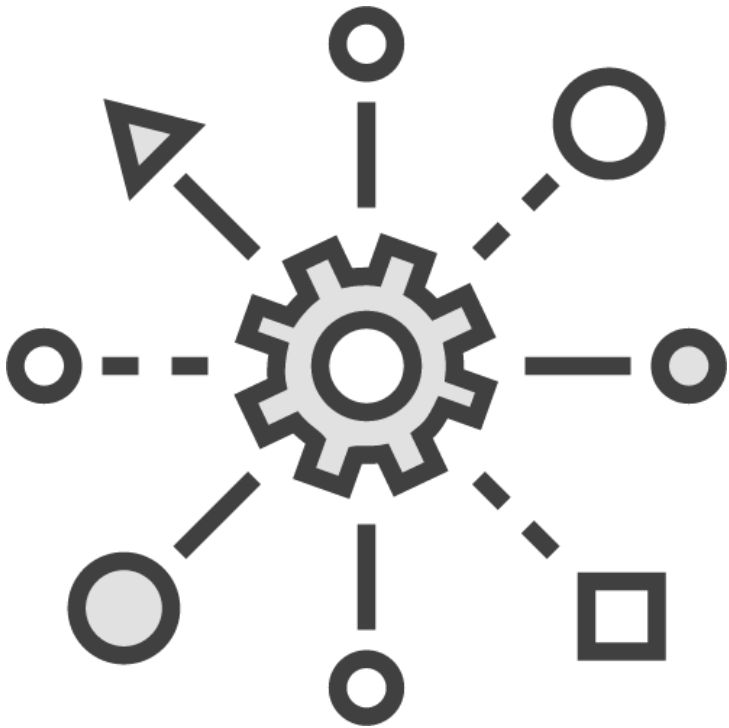
Tells us how much percentage of the data is explained by the relationship but no direction

Var(mean) = 22 and Var(fit) = 4 then $R^2 = 0.81$

Easier to interpret



Others



Median absolute error

Mean squared log error



Demo



Data Segregation

- Train/Test split
- K-Fold Cross Validation



Glitch with Scikit-learn



Summary



Final look to ML pipeline

What is model training?

Foundational concepts

- Slope
- Derivative

Linear regression algorithms

Variance/Bias trade-off

Other regression algorithms

Model evaluation

Demo

