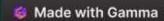


Artificial Intelligence & Machine Learning

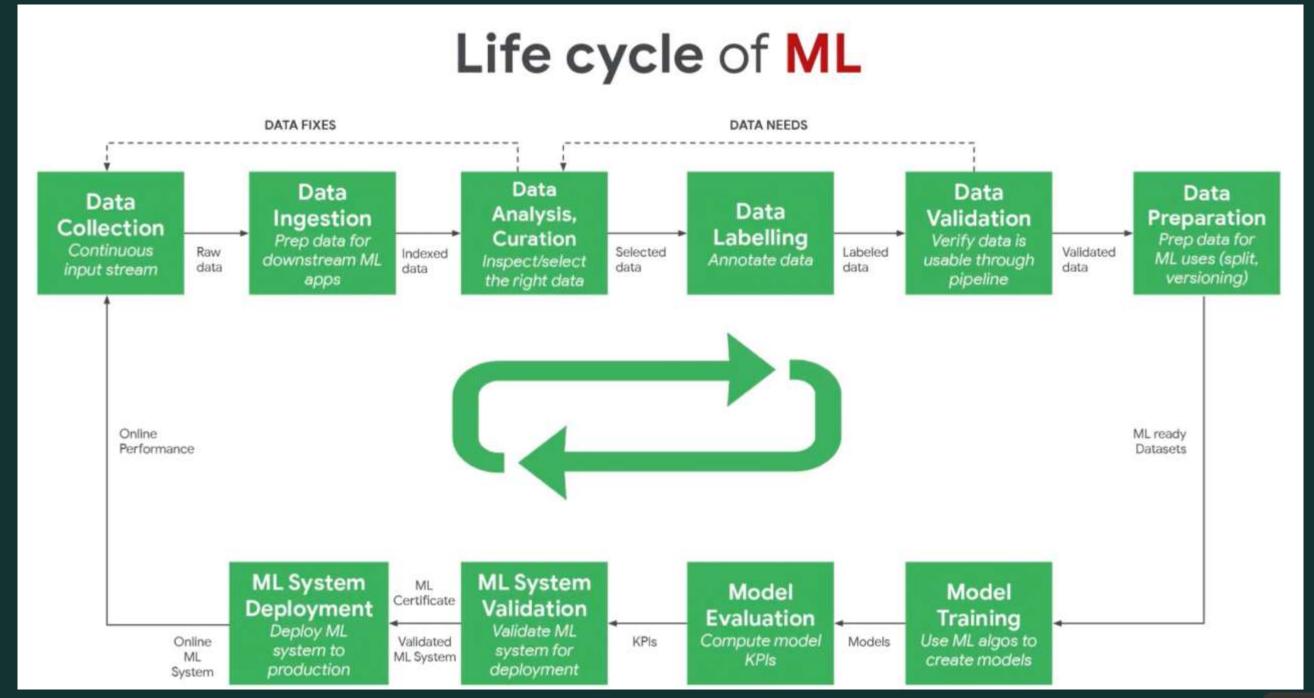
Introduction to AI/ML.

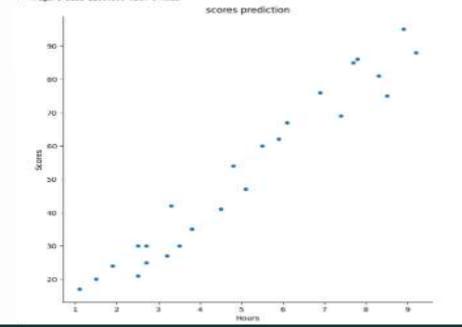
A powerful set of tools for solving complex problems.

Maheza Novrayuda



WORKFLOW AI / ML



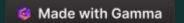


Exploratory Data Analysis

- Understanding DataAnalyze relationships between features.
- Data CleaningAddress missing values and inconsistencies.

- Identify PatternsUncover insights and trends in data.
- Feature Engineering

 Transform data for better model performance.



Supervised Learning

Regression

Predicting continuous values.

Examples: predicting house prices, sales forecast.

Classification

Categorizing data into distinct classes.

Examples: identifying spam emails, classifying images.



Linear Regression

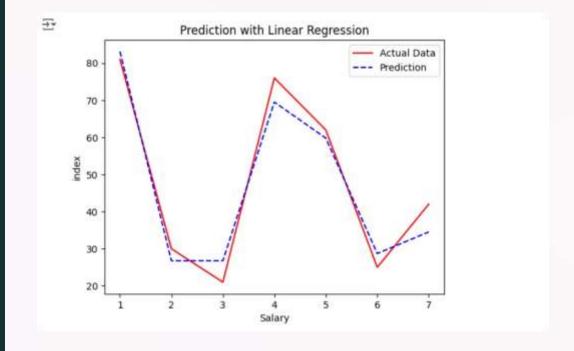
1

Simple Linear

One input variable.

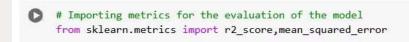
Multiple Linear

Multiple input variables.





Prediction with Decision Tree Actual Data --- Prediction Actual Data --- Prediction Actual Data --- Prediction



- [] # Calculate R square vale
 rsq = r2_score(y_test,y_pred)
- [] print('r square Linear Regression:',rsq)
- → r square Linear Regression: 0.9553509219739938
- [] # Intecept and coeff of the line print('Intercept of the Linear Regression model:',lr_model.intercept_) print('Coefficient of the line Linear Regression:',lr_model.coef_)
- Intercept of the Linear Regression model: 2.4803670915057623 Coefficient of the line Linear Regression: [9.71409219]

Then it is said to form a line with result in Linear Regression

$$y = 2.480367 + 9.714092 x$$

with x is the value of hours of study

Decision Tree

Tree Structure

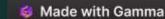
Decision-making branches based on features.

Splitting Criteria

Gini impurity, entropy, or information gain.

Leaf Nodes

Represent predictions or classifications.



Model Evaluation

Model	R-squered
Linear Regression	0.75
Decision Tree	0.88

Conclusion

- Decision Tree models are better than Linear Regression Model
- -AI/ML empowers data-driven decision making.

