Linear Regression

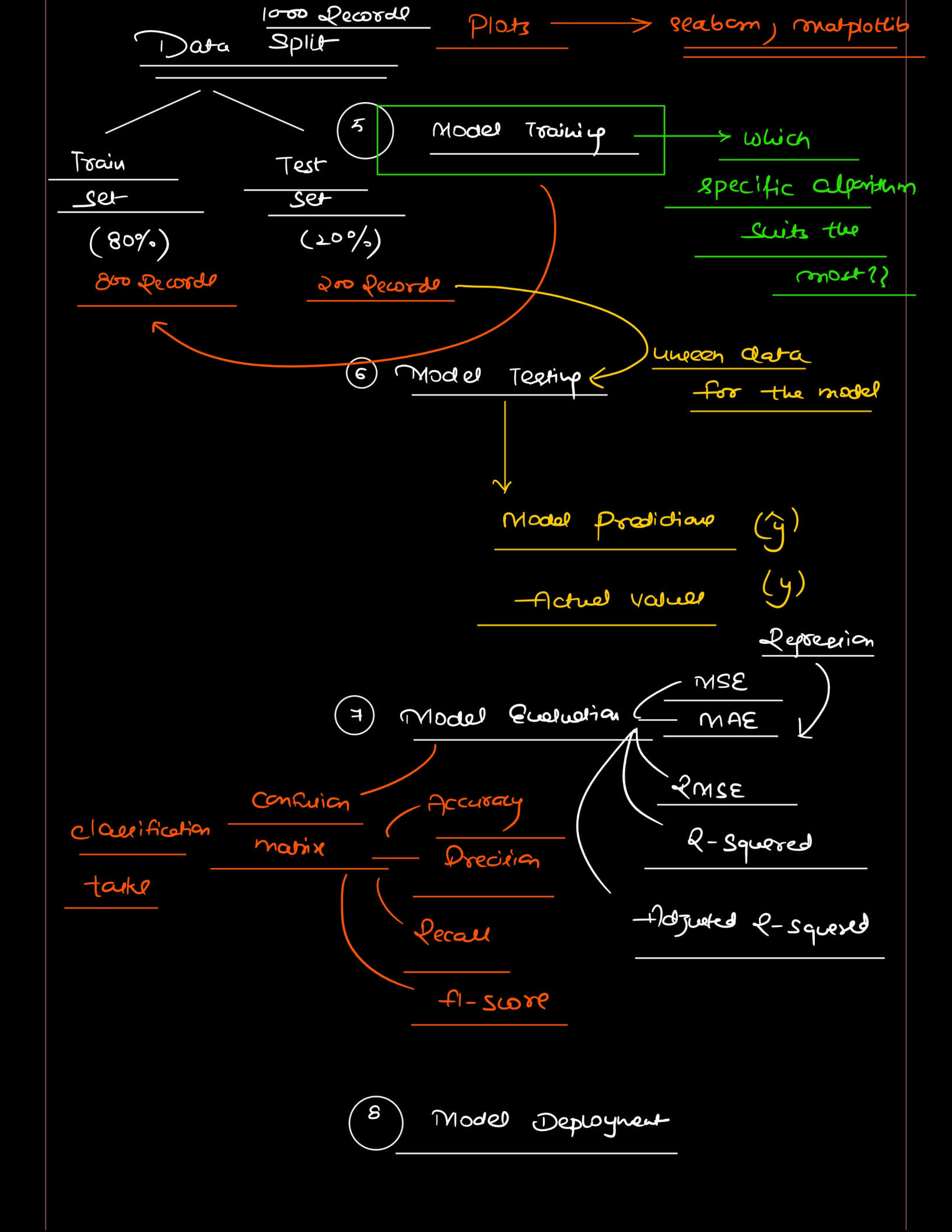
1. Business Problem understanding (Domain Expertise)

Machine Learning Project Pipeline / Workflow

2. Data ingestion / Collection → SQL (Raw data)
3. Data understanding → Rows (Records), Column (Features)

80-90% time → 4. EDA (Exploratory Data Analysis) (Statistics)

- Data Normalization / Standardization
- Outliers → deal
- Missing Values → Imputation
- Quality data
  - 1. Remove
  - 2. Focus on what kind of Cost function
  - 3. To get the best optimal Value
- Visualization → Correlation → Heatmap
- feature Selection
- feature engineering


1000 Records
Data Split
Plots → seaborn, matplotlib
Train set (80%) 800 Records
Test set (20%) 200 Records
5 Model Training → Which specific algorithm suits the most??
6 Model Testing
Unseen data for the model
Model Prediction (ŷ)
Actual values (y)
Regression
7 Model Evaluation
MSE
MAE
RMSE
R-Squared
Adjusted R-squared
Classification task
Confusion matrix
Accuracy
Precision
Recall
f1-score
8 Model Deployment

Data Standardization

Biased Model

2

2500

'MedInc', 'HouseAge', 'AveRooms', 'AveBedrms', 'Population', 'AveOccup', 'Latitude', 'Longitude'

data will be within one specific range

↳ model to be unbiased

① Standard Scaler → Standard normal form

↳ $\mu = 0$

$\sigma = 1$

② Min-max Scaler

fit_transform → train data

(μ, σ)

$z\text{-score} = \frac{x_i - \mu}{\sigma}$

transform → test data

(μ, σ)

Age	SNF
23	
45	
67	
49	
62	
22	

fit()

↳ μ, σ

transform()

↳ $z\text{-score} = \frac{x_i - \mu}{\sigma}$

$$\hat{y} = m_1 x_1 + m_2 x_2 + \dots + m_8 x_8 + c$$

Coefficients = 8

intercepts = 1

MAE vs MSE

$$\frac{1}{n} \sum_{i=1}^{n} |y_i - \hat{y}_i|$$

Residuals

↳ more robust to outliers

$$\frac{1}{n} \sum_{i=1}^{n} (y_i - \hat{y}_i)^2$$

amplify error

$$RMSE = \sqrt{MSE}$$

ideal → house price

↳ 5 lakh

(high error

temp prediction in Celsius

RMSE = 2°C → may be acceptable

↳ 0.5 Lakh

↳ decent

10°C → poor Prediction

Lower RMSE → model prediction will be better