

# Practical Machine Learning

## Day 2: SEP23 DBDA

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# Agenda

- Machine learning Applications
- Domains in Automation
- Machine Learning Modelling
- Evaluating ML techniques

# AI VS Machine Learning VS Deep Learning



- Artificial Intelligence originated around 1950s
- AI represents simulate intelligence in machines
- AI is a subset of data science
- Aim is to build machines which are capable of thinking like humans

## Artificial Intelligence



- Machine Learning originated around 1960s
- Machine learning is the practice of getting machines to make decisions without being programmed
- Machine learning is a subset of AI & Data Science
- Aim is to make machines learn through data so that they can solve problems

## Machine Learning



- Deep Learning originated around 1970s
- Deep Learning is the process of using artificial neural networks to solve complex problems
- Deep Learning is a subset of Machine Learning, AI & Data Science
- Aim is to build neural networks that automatically discover patterns for feature detection

## Deep Learning

# Features of Machine Learning:

- Machine learning **uses data to detect various patterns** in a given dataset.
  - It can **learn from past data** and improve automatically.
  - It is a **data-driven technology**.
  - Machine learning is much **similar to data mining** as it also deals with a huge amount of data.
- Following are some key points that show the importance of Machine Learning:
    - **Rapid increment** in the production of data
    - **Solving complex problems**, which are difficult for a human
    - **Decision-making in various sectors** including finance
    - Finding **hidden patterns and extracting useful information** from data.

# What is Machine Learning Model?

- **Definition:**

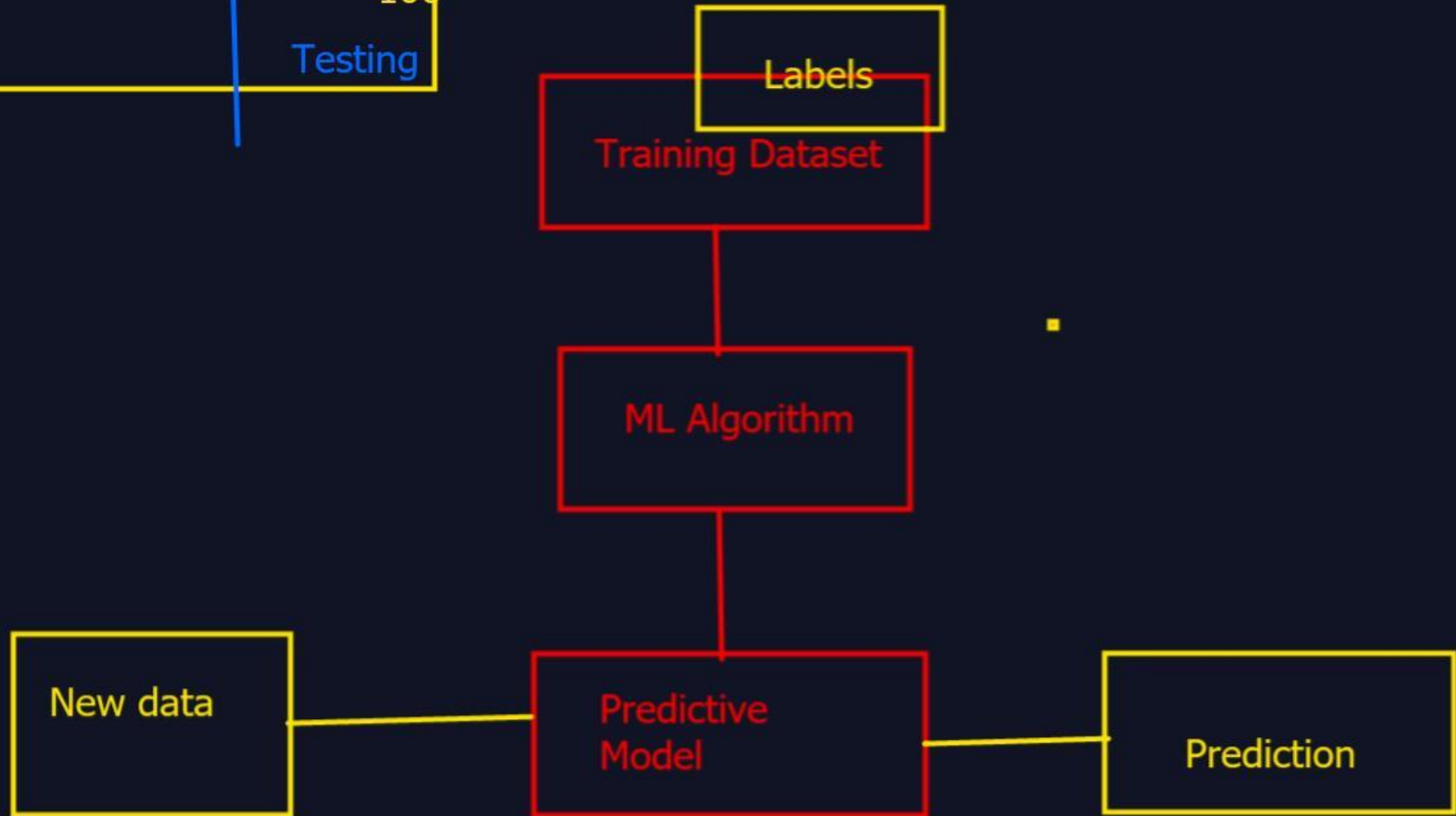
- Machine Learning is a concept which allows the machine
  - **to learn from examples and experience,**
  - and that **too without being explicitly programmed.**

- Machine Learning algorithms are an evolution of **normal algorithms.**
- They make your **programs “smarter”,** by **allowing them to automatically learn** from the data you provide.
- The algorithm is mainly divided into:
  - **Training Phase**
  - **Testing phase**



# 4. Steps in Python Machine Learning

- We follow the following steps in Machine Learning Using Python-
- Collecting data.
- Preprocessing data.
- Learning data.
- Training phase for Evaluation.
- Testing phase for Prediction.
- Using algorithms for future predictions.



# Training Phase

- You take a randomly selected specimen of mangoes from the market (**training data**),
- make a table of all the **physical characteristics** of each mango,
  - like **color, size, shape, grown** in which part of the country,
  - **sold by which vendor**, etc (**features**),
  - along with the **sweetness, juiciness, ripeness** of that mango (**output variables**).
- You feed this data to the machine learning algorithm (**classification/regression**), and it learns a model of the correlation between an average mango's physical characteristics, and its quality.

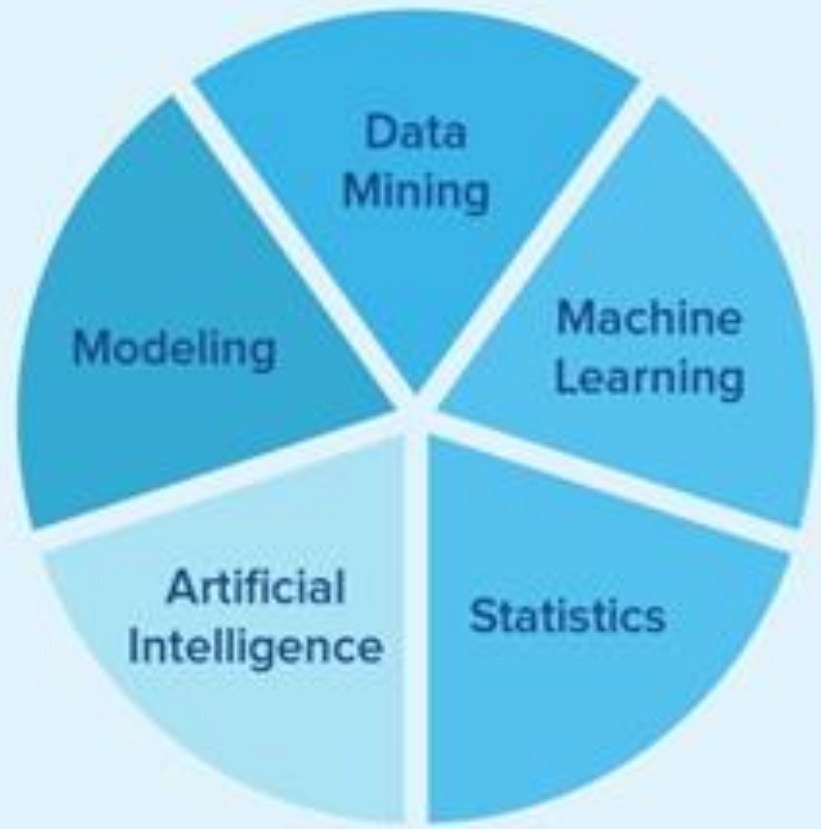


# Testing Phase

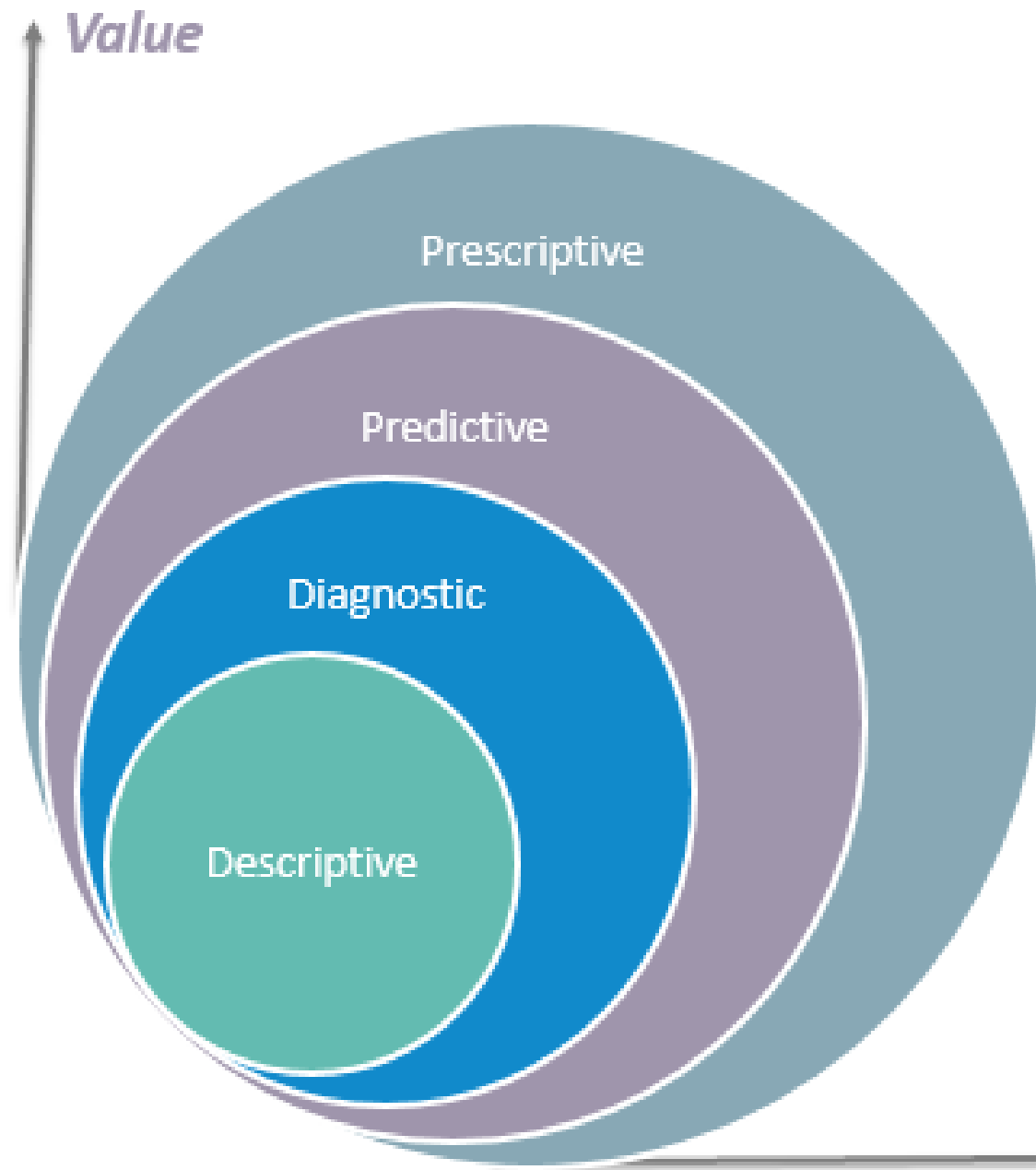
- Next time when you go shopping, you will **measure the characteristics of the mangoes** which you are purchasing(**test data**)and **feed it to the Machine Learning algorithm**.
- It will **use the model** which was **computed earlier to predict**
  - if the **mangoes are sweet, ripe and/or juicy**.
- The algorithm may internally **use the rules**, similar to the one you manually wrote earlier (for eg, a **decision tree**).
- Finally, you can now shop for mangoes with great confidence, without worrying about the details of how to choose the best mangoes.

# Conclusion as an Algorithm

- You know what! you can make your algorithm **improve over time (reinforcement learning)** so that it **will improve its accuracy** as it gets trained on more and more training dataset.
- In case it makes a wrong prediction it will update its rule by itself.
- The best part of this is, you can use the **same algorithm to train different models.**
- You can create one each for predicting the quality of apples, grapes, bananas, or whatever you want.



# 4 types of Data Analytics



## What is the data telling you?

**Descriptive:** *What's happening in my business?*

- Comprehensive, accurate and live data
- Effective visualisation

**Diagnostic:** *Why is it happening?*

- Ability to drill down to the root-cause
- Ability to isolate all confounding information

**Predictive:** *What's likely to happen?*

- Business strategies have remained fairly consistent over time
- Historical patterns being used to predict specific outcomes using algorithms
- Decisions are automated using algorithms and technology

**Prescriptive:** *What do I need to do?*

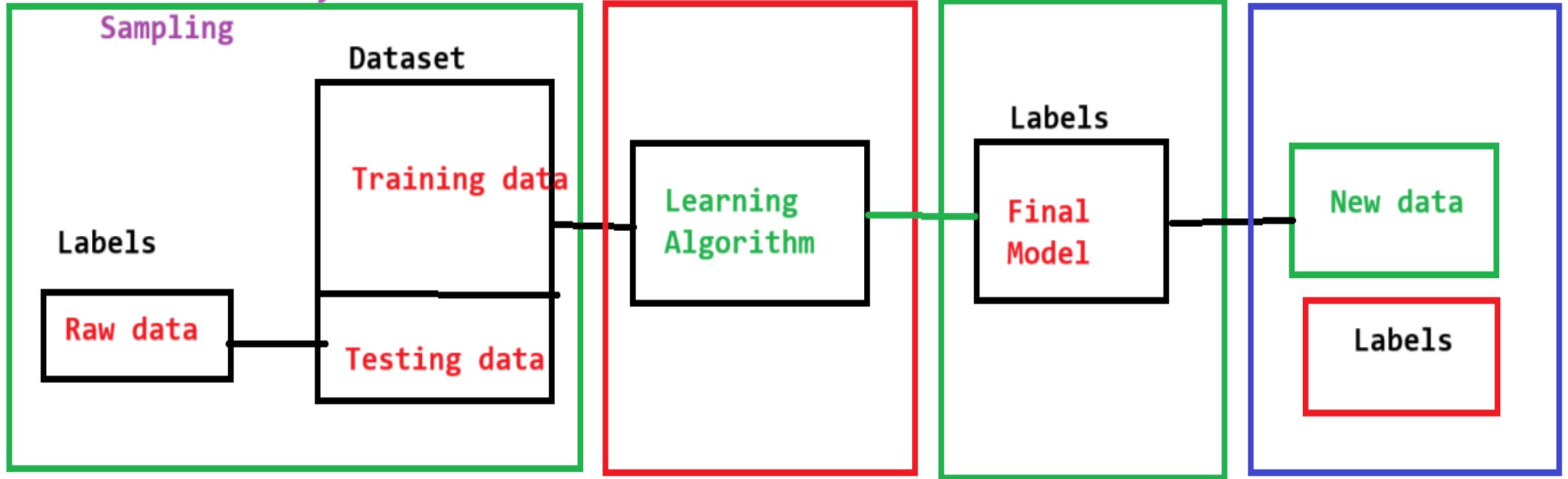
- Recommended actions and strategies based on champion / challenger testing strategy outcomes
- Applying advanced analytical techniques to make specific recommendations

**Complexity**

# Challenges and Limitations of Machine Learning

- The primary challenge of machine learning is the lack of data or the **diversity in the dataset**.
- A machine cannot learn if there is **no data available**.
- Besides, a dataset with a lack of diversity gives the machine a hard time.
- A machine **needs to have heterogeneity** to learn meaningful insight.
- It is rare that an algorithm can extract information when there are no or few variations.
- It is recommended to have at least 20 observations per group to help the machine learn.
- This constraint leads to poor evaluation and prediction.

Feature Extration and scaling  
Feature Selection  
Dimensionality Reduction  
Sampling



Preprocessing

Learning  
Model selection  
Cross-validation  
Performance evaluation

Evaluation

Prediction