

# Day 1 & 2: Modeling Overview and Industry Applications



#### **Session Overview**

#### Modeling Overview and Industry Applications

- This is an introductory session on modeling overview, which will help to understand the application of modeling in different aspects of the business domain like:
  - Insurance
  - Banking
  - Healthcare
  - Procurement
- In this session, we will cover modeling terminology and get to know each term in a nutshell to understand the different types of modeling such as:
  - Small data and large data
  - Use cases in various business domains
  - Model evaluation metrics





#### **Agenda/Session Outcomes**

Modeling Overview and Industry Applications

On completion of the session, you will be able to

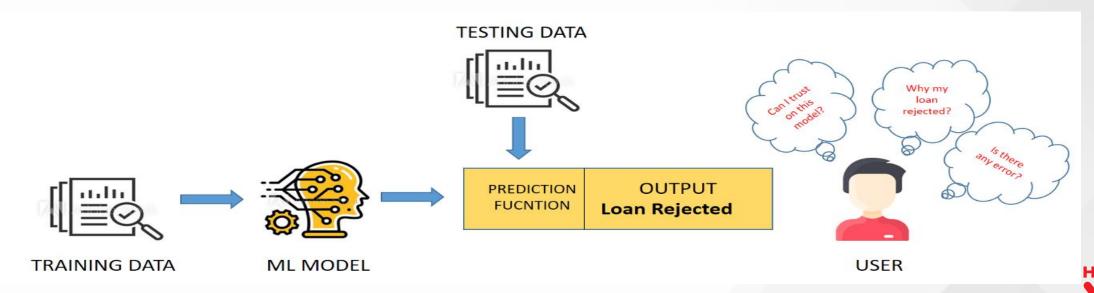
- Describe statistical models
- Discuss about machine learning and large data sets
- Identify different types of machine learning supervised, unsupervised, reinforcement learning
- Describe the real-world applications of machine learning
- Identify the methods used in machine learning such as train, test, validation, prediction etc.





#### **Machine Learning Models**

- Assume there is an activity in a bank to determine whether a customer is eligible for a loan or not.
  To examine and give the approval, bank employees were entrusted to do it manually. Now, to
  make the process faster, we are moving to data-driven decision-making algorithms, which are
  called models.
- Previously, bank employees handled loan processing, but now they rely on machine learning models, which are a set of algorithms that learn patterns in pre-historical customer records and predict whether a new customer is eligible for a loan.





#### What Is a Model?

Assume you want a machine to determine whether a patient is diabetic. You must collect many observations, including patient records and a variety of data. Let's have a look at the data.

Pregnancies	Glucose	${\sf BloodPressure}$	${\bf Skin Thickness}$	Insulin	BMI	DiabetesPedigreeFunction	Age	Outcome
6	148	72	35	0	33.6	0.627	50	1
1	85	66	29	0	26.6	0.351	31	0
8	183	64	0	0	23.3	0.672	32	1
1	89	66	23	94	28.1	0.167	21	0
0	137	40	35	168	43.1	2.288	33	1
5	116	74	0	0	25.6	0.201	30	0
3	78	50	32	88	31	0.248	26	1
10	115	0	0	0	35.3	0.134	29	0
2	197	70	45	543	30.5	0.158	53	1

Now, if the model is learning these several patterns in a random manner, then the algorithm will be able to predict whether, under the specified conditions, the patient will have diabetes or not. And it becomes very easy for the healthcare department to make data-driven decisions.





#### Implementation of the Modeling Techniques

- These modeling techniques are implemented in Python. But this can also be developed in other languages too.
- The reason we prefer Python to implement is because it is a popular and general-purpose programming language.
- We can write machine learning algorithms using Python, and they work well.
- Python is popular among data scientists is because it has a diverse variety of modules and libraries already implemented that make our lives more comfortable..



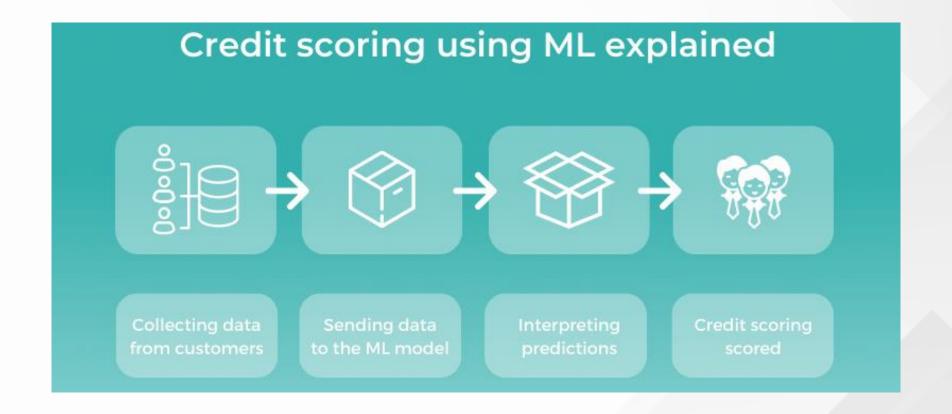


# Some Recent Applications Of Modeling Techniques



#### **Machine Learning Models**

- Analytics is one of the biggest advantages of machine learning technology in the financial industry. And the fields of decision-making and credit scoring are getting the most benefits from it.
- Banking institutions and other fintech firms use machine learning algorithms to perform their operations on the ML-based credit scoring system.

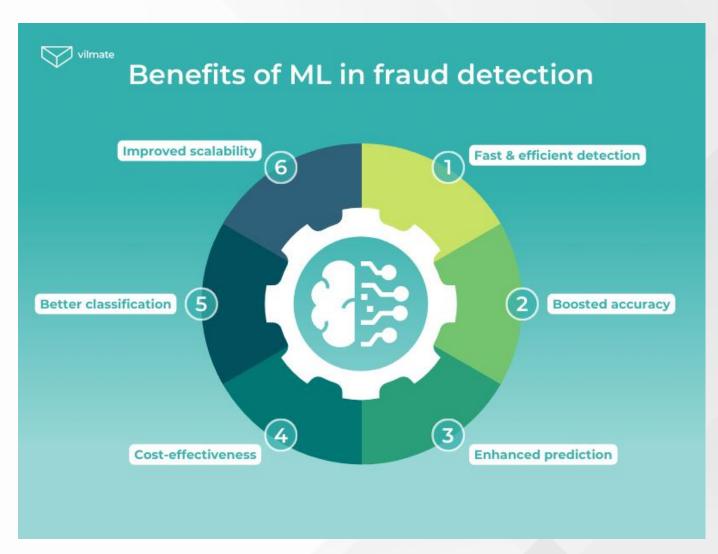






#### Security, Risk Management, and Fraud Detection

- The three biggest concerns for fintech companies are fraud, security, and safety.
- One should be perfectly aware of when and what to buy, what to sell, and when to hold in the market. The analytics feature in AI and ML makes these tasks a lot easier.







#### **Quantitative and Algorithmic Trading**

Building statistical models based on historical data can lead to more profitable investments faster than competitors.



#### Main applications of ML in trading

#### STRATEGY EVALUATION

ML facilitates the process of predicting outcomes by looking for anomalies, patterns, and correlations in huge data

DATA MINING

#### RISK MANAGEMENT

Algorithms based on machine learning enhance risk detection and efficiently target particular risk factors, i.e., value or volatility.

#### STRATEGY AGGREGATION

ML has great potential in aggregating individual signals from many sources into a fully-fledged strategy that can forecast future asset returns.

Back-testing (or testing of historic data) is an important stage in algorithmic trading.

#### STRATEGY REFINEMENT

Algorithmic strategies require extensive scientifically-backed testing based on ideas' performance in various out-of-sample market settings.

#### TRADE IDEAS CREATION

Thanks to deep market research, algorithms can provide specific long and short trading ideas.

#### **ASSETS ALLLOCATION**

ML can efficiently allocate assets by computing a hierarchical form of ris parity using decision-tree models.





#### **Statistical Models**

- Statistical modeling refers to the data science process of applying statistical analysis to datasets.
- A statistical model is a mathematical relationship between one or more random variables and other non-random variables.

# Statistical Modeling



Uses **finite** data sets and <u>reasonable number</u> of observations.



#### Why Statistical Model Is Used on Small Data?

- Statistical models work with finite data sets and a reasonable number of observations.
- Increasing the data might lead to overfitting (when statistical models fit against their training data).
- On the contrary, machine learning models need vast amounts of data to learn and perform intelligent actions.

#### **Small Dataset Example**

Height	Hair	Eyes	Class
short	blonde	blue	+
short	dark	blue	-
tall	dark	brown	-
tall	blonde	brown	-
tall	dark	blue	-
short	blonde	brown	-
tall	red	blue	+
tall	blonde	blue	+





#### When Should You Use Statistical Modeling?

- When data volume isn't too large.
- Errors and uncertainties in prediction are reasonable.
- Independent variables have fewer, pre-specified interactions.
- When it requires high interpretability.
- When it requires the isolation of the effects of a small number of variables.







#### When Should You Use Statistical Modeling?

#### **Example:**

- If a content marketing agency wants to build a model to track an audience's journey, it will prefer a statistical model because, rather than accuracy, the interpretation is important.
- It would help to develop an engagement strategy based on business domain knowledge.

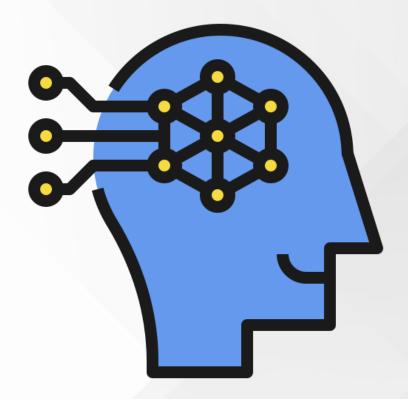






#### **Machine Learning and Large Data**

- Machine learning is a method of data analysis that automates the building of analytical models.
- It is a branch of artificial intelligence based on the idea that systems can learn from data, identify patterns, and make decisions with minimal human intervention.

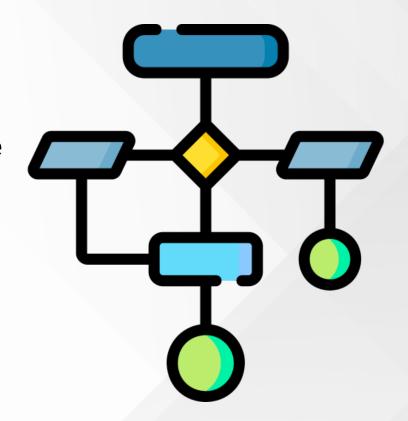






#### **Decision-Making and Credit Scoring**

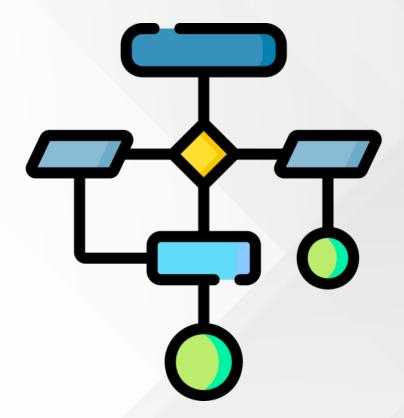
- Analytics is one of the biggest advantages that machine learning technology provides in the financial industry. And the fields of decision-making and credit scoring are getting the most benefits from it.
- Banking institutions & other fintech firms use machine learning algorithms to render their operations on the ML Based credit scoring system.







- Machine learning models are computer programs that are used to recognize patterns in data or make predictions.
- Models are created from algorithms, which are trained using either labeled, unlabeled, or mixed data.
- Different machine learning algorithms are suited to different goals, such as classification or prediction modeling, so data scientists use different algorithms as the basis for different models.

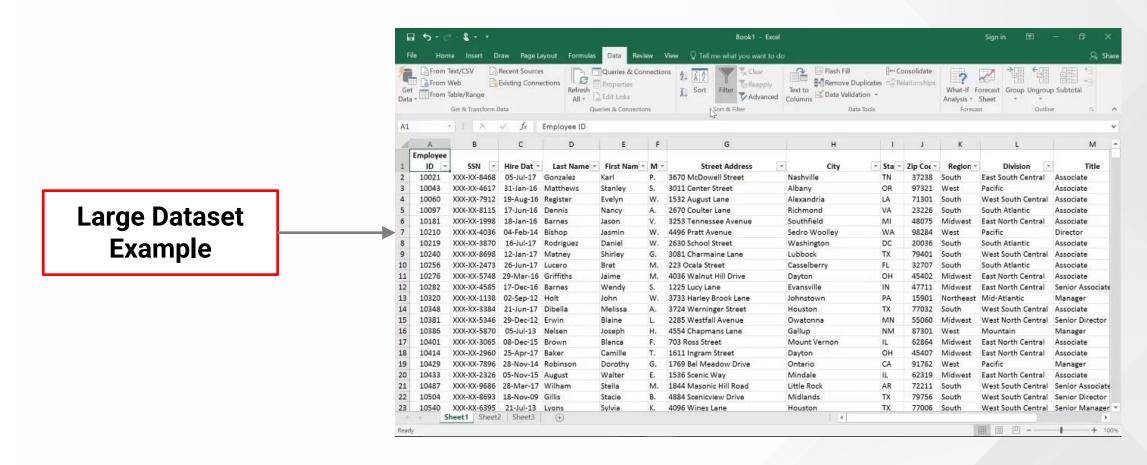






# Why Is a Machine Learning Model Applied to Large Amounts of Data?

More complex algorithms always require a larger amount of data. If your project needs standard ML algorithms that use structured learning, a smaller amount of data will be enough. Even if you feed the algorithm more data than it needs, the results won't improve drastically.

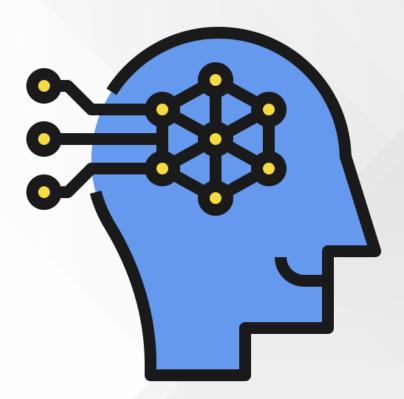






#### When Should You Use Machine Learning?

- Training learning algorithms on infinite data replications
- The goal is to get overall predictions, not relationships between variables.
- Estimating uncertainties in forecasts isn't essential.
- The effect of any variable doesn't need to be isolated.







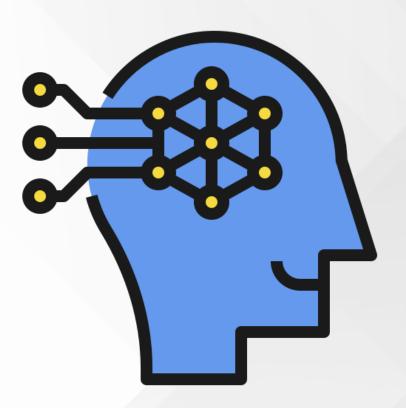
#### When Should You Use Machine Learning?



#### **Example:**

When an e-commerce website such as Amazon wants to recommend products based on previous purchases, they need a powerful recommendation engine. Here, the need for predictive accuracy is more important than the model's interpretability.

In this case, we use a machine learning model.

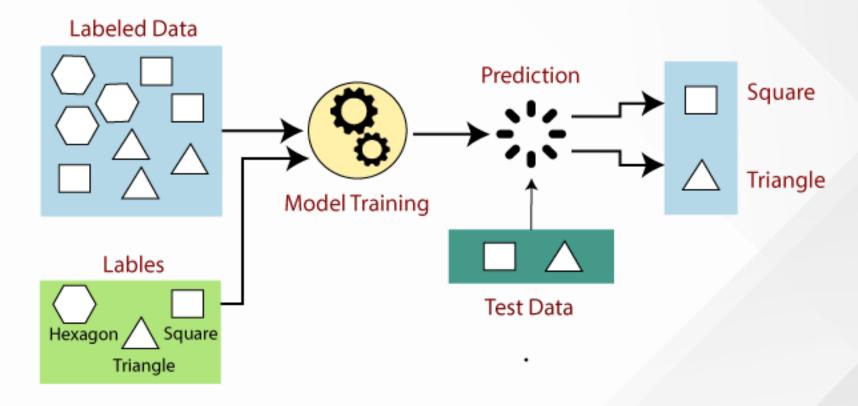






#### **Introduction to Machine Learning**

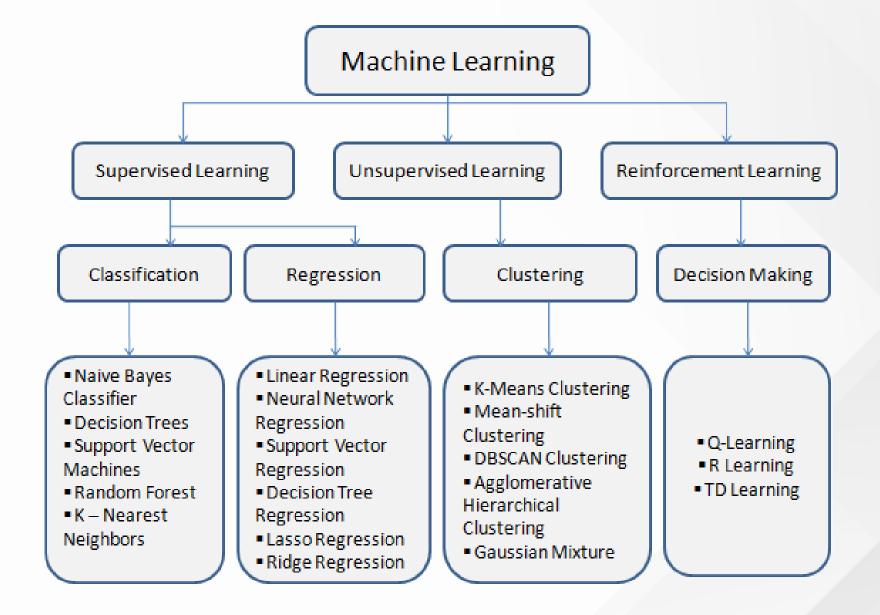
- Machine learning is a subfield of artificial intelligence (AI).
- Machine Learning aims to understand the structure of data and fit that data into models that can be understood and utilized by people.







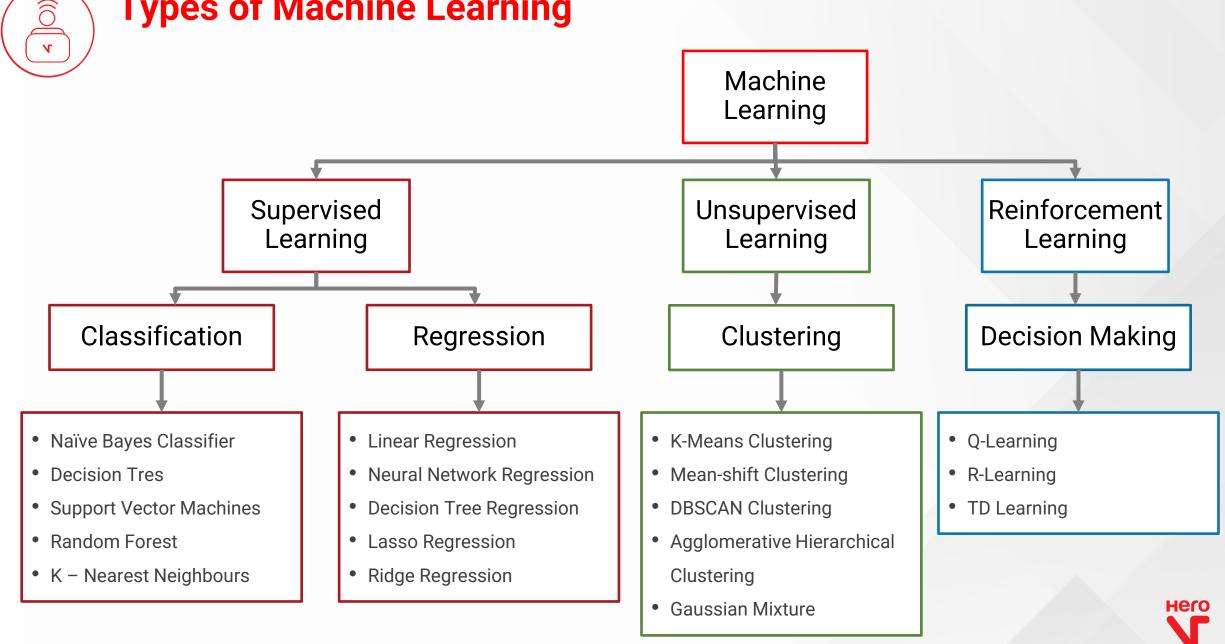
#### **Types of Machine Learning**







#### **Types of Machine Learning**





#### **Industry Specific Applications - Machine Learning**

#### **Machine Learning Applications in Finance**

 Analysts use machine learning in the financial services sector to automate trading activities, detect fraud, and provide financial advising services to their clients.

#### **Machine Learning Applications in Business**

Machine learning offers businesses an extensive number of ways to boost their effectiveness,
efficiency, and offerings. Chatbots, for example, allows businesses to provide faster, more flexible
customer service without employing a call center or making customers wait on hold for the next
available representative.

#### **Machine Learning Applications in Retail**

 As we've already mentioned, machine learning can be incredibly helpful in understanding and decreasing customer churn (i.e., the rate at which a business loses customers each year), which is a large focus point for many retail companies.





#### **Real-World Examples: Machine Learning**

#### **Image recognition**

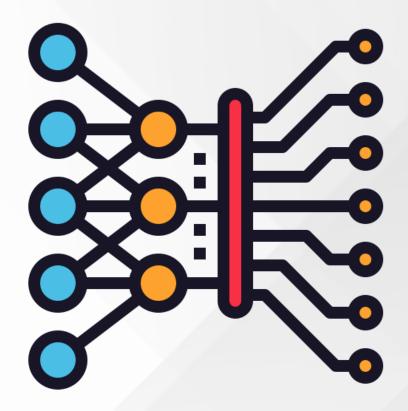
Image recognition is a well-known and widespread example of machine learning in the real world. It can identify an object as a digital image based on the intensity of the pixels in black-and-white or color images.

#### **Speech recognition**

Machine learning can translate speech into text. Certain software applications can convert live voice and recorded speech into a text file.

#### **Medical diagnosis**

Machine learning can help with the diagnosis of diseases. Many physicians use chatbots with speech recognition capabilities to discern patterns in symptoms.







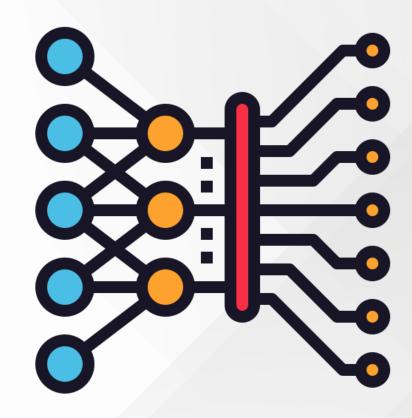
#### **Real-World Examples: Machine Learning**

#### Statistical arbitrage

Arbitrage is an automated trading strategy used in finance to manage a large volume of securities.

#### **Predictive analytics**

Machine learning can classify available data into groups, which are then defined by rules set by analysts.







What is the difference between supervised learning and unsupervised learning?

- 1. Dataset with target labels.
- Dataset without target labels.

Machine learning model requires small data in complex algorithms.

- 1. True
- 2. False

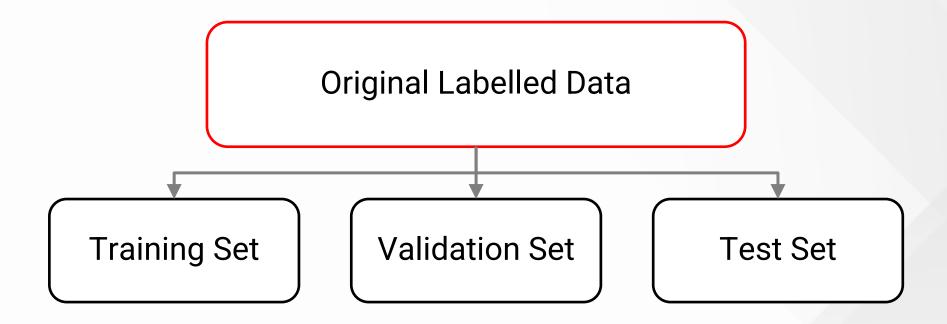
Statistical models work on \_\_\_\_\_ data/ observations.





#### **Methods of Machine Learning**

#### **Data Splitting Techniques**







#### **Methods of Machine Learning**

#### **Data Splitting Techniques**

- **Training set**: The data you will use to train your model will be fed into an algorithm that generates a model. This model maps inputs to outputs.
- Validation set: This is smaller than the training set and is used to evaluate the performance of
  models with different hyperparameter values. It's also used to detect overfitting during the training
  stages.
- Test set: This set is used to get an idea of the final performance of a model after hyperparameter tuning. It's also useful to get an idea of how different models (SVMs, neural networks, random forests, etc.) perform against each other.

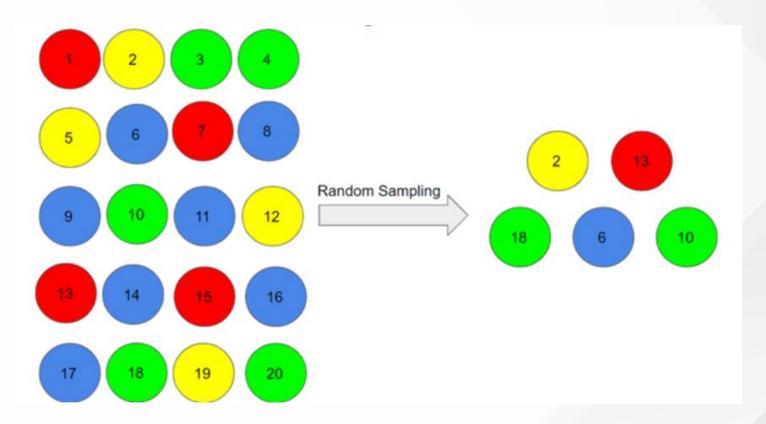




#### **Random Selection**

#### **Random Selection**

A random sample is a subset of a statistical population, whereas the subset should represent the entire population accurately and equally. To ensure that the sample is random, it is important that the population from which the data comes is stable.



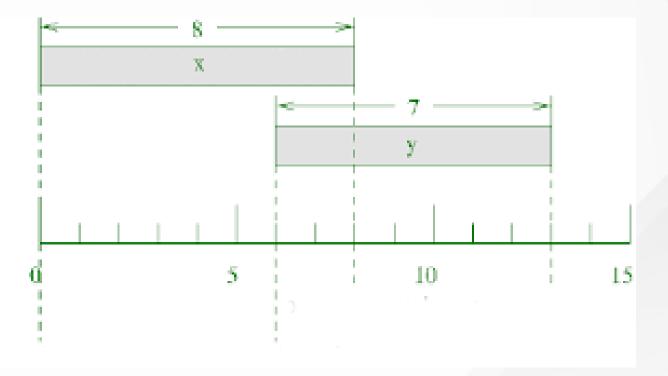




#### **Temporal Selection**

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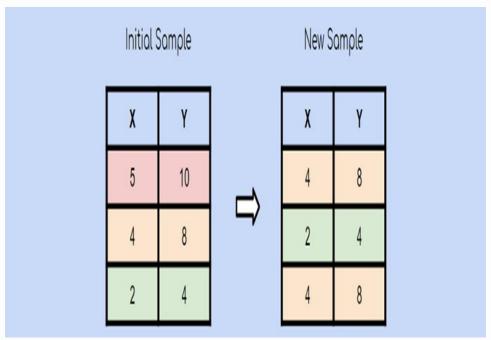
Temporal data is simply data that represents a state in time, such as the land-use patterns of Hong Kong in 1990 or the total rainfall in Honolulu on July 1, 2009. Temporal data is collected to analyze weather patterns and other environmental variables, monitor traffic conditions, study demographic trends, and so on.



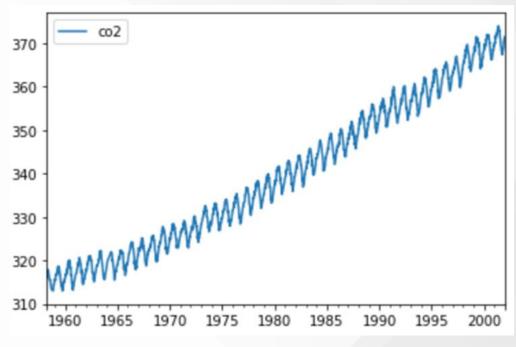




# Understanding Random and Temporal Selection in Machine Learning



Random Data Selection is a subset of data from the entire population that aids in inputting data into models with some random effect to understand the prediction in all possible ways.



Temporal data selection is time seriesdriven, which means it aids in the preparation of data sets that change over time.





#### Random Seeds in Machine Learning

A random seed is used to ensure that results are reproducible. In other words, using this parameter ensures that anyone who runs your code again will get the same results. Reproducibility is an extremely important concept in data science and other fields.

#### There are two common tasks where they are used:

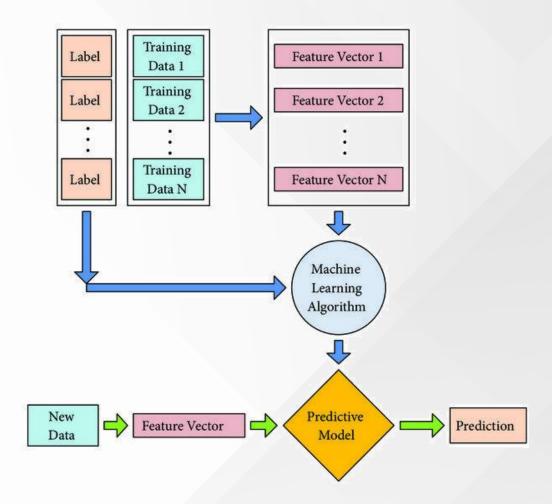
- Splitting data into training, validation, and test sets: Random seeds ensure that the data is divided
  the same way every time the code is run.
- **Model training**: Algorithms such as random forest and gradient boosting are non-deterministic (for a given input, the output is not always the same) and so require a random seed argument for reproducible results.





#### **Prediction in Machine Learning**

- Prediction essentially means to predict a future outcome, similarly to what is accomplished in machine learning.
- It refers to the output of an algorithm after training on a historical data set.
- After learning the historical data set, prediction makes it possible to forecast the outcome of a new data set.



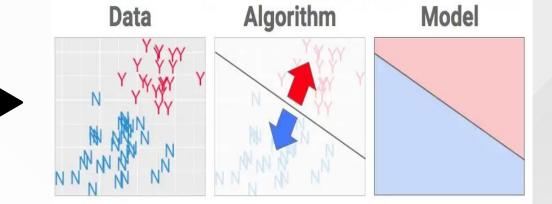




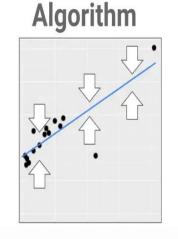
#### **Prediction on Data in Machine Learning**

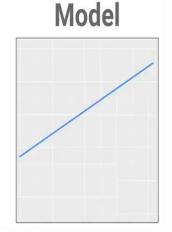
#### Classification

What we have here is a dataset that is labeled with two classes (Y and N). In other words, the algorithm must put a fence in the data to best separate the Ys from the Ns. Our goal is to create a model that can be used to classify new examples correctly.



# Data







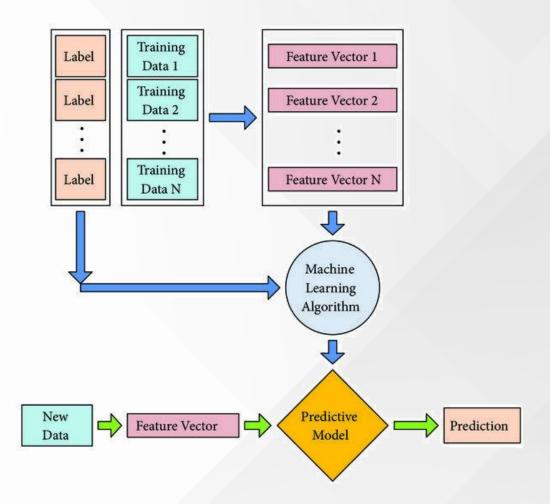
It is often used as a prediction where the model tries to predict any variables that are continuous in nature, such as demand forecasting, sales forecasting, etc.





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#### **Evaluation in Machine Learning**

Model evaluation is the process of using different evaluation metrics to understand a machine learning model's performance as well as its strengths and weaknesses.

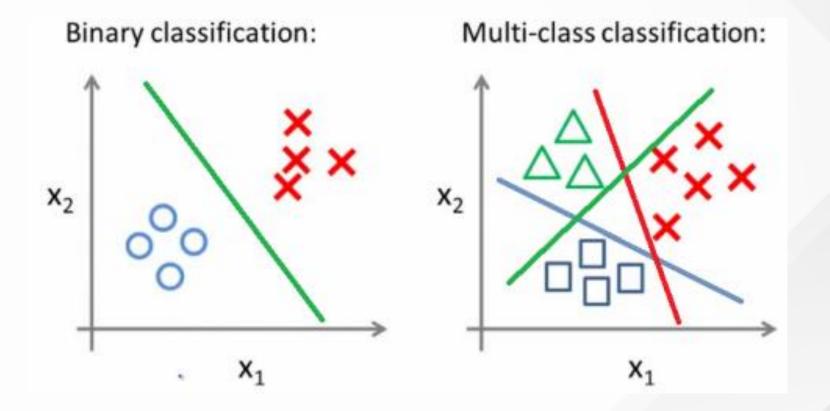






#### **Evaluating Multiclass Classifier Performance**

- Classification is the process of assigning any new data point to a set of categories (sub-populations) based on a mapping function.
- This function is derived from a training set of data where category membership is known for each observation.

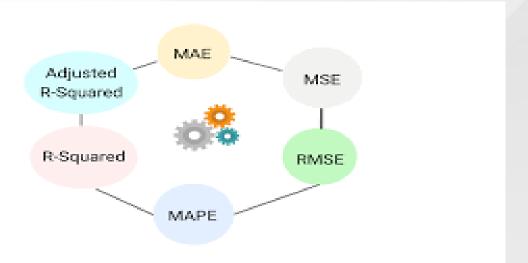






#### **Evaluating Regression Model's Performance**

- Regression refers to predictive modeling problems that involve predicting a numeric value. It is different from classification, which involves predicting a class label.
- Unlike classification, you cannot use classification accuracy to evaluate the predictions made by a regression model.
- Instead, you must use error metrics specifically designed for evaluating predictions made on regression problems.





# Quiz

A bank hires a data analyst to analyze the pre-historical dataset of customer records to determine whether the customers will churn. What will his or her modeling methodology be for the dataset?

- 1. Classification
- 2. Regression
- 3. Clustering
- 4. Reinforcement learning





A large retail store company wants to analyze the sales forecasting for the 52 outlets it has in the Northeast region. Each store generates approximately 3 GB of data per day; as a data scientist, what modeling technique would you choose if you were tasked with sales forecasting?

- 1. Statistical modeling
- 2. Deep learning and modeling
- 3. Modeling with Machine Learning and Deep Learning





### Thank You!

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