- Preface of Machine Learning
- Definition of ML by Tom. M. Mitchell
- What is ML?
- Difference between AI and ML
- Understanding ML by real life examples
- Need For Machine Learning
- Machine Learning Definitions
- Machine Learning Process
- How does Machine Learning Work?
- Machine Learning Types
- Supervised Learning
- Unsupervised Learning



Practical of Linear,
 Multiple and
 Polynomial
 Regression

Lecture-1,2,3,4,5



Supervised Learning

(Introduction to Decision Tree)

Dr. Virendra Singh Kushwah
Assistant Professor Grade-II
School of Computing Science and Engineering

Virendra.Kushwah@vitbhopal.ac.in

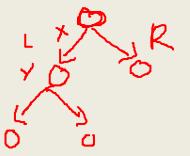
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Decision Tree

- A Decision Tree has many analogies in real life and turns out, it has influenced a wide area of Machine Learning, covering both *Classification and Regression*.
- •In decision analysis, a decision tree can be used to visually and explicitly represent decisions and decision making.





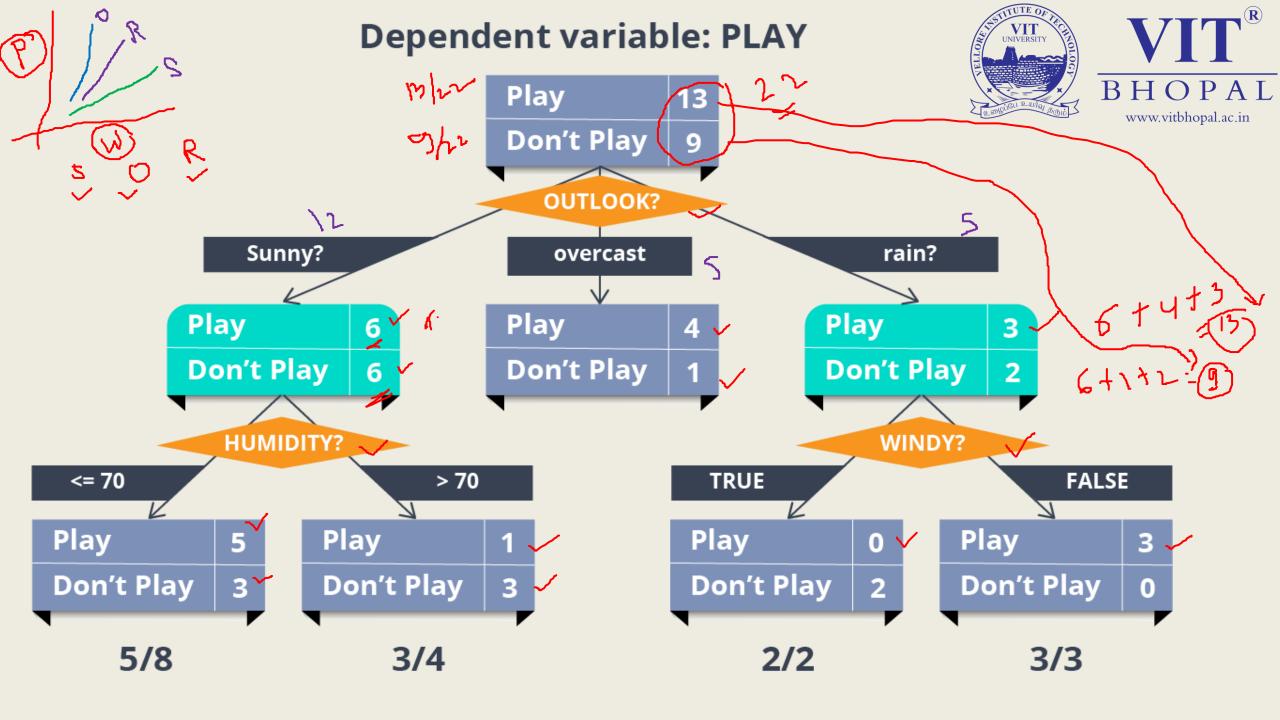
Root not

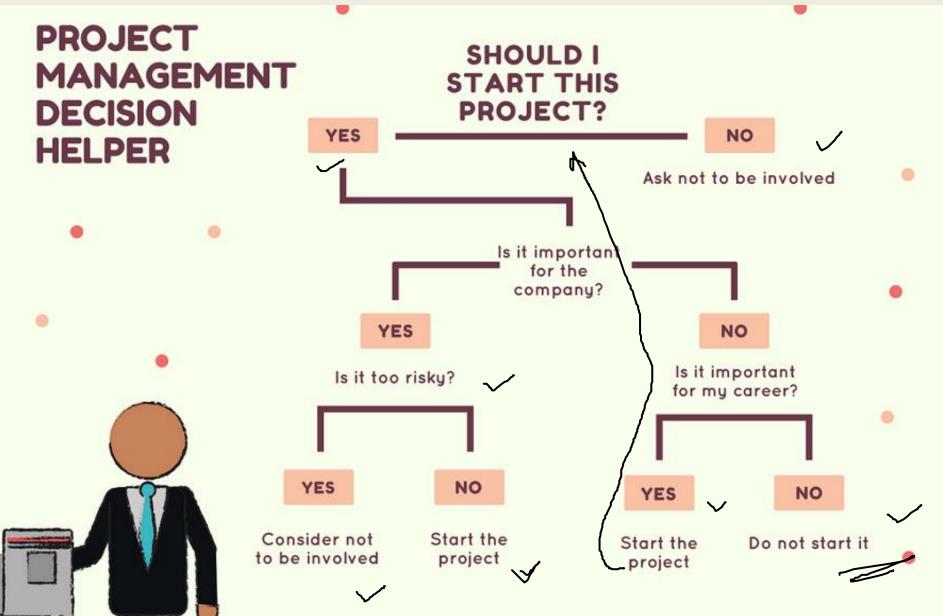


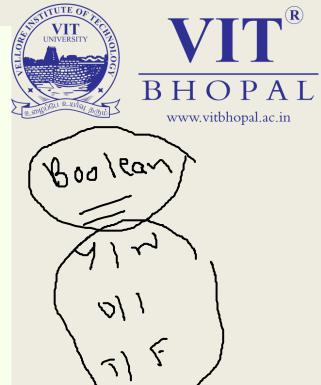


• A decision tree is a map of the possible outcomes of a series of related choices. It allows an individual or organization to weigh possible actions against one another based on their costs, probabilities, and benefits.

- As the name goes, it uses a tree-like model of decisions. They can be used either to drive informal discussion or to map out an algorithm that predicts the best choice mathematically.
- A decision tree typically starts with a single node, which branches into possible outcomes. Each of those outcomes leads to additional nodes, which branch off into other possibilities. This gives it a tree-like shape.













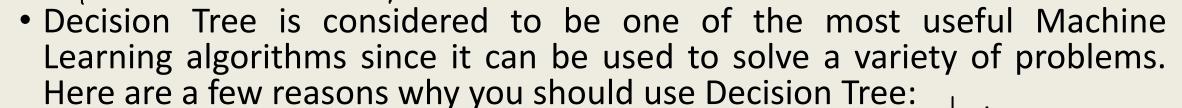


Why Decision Tree Algorithm?

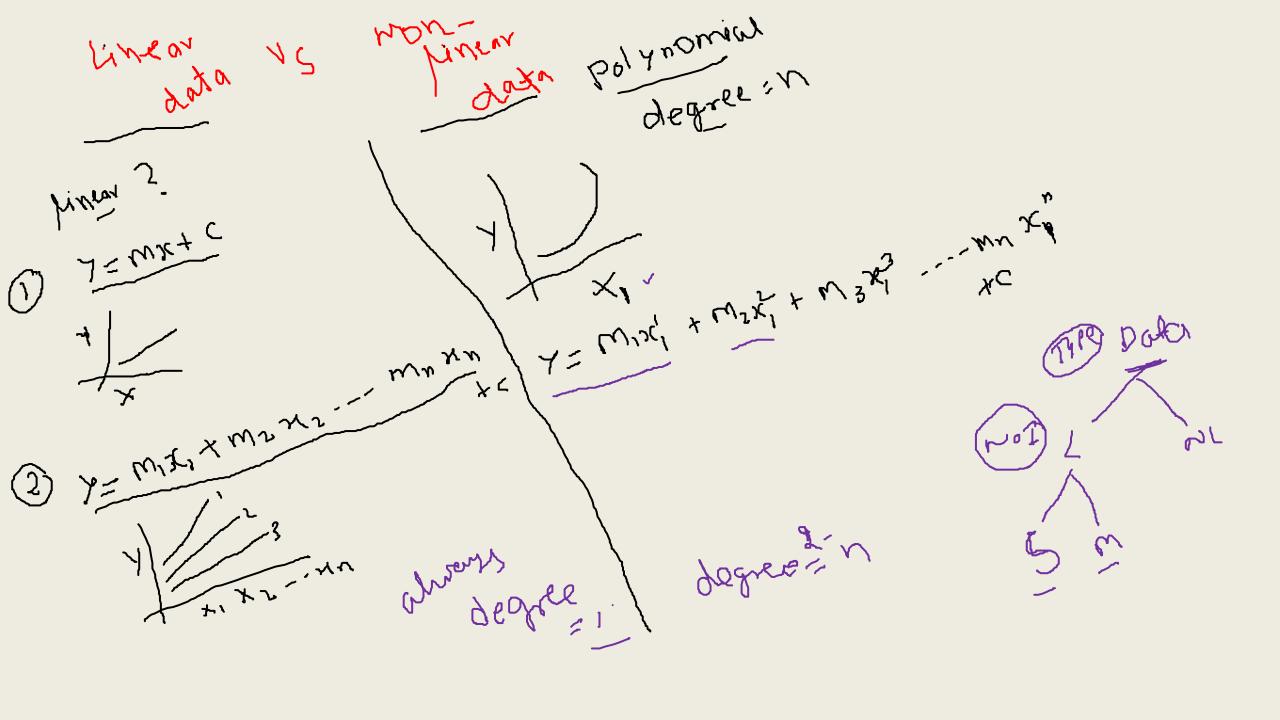








- 1. It is considered to be the most understandable Machine Learning algorithm and it can be easily interpreted.
- 2. It can be used for classification and regression problems.
- Unlike most Machine Learning algorithms, it works effectively with nonlinear data.
- Constructing a Decision Tree is a very quick process since it uses only one feature per node to split the data.



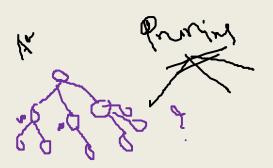
Advantages & Disadvantages of Decision Trees

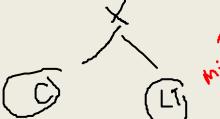


Advantages

- Decision trees generate understandable rules.
- Decision trees perform classification without requiring much computation.
- Decision trees are capable of handling both continuous and categorical variables.
 - Decision trees provide a clear indication of which fields are most important for prediction or classification.

Disadvantages









- Decision trees are less appropriate for estimation tasks where the goal is to predict the value of a continuous attribute.
- ✓ Decision trees are prone to errors in classification problems with many class and a relatively small number of training examples. ✓
- Decision trees can be computationally expensive to train. The process of growing a decision tree is computationally expensive. At each node, each candidate splitting field must be sorted before its best split can be found. In some algorithms, combinations of fields are used and a search must be made for optimal combining weights. Pruning algorithms can also be expensive since many candidate sub-trees must be formed and compared.



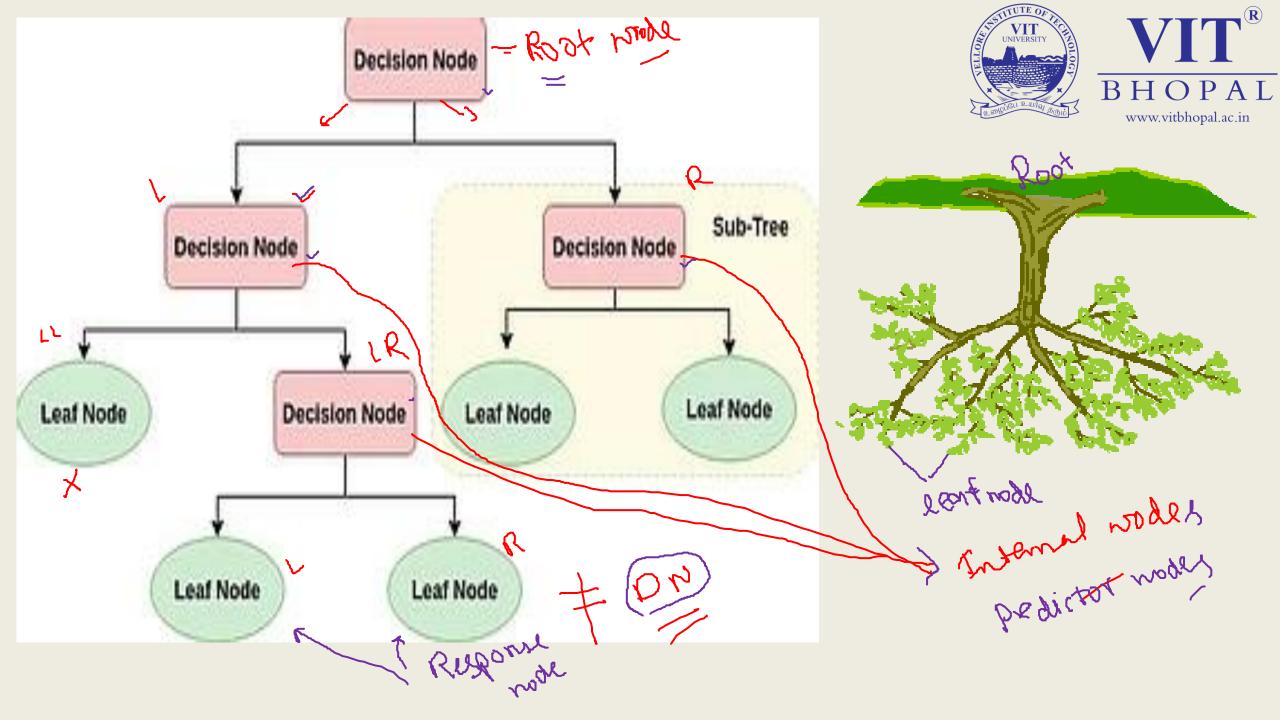
What Is A Decision Tree Algorithm?

• A Decision Tree is a Supervised Machine Learning algorithm which looks like an inverted tree, wherein each node represents a predictor variable (feature), the link between the nodes represents a Decision and each leaf node represents an outcome (response variable).

Bother Do Part



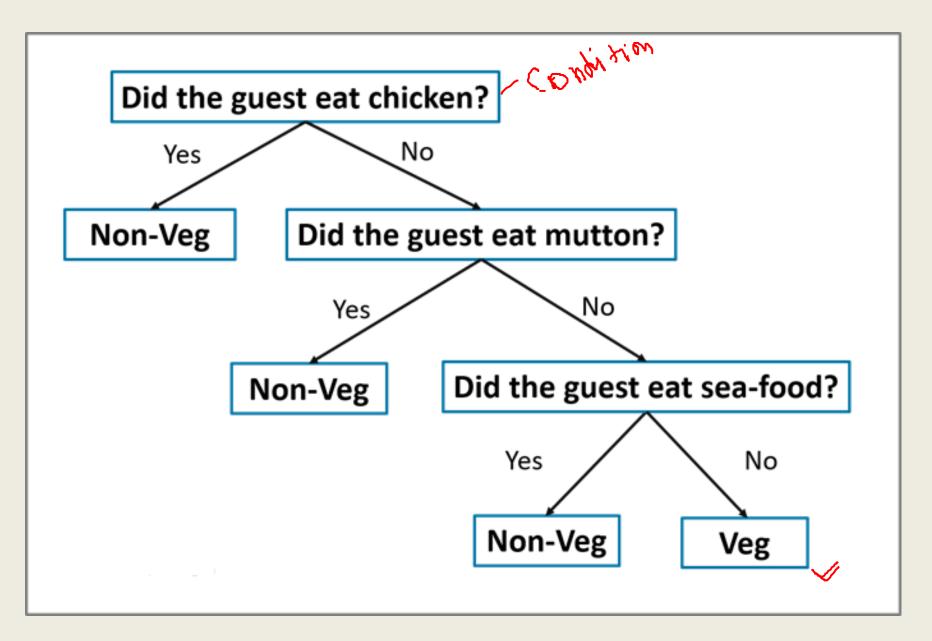
- A Decision Tree has the following structure:
- Root Node: The root node is the starting point of a tree. At this point, the first split is performed.
- Internal Nodes: Each internal node represents a decision point (predictor variable) that eventually leads to the prediction of the outcome.
- Leaf/ Terminal Nodes: Leaf nodes represent the final class of the outcome and therefore they're also called terminating nodes.
- Branches: Branches are connections between nodes, they're represented as arrows. Each branch represents a response such as yes or no.





Let us see an example

- Let's say that you hosted a huge party and you want to know how many of your guests were non-vegetarians. To solve this problem, let's create a simple Decision Tree.
- In the above illustration, I've created a Decision tree that classifies a guest as either vegetarian or non-vegetarian. Each node represents a predictor variable that will help to conclude whether or not a guest is a non-vegetarian. As you traverse down the tree, you must make decisions at each node, until you reach a dead end.











Structure Of A Decision Tree

