Introduction to Traditional Machine Learning

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Agenda

- Overview of Machine Learning
- Traditional Machine Learning vs. Deep Learning
- Key Concepts in Traditional Machine Learning
- Common Algorithms in Traditional Machine Learning
- Conclusion and Future Trends

Traditional Programming



Machine Learning





Overview of Machine Learning

- Definition of Machine Learning: The field of study that gives computers the ability to learn and improve from experience without being explicitly programmed.
- Three main types of Machine Learning:
 - Supervised Learning
 - Unsupervised Learning
 - Reinforcement Learning

Traditional Machine Learning vs. Deep Learning

Traditional Machine Learning:

- Relies on manually engineered features
- Uses statistical algorithms to make predictions
- Suitable for structured data and smaller datasets

Deep Learning:

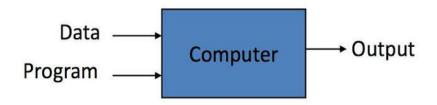
- Learns features automatically from raw data
- Utilizes deep neural networks for prediction
- Performs well on unstructured data and large datasets

Key Concepts in Traditional Machine Learning

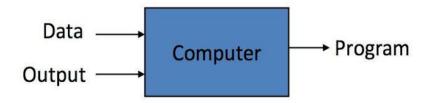
- Features: Representations of the input data used for learning and prediction.
- Training Data: Examples used to train the machine learning model.
- Labels: Known outputs corresponding to the training data used for supervised learning.
- Model: Mathematical representation of the relationship between inputs and outputs.
- Loss Function: Measures the model's performance and guides the learning process.
- Optimization: The process of adjusting model parameters to minimize the loss function.

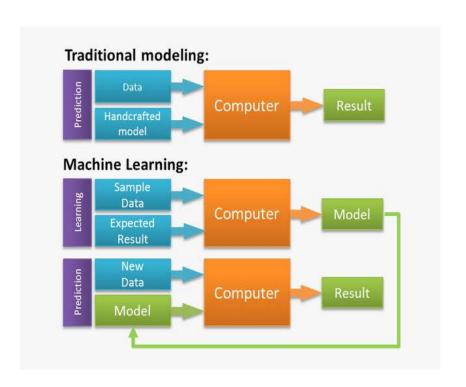
Visual Overview

Traditional Programming



Machine Learning





Common Algorithms in Traditional Machine Learning

- Linear Regression
 - Gradient descent
- Logistic Regression
- Decision Trees
- Random Forests
- Support Vector Machines (SVM)
- Naive Bayes
- k-Nearest Neighbors (k-NN)

Linear Regression

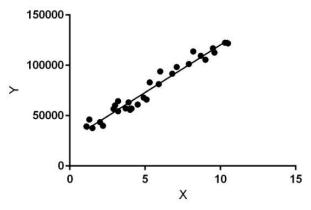
computes the linear relationship between a dependent variable and one or more independent features

Hypothesis function for Linear Regression:

$$\hat{Y} = \theta_1 + \theta_2 X$$
OR
 $\hat{y}_i = \theta_1 + \theta_2 x_i$

Cost function

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



Gradient Descent

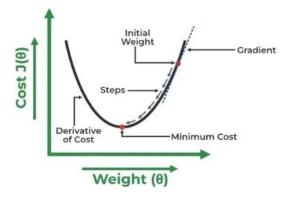
start with random $\theta 1$ and $\theta 2$ values and then iteratively update the values, reaching minimum cost

$$\theta_1 = \theta_1 - \alpha \left(J'_{\theta_1} \right)$$

$$= \theta_1 - \alpha \left(\frac{2}{n} \sum_{i=1}^n (\hat{y}_i - y_i) \right)$$

$$\theta_2 = \theta_2 - \alpha \left(J'_{\theta_2} \right)$$

$$= \theta_2 - \alpha \left(\frac{2}{n} \sum_{i=1}^n (\hat{y}_i - y_i) \cdot x_i \right)$$



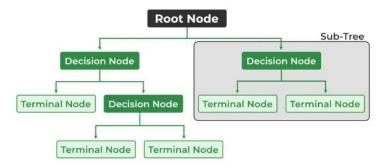
Decision Trees

- flowchart-like tree structure
- recursively splitting the training data into subsets based on the values of the attributes
- Entropy

measure of the degree of randomness or uncertainty in the dataset

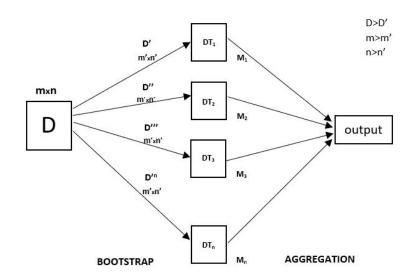
Gini Impurity or index

a score that evaluates how accurate a split is among the classified groups



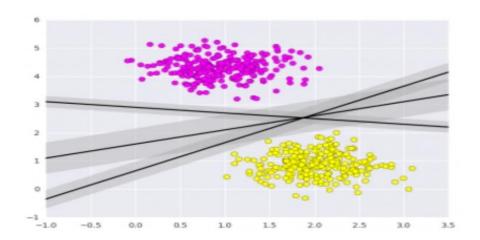
Random Forests

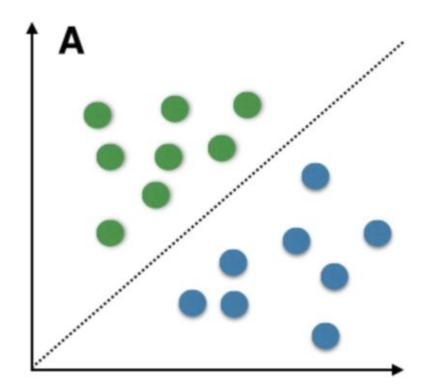
output doesn't depend on one decision tree but on multiple decision trees

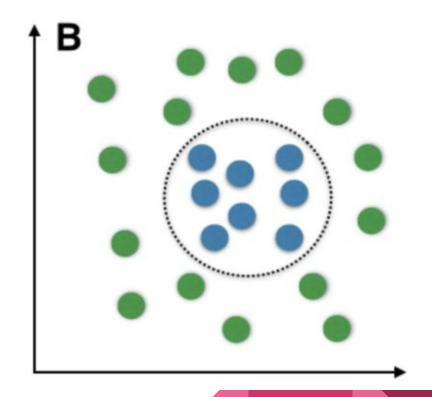


Support Vector Machine

Given a set of training examples, each marked as belonging to one or the other of two categories, an SVM training algorithm builds a model that assigns new examples to one category or the other

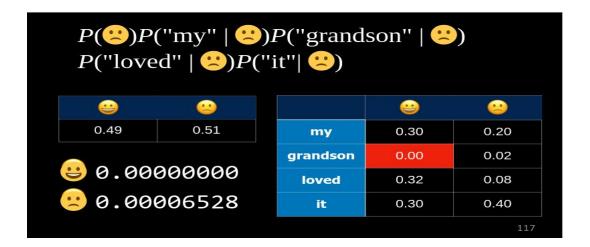






Naive Bayes

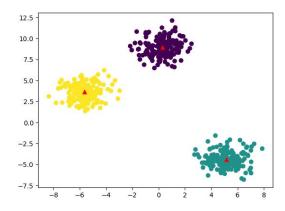
- Naive Bayes is a probabilistic classifier based on Bayes' theorem.
- It assumes that the features are conditionally independent given the class.
- The algorithm calculates the probability of an instance belonging to a particular class and predicts the class with the highest probability.



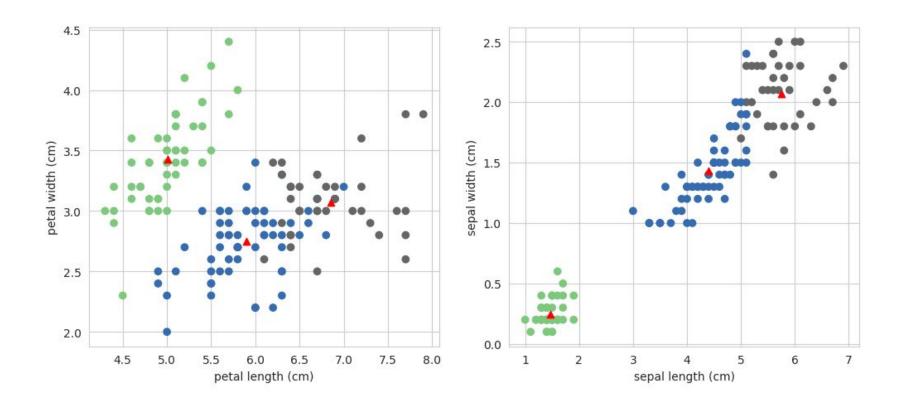
K-nearest Neighbour [Unsupervised]

an Unsupervised Machine Learning algorithm, which groups the unlabeled dataset into different clusters

- randomly initialize k points, called means or cluster centroids
- categorize each item to its closest mean
- and we update the mean's coordinates
- After some iterations, we have our clusters



A practical Example



Conclusion and Future Trends

- Traditional machine learning remains a valuable and widely used approach.
- Deep learning has achieved remarkable success in various domains.
- Future trends:
 - Combination of traditional and deep learning techniques.
 - Advances in interpretability and explainability of machine learning models.
 - Handling unstructured and multimodal data.
 - Development of efficient algorithms for large-scale machine learning.

Acknowledgments

- ChatGPT
- GeeksForGeeks

Any Questions???

Thank You!