



STUDYING ABOUT HISTORY, TYPES & APPLICATION OF MACHINE LEARNING TECHNIQUES

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ABSTRACT

The focus of many large corporations has shifted to AI and machine learning. Machine learning is getting a lot of attention and investment as people try to persuade others that the era of machine intelligence has arrived. Machine learning has sparked a technology revolution that is powering autonomous vehicles, virtual assistants, illness diagnosis, and treatment planning, to name just a few of its many uses. The purpose of this work is to provide philosophical and practical insights into the methodologies, methods, and algorithms of machine learning by reviewing the existing literature. As one of AI's more established subfields, machine learning investigates computer approaches to information management and knowledge discovery. Many fields of study have found success using machine learning techniques. However, in the recent years, new data have been accessible and, as a result, new sectors where machine learning may be used have emerged, thanks to different technical breakthroughs and research activities (such as the completion of the Human Genome Project and the expansion of the Web). Learning from biological sequences, email data, and complex ecosystems like the Web are just a few examples of these cutting-edge uses. In this work, we discuss these three application areas, along with some recent attempts that have used machine learning to evaluate data from these areas.

Keywords: - Machine Learning, Artificial Intelligence, Application, Supervised, Reinforcement

I. INTRODUCTION

Machine learning, a subfield of AI, supposedly gives computers the capacity to pick up new skills and refine their existing ones with little to no human guidance. In the field of machine learning, the goal is to create algorithms that can read and understand data on their own. To improve future

decision-making based on the examples we offer, the learning process starts with observations or facts like examples, direct experience, or teaching. The basic goal is to give computers the ability to learn on their own, without any help from humans, and then behave appropriately. Algorithms are employed in machine learning to determine which data patterns are significant and which are not. The provision of accurate medical diagnoses (for example, for breast cancer), real-time map-based monitoring of environmental catastrophes (for example, for forest fires), and sensory monitoring in the industrial process (for example, for mechanical failure) are all examples of uses of machine learning.

Artificial intelligence (AI) in the form of machine learning allows for computers to be taught without being explicitly programmed. The focus of machine learning is on the additions to software that can be modified when exposed to novel data. Machine learning's development parallels that of data mining. Data mining and machine learning both use comprehensive explorations of data to make inferences about underlying patterns. But unlike data mining tools, which harvest data for human understanding, machine learning makes use of the data to recognize patterns in the data and fine-tune program behaviors.

The fields of computational statistics and data science, which likewise focus on prediction, are closely connected to and frequently overlap with machine learning. Its close linkages to mathematical optimization provide it with tools, theory, and new areas of study. Machine learning is used for a variety of computer jobs where it would be impractical to create explicit methods via programming. Simply defined, machine learning refers to giving computers the capacity to learn, as suggested by the name. The goal of machine learning is to teach computers to analyze and draw conclusions from vast amounts of data with increasing precision. That's why we consider machine learning to be part of AI; it's really just a method for making AI a reality. It's a way to teach computers to make choices on their own. Training an algorithm in machine learning requires providing it with a large amount of data and giving it time to digest and absorb the knowledge.

II. THE HISTORY OF MACHINE LEARNING

Machine learning as we know it now was first used by Arthur Samuel in 1952. Frank Rosenblatt, working at the Cornell Aeronautical Laboratory in 1957, developed the perceptron by fusing Donald Hebb's concept of brain cell interaction with Arthur Samuel's research in machine learning. The first step in modern pattern recognition was taken in 1967 with the development of the closest neighbor algorithm. This technique was one of the early ones developed to solve the "best route" issue faced by road warriors like traveling salesmen.

The advent and use of multilayers in neural network research marked a significant turning point in the 1960s. It was found that the processing capability of a perceptron may be greatly increased by supplying and using two or more layers. Machine learning, so the story goes, developed as a direct

result of efforts to create artificial intelligence. Some early researchers in the field of artificial intelligence were intrigued by the idea of teaching computers new skills using existing datasets.

They tried to solve it using a wide range of symbolic techniques, including neural networks (primarily perceptrons and other models that were subsequently shown to be reimaginings of the generalized linear models of statistics). Automated medical diagnosis, in particular, made extensive use of probabilistic reasoning. After being separated off into its own discipline, machine learning saw rapid growth in the 1990s. The area shifted its focus from developing artificial intelligence to solving real-world issues. It abandoned the symbolic techniques it had acquired from AI in favor of statistical and probabilistic models.

III. TYPES OF MACHINE LEARNING

Categorized machine learning algorithms into supervised, unsupervised and reinforcement learning algorithms: Figure one present the classification in a pictorial form:

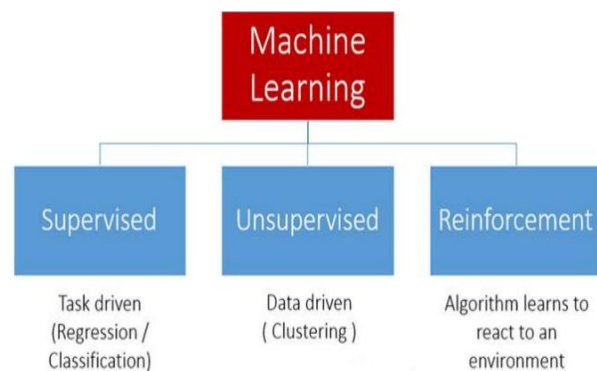


Figure 1: Types of Machine Learning.

1. Supervised Learning

Understanding concepts through seeing like ones is another name for this kind of instruction. Learning a notion or function that is truly a model description is what supervised learning is all about for the cognitive system. In specifically, the system is given a sample to work with. Each example's target function's output is also provided. Based on the function's output, the system must infer the model's description. A model is developed using a subset of the data (the training set) and then tested on the remaining data (the test set).

The field of machine learning relies heavily on supervised learning. The purpose of supervised learning is to acquire knowledge about how inputs relate to desired outcomes. The information sent into the system is descriptive of a set of examples or instances of the things of interest. A supervisor's output is the end result or conclusion reached. Classification is a subset of supervised learning in which the examples are divided into groups based on a mapping (or discriminant

function). The output, known as the class label in machine learning, defines the various categories. Classifiers and models are other names for the discriminant function. The phrase "training set" refers to a collection of examples that have already been labeled. Classification involves optimizing a model's parameters to provide a mapping between training set occurrences and training set labels. New, unseen examples may be labeled or classified using the trained model.

2. Unsupervised Learning

Observational learning is another name for this method. Without being told how many or if any patterns exist, the system in unsupervised learning must figure out what they are based solely on the shared characteristics of the examples.

This theory states that machine learning methods should be utilized if the training data lacks a clear classification or label. The field of unsupervised learning investigates how computers can figure out how to describe a secret structure without any labels. In order to characterize unlabeled data's underlying structures, the system doesn't choose the correct output, but it does explore the data and make conclusions from it.

Unsupervised learning occurs when there is no connection between the input data (X) and the desired outcomes. Learn more about the data by modeling its underlying structure or distribution, as is the case with unsupervised learning. Because there are no right answers and no instructor, this kind of learning is known as unsupervised learning. The onus is on the algorithms themselves to unearth and convey the data's intriguing structure. Clustering and association issues are subsets of unsupervised learning problems. Finding patterns in your data, like customers who exhibit similar buying habits, is an example of a clustering problem, while finding generalizable rules about your data, like "people who buy X also tend to buy Y," is an example of an association rule learning problem. For instance, the k-means method is a well-known unsupervised learning algorithm used to solve clustering issues, while the Apriority algorithm is used to solve association rule learning problems.

3. Reinforcement machine learning algorithms

A reinforcement machine learning algorithm is a kind of machine learning that learns from its mistakes and successes in the real world. Reinforcement learning's most important features are its trial-and-error search and delayed reward. Using this technique, machines and software agents may autonomously learn how to optimize their performance in a given setting. The reinforcement signal teaches the agent the optimal action by providing it with simple reward feedback. Reinforcement learning is the most straightforward area of machine learning to convey to individuals who are unfamiliar with the topic. To put it another way, Reinforcement Learning is like teaching your dog (or cat, if you like a challenge) to do tricks: if he performs the trick you want, you treat him with treats; if he doesn't, you don't treat him or give him lemons. Lemons are

a food that dogs really despise. Beyond debate, Reinforced learning is a more complex and challenging method to be realized, but at its core, it is concerned with learning by interacting with and receiving feedback from one's environment, or learning to solve a task through trial and error. In essence, an agent (or several agents) is developed with the capacity to detect, analyze, and learn from its surrounding environment.

IV. Applications of Machine learning

The field of machine learning is a hot topic in today's technology, and its popularity is only increasing. Without realizing it, we use machine learning every day in the form of services like Google Maps, Google Assistant, Alexa, etc. Some of the most popular uses of Machine Learning in the real world are listed below.

1. Image Recognition:

One of machine learning's most widespread uses is in image identification. Its purpose is to assign unique identifiers to things, people, locations, digital photos, etc. Automatic friend tagging recommendation: Facebook gives us a function of auto friend tagging suggestion, which is a common application of image recognition and facial detection.

The technology underlying Facebook's automatic tagging recommendation with name when uploading a picture with friends is Facebook's facial detection and recognition system, which relies on machine learning.

2. Speech Recognition

When using Google, we have the option to "Search by voice," which is classified as speech recognition and is a well-known example of how machine learning is put to use.

Computer speech recognition, also known as speech recognition or speech to text, is the process by which spoken commands are translated into text. Many voice recognition programs now use machine learning techniques. Assistants like Google's, Apple's, Microsoft's, and Amazon's all use speech recognition technology to act on user requests.

3. Traffic prediction:

If we want to visit a new place, we take help of Google Maps, which shows us the correct path with the shortest route and predicts the traffic conditions.

4. Product recommendations:

Amazon, Netflix, and many other online retailers and media providers employ machine learning to better propose products to their customers. Thanks to machine learning, we now get ads for the same goods we were looking for on Amazon whenever we use that browser to explore the web.

5. Self-driving cars:

Self-driving automobiles are one of the most fascinating uses of machine learning. The development of autonomous vehicles relies heavily on machine learning. The most well-known automaker, Tesla, is developing an autonomous vehicle. Vehicle models for traffic-related item and person detection are being trained via unsupervised learning.

V. CONCLUSION

The definition, development, varieties, models, tools, and real-world applications of machine learning are all covered in this article. Today's machine learning systems are showing that they can produce and finish complicated jobs with previously imagined efficiency and effectiveness. A machine learning system "learns" to estimate accurately by observing a sample of previously completed jobs. This article provides an overview of machine learning, its many applications, and the most popular applications software, including WEKA, Salford's Predictive Modeler, OpenCV, Torch, LIONSolver, NeuroSolution, KXEN Modeler, RapidMiner, Databricks, and H2O.

Although Web and biological data mining are anticipated to play an important role in the future of the field, this study does not cover all of the relevant aspects of this exciting sector. Imaging and astronomy data, robotics, gaming and music data, and so on are only a few examples. Research in machine learning has also recently shifted toward an emphasis on spatial and visual data. Machine learning has the potential to enhance the performance of artificial vision systems and provide light on how humans process and make sense of the world around them.

The fundamentals of machine learning, its three subfields—supervised, unsupervised, and reinforcement learning, and the most popular supervised and unsupervised learning algorithms—have been covered in this survey study.

Applications of machine learning in healthcare, business, commerce, tourism, the media, computer vision, NLP, automated trading, the auto industry, the aerospace industry, and manufacturing are also explored.

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