

Artificial Intelligence  
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## Introduction

As human beings, we have always been fascinated by technological change and fiction, and we are currently living in the greatest development in history. Artificial intelligence is emerging as the next big thing in technology and it is emerging in organizations around the world. Given this growth rate, we will continue to act as technological innovators for the foreseeable future. Artificial intelligence is shaping the future of virtually all industries and people (Haibe-Kains et al., 2020). It is an emerging technology and it will influence and influence the future of all industries and all people. Given this growth rate, we will continue to act as technological innovators for the foreseeable future. Artificial intelligence is an emerging technology that shapes the future of virtually all industries and people. Machine learning and computer vision are used to detect and classify various security events. We do not see shoebox-sized devices, but people see them in abundance, and they are not considered a threat to human life, health, safety, or the environment. There are various laws, regulatory policies, and procedures that monitor the use of artificial intelligence in today's world. IFM software automatically detects the presence or absence of a specific type of equipment in a warehouse or storage area and the warehouse manager can be immediately active. Human safety is always a priority here, but there is no need to monitor it because of the various laws, regulations, directives, or procedures that apply today.

## AI Approaches

### Machine Learning

Machine learning has had practical effects in a range of industries and sectors, including education, health, health care, finance, and even the military. AI systems will generally have the following characteristics, such as the ability to solve problems, make decisions, recognize the language, and process language (Haibe-Kains et al., 2020). Machine learning is an artificial intelligence (AI) branch that focuses on applications that learn from data and improve its accuracy over time by programming it in this way. It is trained on algorithms that find patterns and features in huge amounts of data to make decisions and predictions based on new data. The more data the algorithm processes, the more accurate the decision or prediction becomes about how it is processed. Generative Adversarial Networks, for example, are an example of an algorithm for machine learning that is trained to generate more images. More data must be generated using historical data to train the algorithm to learn from past data cases through statistical analysis and pattern matching.

Mathematics is useful for the development of models for machine learning, and finally, computer science is used to implement the algorithm. Below is a list of the most common and useful terms for machine learning. Some of these terms are confusing and many of you may find them confusing, so here are a few additional tips for users. Machine learning is a part of artificial intelligence that involves the implementation of an algorithm that can learn from data from previous instances and perform a task without explicit instructions. The process of learning from data involves adapting a model to evaluate the data more accurately and deliver precise results. Monitored machine learning facilitates training, because the model's results can be compared with the actual labeled results, but requires the use of more complex algorithms, such as neural

networks and more sophisticated algorithms. Machine learning can also be used in the field of computer vision and machine translation.

Properly labeled data is expensive to prepare, and there is a risk that models will be created that is so closely linked to training data and so biased that they will not properly handle deviations in new data. Unsupervised learning is the process of automating decision-making and identifying patterns and relationships in data that people would miss. It collects unmarked data, uses algorithms to extract the useful features needed to mark and classify the data in real-time without human intervention. Deep learning allows algorithms to be monitored, pattern analyses to be performed and data to be classified. Unlike machine learning algorithms currently used or developed, deep learning can absorb most data and beat people in cognitive tasks. Computer vision and speech recognition have made significant progress in deep learning approaches.

IBM's Watson is a well-known example of a system that uses deep learning, as well as several other systems, such as Google's Google Glass. Machine learning is a powerful tool for predicting the future and helping people make the necessary decisions. To gain deeper insights and understand why machine learning is trending, read DataFlair's latest machine learning tutorial. Machine learning is trained by learning from past experiences, for example, about past experiences, but also by analyzing historical data. Therefore, we can identify patterns to make predictions through training examples. Complex models can make accurate predictions, but it can be difficult to explain to a layman how the output is determined.

## **Deep Learning Ai**

One of the highest-rated Coursera courses, Machine Learning, taught by Andrew Ng, teaches you the most effective machine learning techniques. This course is excellent for those

who want to start a new career in artificial intelligence, as well as for those who want to learn about the business aspects of artificial intelligence. Therefore, the graduate program in machine learning and AI is ideal for professionals interested in expanding their existing quantitative skills to become more effective with their machine learning skills (Haibe-Kains et al., 2020). They will learn about deep learning, artificial intelligence, machine learning, and deep neural networks, as well as the use of machine learning in business. Topics include supervised learning, which includes unsupervised learning, the use of machine learning techniques such as neural networks, explainable and interpretable deep learning models, and best practices in machine learning that explain prejudice and theories of variance. Interesting work includes the fact that deep learning is currently the most advanced artificial intelligence technology, that neural networks can develop behavior with less training data, and that a deeply learned algorithm can perform a task much more efficiently than a conventional machine learning algorithm. Although deep learning is the ultimate goal of the AI industry today, it is not the end - everything, be it AI, but the beginning of it (Jin et al., 2020).

The development of deep learning and neural networks could make way for completely new architectures. When people use the term "deep learning," they refer to a deep artificial neural network, and that is a new record - the setting of algorithms (Jin et al., 2020). Deep enhancement learning uses in-depth enhancement training and deep neural networking as the basis for the development of new algorithms. Deep learning machine learning algorithms like DeepMind's AlphaGo and Google's TensorFlow outperform their human counterparts in a variety of tasks. Machine learning is typically divided into two types of learning: supervised learning, in which data is flagged, and unsupervised learning, in which computers group similar data. In artificial intelligence, we have machine learning, which uses a set of algorithms to search

through data to make a decision - to make a process and improve it. With machine learning comes deep learning that can make data meaningful by using multiple levels of abstraction. It includes the use of deep neural networks, machine-learning algorithms, and deep data processing.

### **Artificial Neural Network**

Deep learning as a subset of machine learning that uses algorithms inspired by CNN models can be divided into two categories: Regression Neural Networks and Backpropagation Neural Network (NLP). There is a wide range of applications for generalizing the regression of neural networks, such as computer vision, machine learning, and artificial intelligence (AI). NLP - trained back propaganda learning algorithm for neural networks, widely used in hydrological prediction models and one of the most popular models for predicting climate change. This is the basis for a large number of applications in the fields of computer vision, machine learning, and artificial intelligence (AI). Electrically adjustable resistance materials are used to mimic the function of synapses in software simulations performed on neural networks. Software simulations are carried out using neural networks, using a combination of computer vision, machine learning, and artificial intelligence (AI) techniques (Jin et al., 2020).

It is easy to get confused, but once neural network algorithms are developed to do useful things with 1,000 neurons and 10,000 synapses, high-performance neural network hardware becomes essential for practical operation. Machine learning is a subset of AI that focuses on how a machine can learn about itself and how it can "learn" through machine learning. Deep learning is another subset that focuses on deep learning, or how each layer of a neural network can be used to generate output. Artificial neural networks (ANN) are computer systems designed to simulate the way the human brain processes information (Lalmuanawma et al., 2020). Neural

network software is simulated and ultimately used to develop and apply software concepts adapted from biological neural networks. The goal of artificial neural networks (AI) and deep learning software is to replicate the system's data processing and output to unravel the neural system.

In some cases, the human brain also benefits from the ability to process information more efficiently and efficiently than biological neural networks. Deep learning is a subset of machine learning, and neural networks form the backbone of deep learning algorithms. Indeed, there is no way to distinguish a single neural network from an algorithm of "deep learning" that must have more than three. Perhaps the easiest way to think about this is to think about the difference between a deep learning algorithm and a human brain, or even a biological and an artificial brain. At a basic level, a neural network consists of a collection of layers, each with its own rules and parameters. In practice, neural networks provide a sorting and classification level that is based on the managed data to bundle cluster and group data based on similarities (Lalmuanawma et al., 2020).

It is possible to develop customer relationship tools to detect fraudulent behavior and accurately assess moods, customer service, and customer service, or accurately evaluate customer relationships and tools using artificial neural networks. These networks are built with neurons - like computer nodes, which are Bebe neurons that communicate with each other via channels similar to the synapse. This means that the output of one arithmetic node affects the processing of another node and vice versa the output of other arithmetic nodes. But perhaps one of the greatest advantages of deep neural networks is the ability to process and cluster large amounts of data, such as large data sets and large data sets. When a neural network is connected to multiple systems, it can be routed around missing nodes that cannot communicate. It recreates



the data and helps determine the nodes and routes that are not working, and to identify the parts of a network that is no longer working. Given the ability of deep neural networks to process and bundle large amounts of data, they can process huge amounts of unstructured data to find similarities before they reach humans (Lalmuanawma et al., 2020).

### **Convolutional Neural Network**

In machine learning, a coevolutionary neural network (CNN or ConvNet) is a deep artificial neural network most commonly used to analyze visual images. This network is one of the most popular deep learning algorithms for image processing. It can take input from a deep learning algorithm (an image) and effectively weight and distinguish the object from others. It has been around for several decades and has proven to be very powerful, especially when large labeled datasets (e.g. images) are used. For example, Convolutional Neural Networks (ConvNets or CNNs) are used to identify objects in a scene (e.g. cars, buildings, and other objects). The effectiveness of coil webs in image recognition is one of the main reasons why the effectiveness of deep learning is awakening the world. The convolutionally of neural networks is a method of using neural networks to classify images (i.e., name the images you see) and perform object detection in scenes (Li et al., 2020).

A convolutional neural network is a complex neural network that can recognize complex features in data. CNN's are used in vision-powered robots, self-driving vehicles, and even artificial intelligence (AI) applications. A deep revolutionary architecture called AlexNet is the most advanced deep learning system for hearing gunshots in the world. Convolutional Neural Networks (ConvNets, or CNNs) is a deep learning system that has proven itself in a variety of applications including speech recognition, image classification, and machine learning. (CNN) -- Image classifiers can be used in myriad ways, for example, to classify cats and dogs or to

determine whether a brain image contains a tumor. ConvNets can recognize faces, objects, and road signs, and to promote vision in robots and self-driving cars.

ConvNets have also been the subject of several recent research projects in the field of artificial intelligence (AI). They can also be used in natural languages processing projects, such as speech recognition and speech to word processing. To understand what makes convolutionary neural networks so useful in computer vision applications, you can read our previous article on the subject in our section Computer Vision and Neural Networks. The conventional neural network is used in ComputerVision applications where the input is an image and regular, fully networked neural networks do not function well. One of the main reasons it is useful and a fast-growing area is its high performance, low cost, and low latency, which is a major reason why we are embedding it in a wide range of applications today such as image recognition, voice recognition, and image processing. Because if every pixel in the image is input, if more layers are added, the number of parameters will increase exponentially. Each weight matrix has several different parameters that can be used in different ways, such as hidden layers, coevolutionary neural networks, or coils.

Convolutional Neural Networks (CNN) -- A neural network designed to process structured data fields such as images. Normal neural networks are not scalable for image classification, because processing large inputs is very expensive and not feasible. Revolutionary neural networks are widespread in computer visualization and have also found their way into many other areas, such as machine learning and artificial intelligence (AI). They are also becoming increasingly popular in the field of image recognition and image processing. Convolutional neural networks can pick out patterns in a large number of different data types, such as colors, shapes, and shapes of objects. They are pushing the development of deep neural

networks, which are becoming increasingly important in visual recognition and classification. The architecture of these networks is loosely inspired by the brain, where several groups of neurons communicate with each other to respond to different inputs. In a regular deep neural network, we can observe a single vector input passing through a series of hidden layers.

These hidden layers consist of two completely connected layers, one for each class, and then the output layer is joined, resulting in a certain class value. The figure below is an example of a CNN model that deeply engages in education and testing, processes an input image, and classifies an object based on these values. To train, we run the entered images through all previously connected layers and use the Softmax function to classify the object with probability values between 0 and 1. Then we connect the outputs of each level, which leads to the determination of the class values. It classifies objects based on these values and converts them into a class of objects. The folding preserves the relationship between the pixels so that the image properties can be learned from the small squares of the input data without the need for a large number of turns. Proposed by Yan LeCun in 1998, evolutionary neural networks can identify the number present in a given input image. Before we start with a convolutional neural network, it is important to understand the functioning of a neural network. Neural networks mimic how the human brain solves complex problems and finds patterns in any data set.

## Artificial Intelligence Evaluation Techniques

Common model evaluation metrics for classification and regression problems with Python. The above problems are solved by evaluating machine learning models, which are an integral part of any data science project. The model evaluation aims to estimate the performance of a particular model, such as a classification model or a regression model. A classifier is a machine learning algorithm that is nowadays often learned during the training phase of a data

science project, such as classification or regression. Classifiers are a series of hypotheses (discrete - evaluated functions) that are used to assign categorical class names to certain data points. They are the most common type of learning algorithm in this area and they are derived from modeling an unknown target function before formulating a definitive hypothesis. For example, in the classification of e-mails, a classifier could be an algorithm for classifying e-mail addresses or other types of data such as SMS. It is worth noting that Panaseer does not have AI or machine learning (Li et al., 2020).

Instead, the tool is based on a process called Entity Resolution, which will be discussed in more detail in a forthcoming post. The solutions available today fall within the subset of machine learning in artificial intelligence. Generally, the methods contained in machine-learning algorithms are used to create prediction models. An example of a machine-learned solution is an algorithm trained on a data set for detecting malware, and based on its observed data, it provides the logic that can be used to predict future malware detections. In literature, the methods mentioned here are divided into two categories: knowledge-driven models and data-driven models or data-driven model groups. Using an algorithm in a data driver model group to study copper extraction is similar to using a knowledge-based mineral extraction or hydrometallurgy method, although, like all knowledge-gathering methods, you should consult a mineral extraction and hydronautics expert for this task (Li et al., 2020).

ANN, RT, and SVM models require enough records and parameters to achieve good quality output generated by the model. The most successful machine learning structures so far have been artificial neural networks, complex mathematical functions that can generate complex mappings between input and output. The second approach of machine learning systems is to provide data and problem space to AI models and allow them to develop their behavior. AI to

study data and develop mathematical models that represent common features and cancer patterns. Here are four explainable AI techniques that will help companies develop more transparent models of machine learning while maintaining the level of learning performance. Data scientists can explain the behavior of a model by looking at the data on which it is trained. The results of a machine learning model could be explained by data from the training data itself or by how the neural network interprets this data set.

A time series is simply a set of data points that are indexed in the order of the data, such as the number of days in a week, the time of day, and the date of the last day. This helps researchers determine what happens to model predictions when the various features are adjusted. A common goal of quantitative research is to identify the most useful features of a time series using statistical or machine learning methods. The first method we will examine is the Partial Dependency Plot (PDP), which was invented decades ago. It shows the marginal effect that one or two features have on the predicted outcome of the data. To create a prediction algorithm, many researchers take the first group of subjects to test it on a new population. The training data set is composed of all subjects who were used in the creation of the algorithm that carries out the prediction. To test whether the prediction algorithms work, researchers must test them on this population and test their work.

Therefore, it is important to expand and maintain the database so that the reported performance can be checked and competing algorithms can be pulled from the same data pool. A real comparison is difficult when a prediction tool is trained on a separate data set than another prediction tool because each of them trains on its own data set. Each iteration improves the performance of the model because the program can compare the results of the training sessions to see which works best and can change its overall predictive abilities. This improves the

generalization of results and leads to more accurate predictions for different populations.

Optimizing the rating set is one of the most important steps in developing a predictive algorithm.

Since the success of an algorithm in developing supportive technologies - that is, improving service and patient outcomes - depends on the tools doctors use, we advocate that the end product of developing prediction algorithms should include a comprehensive assessment of its performance in a variety of data sets. The tools for assessing the intelligence of developed systems are designed so that developers do not have to solve tasks in advance or program hard (Li et al., 2020)

Artificial intelligence will play a very important role in the future for the global economy and society as a whole, not only in manufacturing and logistics. AI and other technologies will revolutionize our world in the coming years, creating tremendous opportunities for economic and social development. Modern artificial intelligence is a broad and complex field, so most people need at least some general understanding of the term. The terms are often interchangeable and there are several different definitions of the term, such as "artificial intelligence," "human-like intelligence" and "intelligent intelligence." Artificial intelligence, in short, means programming a computer to do something that previously required human intelligence. AI is a very broad category, dating back to the very first computers. This is the subset that focuses on creating technologies that perform tasks, make their own decisions based on previous experience, that is, imitate human decisions – making (Paschen et al., 2020).

The use of artificial intelligence raises ethical questions, as AI tools offer companies a range of new features, and AI systems, for better or worse, will reinforce what they have already learned. This is problematic because the data and machine learning algorithms that underlie many of the most advanced AI tools are themselves the result of their training (Paschen et al.,

2020). As a result, the potential for machine learning distortions is inherent and needs to be monitored, regardless of what data is used to train the AI program. By enabling computers to learn and think independently, AI systems have the potential to help develop more efficient, intelligent, and powerful products and services. If machine learning is to be used as part of the real world, production systems must incorporate ethical aspects into their AI training and strive to avoid prejudice. By combining the human intellect with increasingly powerful AI, this technology is about to transform almost all areas of modern life. Although it would be impossible to study every single use of AI in a single article, we have identified some of the key areas where AI is set to revolutionize, listed below.

## Conclusion

One thing is certain: Artificial Intelligence is changing the world, and we must be part of it. One of the most important things that motivate a person to work is wisdom and exaggerated prospects, namely the use of artificial intelligence in modern medicine. The patient needs care, the procedure is programmed and the needs of the patient are taken into account. From the very first computers to do simple calculations to machine learning, to self-driving cars, AI is making our lives easier and more efficient. The everyday thing we use every day is machine learning, a more complex subset of AI that programs computers to make decisions based on past experiences, much like humans. Medical research is probably one of the areas where artificial intelligence could have the greatest impact, and not only in the field of medical care. With important medical issues, the potential of artificial intelligence (AI) technologies can decouple healthcare - appropriate information that is separated by vast amounts of data that can be stored

for health-care decisions - and make decisions. Artificial intelligence and AI techniques are one of the most effective technologies used in modern health care. With the rapidly growing availability of medical research technologies, much research on the use of artificial intelligence in healthcare is already complete.



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