

A British science fiction writer Arthur C. Clarke famously wrote, “Any \_\_\_\_\_ technology is \_\_\_\_\_ from magic,” and that is very \_\_\_\_\_ with Artificial Intelligence today. In 2016, Google’s Go-playing algorithm called AlphaGo \_\_\_\_\_ Go champion, Lee Sedol. Go is the most complex board game \_\_\_\_\_ and creativity, so the game had \_\_\_\_\_ to be difficult for machines. But AI \_\_\_\_\_ humans much earlier than \_\_\_\_\_, making headlines around the world. Machine Intelligence has \_\_\_\_\_ results in practical fields as well, such as \_\_\_\_\_ pictures, \_\_\_\_\_ language, and driving cars. \_\_\_\_\_ will make AI \_\_\_\_\_ even more like magic. But it is \_\_\_\_\_ a computer software, not \_\_\_\_\_. The only thing computers can do is computing—\_\_\_\_\_. Instead of \_\_\_\_\_ by \_\_\_\_\_, we need to understand what exactly is AI—what it \_\_\_\_\_ do and what it \_\_\_\_\_ do—and properly \_\_\_\_\_ to the economy \_\_\_\_\_ daily lives.

A British science fiction writer Arthur C. Clarke famously wrote, “Any sufficiently advanced technology is indistinguishable from magic,” and that is very much the case with Artificial Intelligence today. In 2016, Google’s Go-playing algorithm called AlphaGo defeated the world Go champion, Lee Sedol. Go is the most complex board game that requires intuition and creativity, so the game had long been considered to be difficult for machines. But AI prevailed over humans much earlier than predicted, making headlines around the world. Machine Intelligence has achieved impressive results in practical fields as well, such as labeling pictures, recognizing spoken language, and driving cars. News like this will make AI seem even more like magic. But it is essentially a computer software, not an almighty god. The only thing computers can do is computing—mathematical calculations. Instead of being tricked by its illusion, we need to understand what exactly is AI—what it can do and what it can’t do—and properly evaluate its potential impact to the economy and to our daily lives.

sufficient	十分な、足りる	prevail	打ち勝つ、勝る
indistinguishable	見分けがつかない	headline	見出し、主な項目
require	必要とする、要求する	label	識別する、レッテルを貼る
intuition	直感	almighty	全能の

Most computer scientists \_\_\_\_\_ AI \_\_\_\_\_ a computer algorithm \_\_\_\_\_ by \_\_\_\_\_ machine learning. Let's start \_\_\_\_\_ algorithm. An algorithm is \_\_\_\_\_ can be used to reach decisions and \_\_\_\_\_ problems. An algorithm \_\_\_\_\_ particular solution or decision, but \_\_\_\_\_ process \_\_\_\_\_ when providing a solution. For example, if you want to \_\_\_\_\_ the \_\_\_\_\_ between two numbers, you can use a simple algorithm. The algorithm says: 'First step: \_\_\_\_\_ the two numbers together. Second step: \_\_\_\_\_ by two.' When you \_\_\_\_\_ the numbers 3 and 5, you will get 4. When you enter 64 and 82, you will get 73. \_\_\_\_\_, an algorithm is a set of \_\_\_\_\_ that \_\_\_\_\_ based on specific input. Computer algorithms \_\_\_\_\_ essentially the same way. But they \_\_\_\_\_ much more complicated problems by \_\_\_\_\_ computing power. For example, go-playing algorithm receives input such as the \_\_\_\_\_ and the last \_\_\_\_\_ by \_\_\_\_\_, and generates output, which is your next move that most likely \_\_\_\_\_. Similarly, text-to-image algorithms like \_\_\_\_\_ are given text \_\_\_\_\_ input and generate \_\_\_\_\_ picture as the output. Algorithms for medical \_\_\_\_\_ skin images \_\_\_\_\_ patient, and then \_\_\_\_\_ and other potential health risks.

Most computer scientists define AI as a computer algorithm powered by advanced machine learning. Let's start with an algorithm. An algorithm is a set of steps that can be used to reach decisions and resolve problems. An algorithm isn't a particular solution or decision, but the method or process followed when providing a solution. For example, if you want to calculate the average between two numbers, you can use a simple algorithm. The algorithm says: 'First step: add the two numbers together. Second step: divide the sum by two.' When you enter the numbers 3 and 5, you will get 4. When you enter 64 and 82, you will get 73. In other words, an algorithm is a set of procedure that produces output based on specific input. Computer algorithms run essentially the same way. But they work on much more complicated problems by harnessing its massive computing power. For example, go-playing algorithm receives input such as the current face of the board and the last move by the opponent, and generates output, which is your next move that most likely claims victory. Similarly, text-to-image algorithms like Stable Diffusion are given text descriptions as input and generate a corresponding picture as the output. Algorithms for medical diagnosis are fed with skin images of a patient, and then detect cancer and other potential health risks.

define	定義する	diagnosis	診断
harness	動力化する、利用する	feed	餌をやる、投入する
description	記述、描写		

19 years before AlphaGo defeated the human champion, IBM’s chess-playing computer “Deep Blue” \_\_\_\_\_ the human chess champion, Garry Kasparov. These two achievements look almost \_\_\_\_\_, but these two board game algorithms \_\_\_\_\_ quite differently. Deep Blue was programmed by expert chess players and computer scientists, \_\_\_\_\_ AlphaGo learned \_\_\_\_\_. Learning is the key thing \_\_\_\_\_. AlphaGo self-learned how to play the game of \_\_\_\_\_ data. In other words, Deep Blue had human \_\_\_\_\_, but AlphaGo didn’t. Instead, AlphaGo was \_\_\_\_\_ by data. First, AlphaGo was \_\_\_\_\_ 100,000 games that strong \_\_\_\_\_ played as a training data set and \_\_\_\_\_ human players. Then, it \_\_\_\_\_ different \_\_\_\_\_ millions of times and learned from \_\_\_\_\_ data. This self-learning mechanism is \_\_\_\_\_ “machine learning.” The idea itself is nothing new, but \_\_\_\_\_ the massive increases in computing power in recent years, machine learning has become extremely powerful. Machine learning \_\_\_\_\_ a variety of methods \_\_\_\_\_, including Deep \_\_\_\_\_ network, which essentially \_\_\_\_\_ neurons work in the brain. The \_\_\_\_\_ of machine learning algorithms is \_\_\_\_\_ they are learning for themselves, they can \_\_\_\_\_ what we human \_\_\_\_\_ know how to do. They make \_\_\_\_\_ in many practical areas, \_\_\_\_\_ and self driving.

19 years before AlphaGo defeated the human champion, IBM’s chess-playing computer “Deep Blue” beat the human chess champion, Garry Kasparov. These two achievements look almost identical, but these two board game algorithms were built quite differently. Deep Blue was programmed by expert chess players and computer scientists, while AlphaGo learned by itself. Learning is the key thing here. AlphaGo self-learned how to play the game of go from data. In other words, Deep Blue had human teachers, but AlphaGo didn’t. Instead, AlphaGo was taught by data. First, AlphaGo was fed with 100,000 games that strong amateurs played as a training data set and mimicked human players. Then, it played against different versions of itself millions of times and learned from its own data. This self-learning mechanism is called “machine learning.” The idea itself is nothing new, but due to the massive increases in computing power in recent years, machine learning has become extremely powerful. Machine learning involves a variety of methods as its subset, including Deep neural network, which essentially imitates the way neurons work in the brain. The whole beauty of machine learning algorithms is that because they are learning for themselves, they can go beyond what we human programmers know how to do. They make breakthroughs in many practical areas, such as image recognition and self driving.

identical	まったく同じの	subset	下部集合
mimic	まねる	imitate	見習う、模範とする

## Artificial Intelligence #4

One of the technologies that have been \_\_\_\_\_ by \_\_\_\_\_ machine learning is image labeling, which is essential \_\_\_\_\_ self-driving cars. In order to drive safely, the machine has to be able to tell the difference between a plastic bag, which can be \_\_\_\_\_, and a cat on the road, \_\_\_\_\_ definitely \_\_\_\_\_. Previously, if you \_\_\_\_\_ algorithm to identify a strawberry, for example, the only choice was to teach machines what exactly is a strawberry. We had to \_\_\_\_\_ the fruit; red, almost \_\_\_\_\_-shaped with \_\_\_\_\_ and regular dots, \_\_\_\_\_. But what \_\_\_\_\_ it's not \_\_\_\_\_? What \_\_\_\_\_ it's \_\_\_\_\_ and not \_\_\_\_\_ of a cone? You have to \_\_\_\_\_ every possible case, including very unusual ones, and \_\_\_\_\_ the algorithm. Machine learning \_\_\_\_\_ this ridiculously \_\_\_\_\_ work. Instead of teaching machines by human hands, we \_\_\_\_\_ of pictures as training data. Some of them have to include strawberries. Then the machine figures out \_\_\_\_\_ of the fruit and self-learns how to identify the object. Machine learning seems to \_\_\_\_\_ possibilities, but we have to \_\_\_\_\_ its limitations. First, machines don't understand. They successfully identify the object, not by understanding what it means, but by \_\_\_\_\_ and \_\_\_\_\_ recognition. They only know that the new image is similar to \_\_\_\_\_ images \_\_\_\_\_ labeled "strawberry." The same is true for machine translation algorithms. When you type in "strawberry" on Google translate, \_\_\_\_\_ you get the result. But it doesn't know what exactly is strawberry. It only learned from data to give back "ichigo" \_\_\_\_\_ the translation of "strawberry." Second limitation is that they depend on \_\_\_\_\_ data. For human \_\_\_\_\_ to learn how to identify strawberries, they only have to see the object several times. But computers \_\_\_\_\_ millions of images \_\_\_\_\_ training data. The \_\_\_\_\_ in machine learning capability in \_\_\_\_\_ is \_\_\_\_\_ the easier and cheaper access to big data.

One of the technologies that have been accelerated by the latest machine learning is image labeling, which is essential for self-driving cars. In order to drive safely, the machine has to be able to tell the difference between a plastic bag, which can be run over, and a cat on the road, which should definitely be avoided. Previously, if you wanted an algorithm to identify a strawberry, for example, the only choice was to teach machines what exactly is a strawberry. We had to mathematically define the fruit; red, almost cone-shaped with shiny texture and regular dots, and so on. But what if it's not ripe? What if it's bitten and not the shape of a cone? You have to predict every possible case, including very unusual ones, and put them into the algorithm. Machine learning gets humans free from this ridiculously troublesome work. Instead of teaching machines by human hands, we feed vast amount of pictures as training data. Some of them have to include strawberries. Then the machine figures out distinctive characteristics of the fruit and self-learns how to identify the object. Machine learning seems to possess infinite possibilities, but we have to be aware of its limitations. First, machines don't understand. They successfully identify the object, not by understanding what it means, but by relying on statistics and pattern recognition. They only know that the new image is similar to previously seen images that are labeled "strawberry." The same is true for machine translation algorithms. When you type in "strawberry" on Google translate, soon you get the result. But it doesn't know what exactly is strawberry. It only learned from data to give back "ichigo" when asked the translation of "strawberry." Second limitation is that they depend on enormous amount of data. For human infants to learn how to identify strawberries, they only have to see the object several times. But computers require millions of images as training data. The drastic increase in machine learning capability in the last decade is due to the easier and cheaper access to big data.

Vocabulary List: No. 19~24



There is a type of algorithm \_\_\_\_\_ “\_\_\_\_\_ learning,” which doesn’t \_\_\_\_\_ training data and learns from \_\_\_\_\_ alone. Just one year \_\_\_\_\_ AlphaGo \_\_\_\_\_ Lee Sedol, \_\_\_\_\_ a much more elegant algorithm \_\_\_\_\_ AlphaZero. Unlike its previous version, AlphaZero knows nothing \_\_\_\_\_ of the game and learns \_\_\_\_\_ from \_\_\_\_\_. Instead of learning from human data, it simply \_\_\_\_\_ completely randomly \_\_\_\_\_ every game \_\_\_\_\_. In 2018, AlphaZero \_\_\_\_\_ Stockfish8, the 2016 top algorithm in chess. Can you guess how long it took AlphaZero to learn chess from scratch, \_\_\_\_\_ the match \_\_\_\_\_ Stockfish? Only four hours. Not four years or four days. Just four hours of \_\_\_\_\_ learning \_\_\_\_\_ AlphaZero to be the strongest chess algorithm in the world. Reinforcement learning \_\_\_\_\_ any tasks which \_\_\_\_\_ and limiting conditions. In self-driving, for example, \_\_\_\_\_ is to move from point A to point B as quickly as possible. Limiting conditions include \_\_\_\_\_ other cars and not hitting the \_\_\_\_\_. Then you \_\_\_\_\_ for a number of times. At first, cars on the software drive completely randomly and perform quite \_\_\_\_\_. However, they gradually learn to drive safe and \_\_\_\_\_ human drivers. Machine intelligence still \_\_\_\_\_ of human intelligence. \_\_\_\_\_ machines can learn to avoid a cat on a road, they don’t understand what a cat is, and never find \_\_\_\_\_. A Human driver can go home and make pancakes for his daughter and read books, while a self-driving algorithm can do nothing else \_\_\_\_\_. However, \_\_\_\_\_ a taxi is to \_\_\_\_\_ to the \_\_\_\_\_ quickly, safely and \_\_\_\_\_. What \_\_\_\_\_ the \_\_\_\_\_ in the future look like when computers outperform humans in specific tasks like driving a car, \_\_\_\_\_ cancer and even drawing a picture?

There is a type of algorithm called “reinforcement learning,” which doesn’t require training data and learns from trial and error alone.

Just one year after AlphaGo beat Lee Sedol, the team developed a much more elegant algorithm named AlphaZero. Unlike its previous version, AlphaZero knows nothing beyond the rules of the game and learns literally from scratch. Instead of learning from human data, it simply plays against itself completely randomly and continually improves from every game it plays. In 2018, AlphaZero defeated Stockfish8, the 2016 top algorithm in chess. Can you guess how long it took AlphaZero to learn chess from scratch, prepare for the match against Stockfish? Only four hours. Not four years or four days. Just four hours of self-reinforcement learning brought AlphaZero to be the strongest chess algorithm in the world.

Reinforcement learning is fit to any tasks which can clearly state its goal and limiting conditions.

In self-driving, for example, the goal is to move from point A to point B as quickly as possible. Limiting conditions include not crushing with other cars and not hitting the wall. Then you let the simulation run for a number of times. At first, cars on the software drive completely randomly and perform quite miserably.

However, they gradually learn to drive safe and eventually outperform human drivers.

Machine intelligence still falls far short of human intelligence. Though machines can learn to avoid a cat on a road, they don’t understand what a cat is, and never find cats adorable. A Human driver can go home and make pancakes for his daughter and read books, while a self-driving algorithm can do nothing else but drive.

However, all we want for a taxi is to bring us to the destination as quickly, safely and cheaply as possible. What will the labor market in the future look like when computers outperform humans in specific tasks like driving a car, diagnosing cancer and even drawing a picture?