

Artificial Intelligence in the Game of Go

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【编者按】编程人员认为,围棋比象棋更能精确地反映人类的思维方式。这是因为国际象棋虽然错综复杂,但仍可简化成强力运算。而围棋却不同,无论是人还是电脑,必须经过多年探索,运用积累的知识和逻辑,综合自己的模式匹配能力,才有可能掌握精湛的棋艺。至今仍没有电脑能够超出业余围棋手的水平。鉴于此,大卫·福兰指出,“编写一套好的围棋程序比编写一套好的象棋程序更具挑战性,它能让电脑更接近人的思维方式”。电脑围棋与人工智能究竟有何渊源?近几年来,专攻人工智能的科学家们有望在电脑模拟人类思考这一人工智能的核心技术中有所突破。

Early in the film “A Beautiful Mind”, the mathematician John Nash is seen sitting in a Princeton courtyard, hunched over a playing board covered with small black and white pieces that look like pebbles. He was playing Go, an ancient Asian game. Frustration at losing that game inspired the real Nash to pursue the mathematics of game theory^①, research for which he eventually was awarded a Nobel Prize.

In recent years, computer experts, particularly those specializing in artificial intelligence, have felt the same fascination and frustration. Programming other board games has been a relative snap. Even chess has succumbed to the power of the processor. Five years ago, a chess-playing computer called Deep Blue not only beat but thoroughly humbled Garry Kasparov, the world champion at that time. That is because chess, while highly complex, can be reduced to a matter of brute force computation. Go is different. Deceptively^② easy to learn, either for a computer or a human, it is a game of such depth and complexity that it can take years for a person to become a strong player. To date, no computer has been able to achieve a skill level beyond that of the casual player.

The game is played on a board divided into a grid of 19 horizontal and 19 vertical lines. Black and white pieces called stones are placed one at a time on the grid's intersections^③. The object is to acquire and defend territory by surrounding it with stones. Programmers working on Go see it as more accurate than chess in reflecting the ways the human mind

① game theory: 博弈论。 ② deceptively: 迷惑地, 虚伪地。 ③ intersection: 横断, 交点, 交叉线。

works. The challenge of programming a computer to mimic that process goes to the core of artificial intelligence, which involves the study of learning and decision-making, strategic thinking, knowledge representation, pattern recognition and perhaps most intriguingly, intuition.

Danny Hillis, a computer designer and chairman of the technology company Applied Minds, said the depth of Go made it ripe for the kind of scientific progress that came from studying one example in great detail. "We want the equivalent of a fruit fly to study," Hillis said. "Chess was the fruit fly for studying logic. Go may be the fruit fly for studying intuition."

Along with intuition, pattern recognition is a large part of the game. While computers are good at crunching numbers, people are naturally good at matching patterns. Humans can recognize an acquaintance at a glance, even from the back.

Daniel Bump, a mathematics professor at Stanford, works on a program called GNU Go in his spare time. "You can very quickly look at a chess game and see if there's some major issue," he said. But to make a decision in Go, he said, players must learn to combine their pattern-matching^④ abilities with the logic and knowledge they have accrued in years of playing.

One measure of the challenge the game poses is the performance of Go computer programs. The past five years have yielded incremental^⑤ improvements but no breakthroughs, said David Fotland, a programmer and chip designer in San Jose, California, who created and sells The Many Faces of Go, one of the few commercial Go programs.

Part of the challenge has to do with processing speed. The typical chess program can evaluate about 300,000 positions in a second, and Deep Blue was able to evaluate some 200 million positions in a second. By midgame, most Go programs can evaluate only a couple of dozen positions each second, said Anders Kierulf, who wrote a program called SmartGo.

In the course of a chess game, a player has an average of 25 to 35 moves available. In Go, on the other hand, a player can choose from an average of 240 moves. A Go-playing computer would need about 30,000

④ pattern-matching; 模式匹配。 ⑤ incremental; 增加的, 逐渐增长的, 递增的。

⑥ prune; 删除, 减少。

