



Methodologies and Tools for Creating Competitive Poker Playing Agents

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Abstract

Many researchers have devoted their time to develop software agents intended for strategic games. These agents obtained outstanding results in popular games such as Chess in which current agents cannot be consistently beaten by the best human players. However, for stochastic games with incomplete information there are still no optimal solutions, especially for games with large search spaces, where research is limited by current hardware. Poker is a game that is frequently used to measure progress in this domain, given its key features: simplicity; large number of decision points; hidden cards. Major scientific advances have already been achieved: agents are unbeatable in Head's up Limit Poker. However, in more popular Poker variants, agents are still far from perfect. In this thesis Poker is approached in-depth by addressing all necessary aspects to create Poker software agents, both in scientific and engineering terms. First, new tools for creating and testing agents are shown, namely a tool for automatic online playing. Next, advances on abstraction techniques are shown, namely a new no-domain specific method. Finally, techniques to enhance game play and decision making are analysed and compared. This includes agent architectures based on expert knowledge and optimizations in the usage of the current state-of-the-art algorithm for game playing (Counterfactual Regret Minimization). All developed methodologies were validated on simulated games or real games. Simulations show great efficiency improvements on current techniques. In real games the developed agents achieved a good result on the AAAI Annual Computer Poker Competition (2nd place in the Kuhn track) and, for the first time reported, they were also profitable in real money multiplayer online matches, against human players.

Keywords: Poker; Game Theory; Opponent Modelling; Simulation; Bot; General Game Playing; Game Description Languages.

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Resumo

Muitos investigadores dedicam tempo a desenvolver agentes destinados a jogos estratégicos. Estes obtiveram excelentes resultados em jogos populares como Xadrez, tendo ultrapassando o desempenho de jogadores humanos. No entanto, nos jogos estocásticos de informação incompleta não existem soluções ideais, especialmente para jogos com grande espaço de pesquisa, devido a limitações de *hardware*. O póquer é atualmente o jogo mais popular para medir os avanços nesta área pois tem regras simples, elevado número de pontos de decisão e cartas escondidas. Avanços científicos relevantes foram alcançados onde, inclusivamente, foram criados agentes imbatíveis na variante *Head's up Limit*. No entanto, em variantes mais populares, os agentes ainda não são perfeitos. Nesta tese são abordados todos os aspetos essenciais para a criação de agentes póquer, tanto em termos científicos como da engenharia da solução. Primeiro, foram criadas novas ferramentas para criar e testar agentes, com destaque para um programa que permite aos agentes jogarem *online*. De seguida, são abordadas técnicas de abstração, incluindo um novo método independente do domínio do jogo. Por fim, demonstram-se técnicas para melhorar a tomada de decisão, baseadas em arquiteturas de agentes com base no conhecimento de especialistas e otimizações no uso do algoritmo *Counterfactual Regret Minimization*, abordagem com melhores resultados teóricos nesta área atualmente. As metodologias desenvolvidas foram validadas via simulação e jogos reais. Nas simulações foi possível observar melhoramento da velocidade dos algoritmos. Nos testes em ambiente real, os agentes obtiveram bons resultados na competição do AAI (2º lugar em *Kuhn*) e nos jogos *online*, demonstrando-se que um agente pode ser rentável a jogar contra humanos.

Palavras-chave: Póquer; Teoria de Jogos; Modelação de Oponentes; Simulação; Agentes Automáticos; Jogos Genéricos; Linguagens de Descrição de Jogos.

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Luís Filipe Guimarães Teófilo

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"Where there's a will there's a way"

Samuel Smiles

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Abbreviations and Acronyms

AAAI – Association for the Advancement of Artificial Intelligence.

ACPC – Annual Computer Poker Competition (organized in the AAAI conference).

AI – Artificial Intelligence.

CIG – Complete Information Games

CFR – Counterfactual Regret Minimization.

CPRG – Computer Poker Research Group (University Alberta, Canada).

DEI – Informatics Engineering Department (FEUP).

EGT – Evolutionary game theory (EGT)

FAI – Fake Artificial Intelligence

FCT – Foundation for Science and Technology

FEUP – Faculty of Engineering, University of Porto

GDL – Game Description Language

IIG – Incomplete Information Games

IRC – Internet Relay Chat

LIACC – Artificial Intelligence and Computer Science Laboratory

NE – Nash-Equilibrium

UP – University of Porto

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Chapter 1

Introduction

This chapter provides an overview of this thesis and all the developed work that led to its writing. First, the context and motivation of the work are presented in order to justify it. To further emphasize the work's motivation, the main scientific challenges of this kind of work are also listed. Next, the main objectives and contributions of this thesis are described. The chapter is finalized by outlining the structure of this thesis.

1.1 Context

Artificial intelligence (AI) research is a field of study aimed at developing pieces of software and/or hardware that can replace or assist human beings in performing tasks that require intelligence i.e. tasks that are not methodical and that require expert knowledge about on how to deal with unforeseen events. This is contrary to common machines (such as appliances¹) or software (such as Notepad) whose aim is the systematic fulfilment of tasks that are composed by sets of instructions. Software applications or machines with intelligence also have the capacity to make decisions and control other systematic systems.

In the beginning, the main goal of AI research was to create a Strong AI: an intelligence that imitates brain functions or human behaviour, with the goal of creating intellects that match or exceed humans' one – towards a technological singularity² [1]. As years and research went by it was verified that such project was impracticable at

¹ Some modern appliances can perform intelligent tasks such as heat controlling.

² A time when machines are so smart that they are able to create smarter versions of themselves.