

# General Video Game playing AI: A Survey

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# Overview

## Introduction

- Research Scope

- Motivation

## Problem

- Non-human game-playino

- Generality

## Approachs

- Neural Network

- Reinforcement Learning

- Monte-Carlo Tree Search

## Competitions

## Application

- AlphaGo

- StarCraft

- Atari Games

# Video Game Playing AI

Video games have long been popular benchmarks for Artificial Intelligence(AI)[1]. Many researches have been done in building a optimal **Video Game Playing AI(VPAI)** in certain games.

Such attempts comprises Chess, Go, Car Racing games, Ms.PacMan, Real-Time Strategy (RTS) games and Super Mario Bros, etc.

# Research Scope

In this research, we focus on **General Video Game Playing AI(GVPAI)**.

GVPAI aims at building an no-human-intervening game-playing agent that is able to playing multiple games rather than specialized VPAI designed for certain one game[1].

## Motivation[2]

The objective of General Video Game Playing (GVGP) is to by-pass the addition of game specific knowledge, especially if the algorithm is tested in games that have not been played before.

Obviously, algorithms that approach GVGP problems may still count on some kind of domain knowledge, and the questions raised above could still be asked. Indeed, many different algorithms can be employed for GVGP, and chances are that heuristics will still make a big difference.

However, by reducing the game-dependent knowledge, approaches are forced to be more general, and research conducted in this field is closer to the open domain of General Artificial Intelligence.

# Problem Definition

- ▶ Non-player game-playing
- ▶ Generality

# Non-human Game-playing[3]

## Game-playing

How AI agent playing games is the core problem. This may involve modeling the way human playing games. Some characteristics such as short-term memory, reaction time and perceptual capabilities should be take into concern to make AI more like a real human player rather than dull Bots.

## Non-human behavior

GVPAI should act like human rather than autonomous machine program, Many games have non-player characters (NPCs), and AI can help in making NPCs believable, human-like, social and expressive.

# Generality[4]

Develop AI methods that work with not just one game, but with multiple game.

Develop methods that can do not only one task but several different related tasks.

Develop methods that can model, respond or reproduce the large variability among humans in design style, playing style, preferences and abilities.

## Heading

1. Game generality
2. Task generality
3. Player generality



# Approachs

- ▶ Neural Network(NN)
- ▶ Reinforcement Learning(RL)
- ▶ Monte-Carlo Tree Search(MCTS)

# NN-based GVGP

M. Hausknecht et al. [9] employed evolutionary neural networks to extract higher-dimensional representation forms from the raw game screen.

# RL-based GVGP

It is a best-first search algorithm that incrementally builds a tree representation of the search space of a problem (e.g., a game) and estimates the values of states by performing simulations.

# MCTS-based GVGP[5]

The parameters of an AI agent can be seen as a vector of integers and doubles. The tuning of parameters is therefore a problem of searching optimal numerical vector(s) in a given parameter search space.

# Competitions

One of the main challenges of General Video Game Playing is to create a software framework that allows for games to be designed and represented and different game-playing agents tested via some form of long-running competition.

- ▶ Ms pac-man competition
- ▶ The 2k botprize
- ▶ The Physical Travelling Salesman Problem

# Ms pac-man competition[2]

This competition is focused on the tasks of programming computer agents to play as either Ms. Pac-Man or as the ghosts.

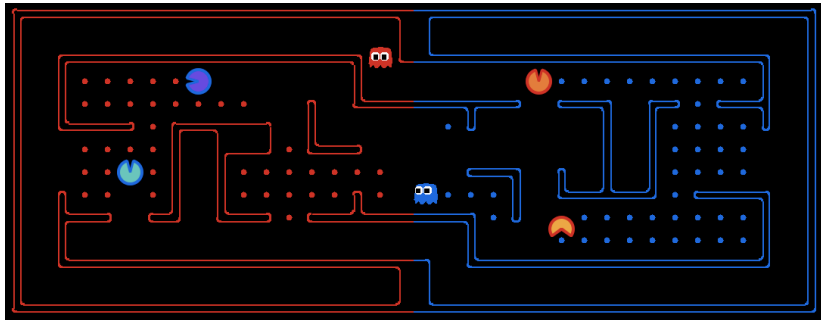


Figure: Pacman Capture the Flag

## The 2k botprize[3]

In the contest, bots try to convince a panel of expert judges that they are actually human players.

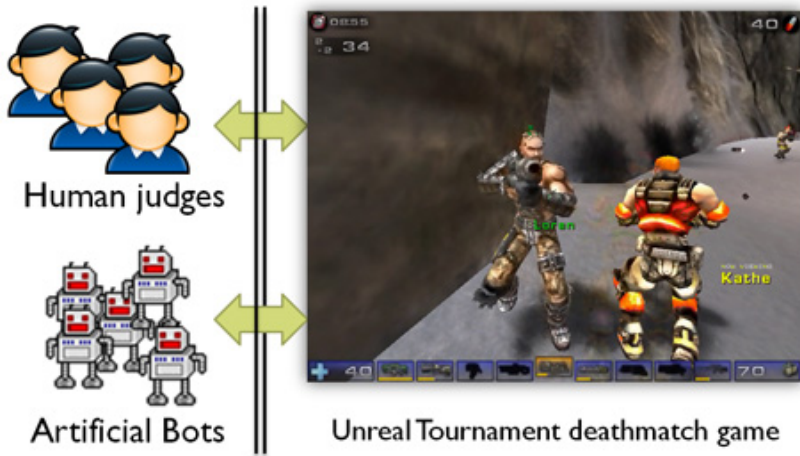


Figure: BotPrize judging protocol

# The physical travelling salesman problem competition[4]

It is a classic algorithmic problem in the field of computer science and operations research. It is focused on optimization.

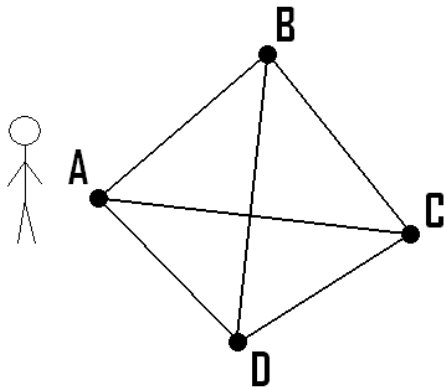


Figure: A salesman wants to visit all cities, A, B, C and D. What is the best way to do this



# Applications

- ▶ AlphaGo
- ▶ StarCraft
- ▶ Atari Games

# AlphaGo - Neural Network and Tree Search

Sliver et.al. [9] introduce a new approach to computer Go that uses 'value networks' to evaluate board positions and 'policy networks' to select moves.

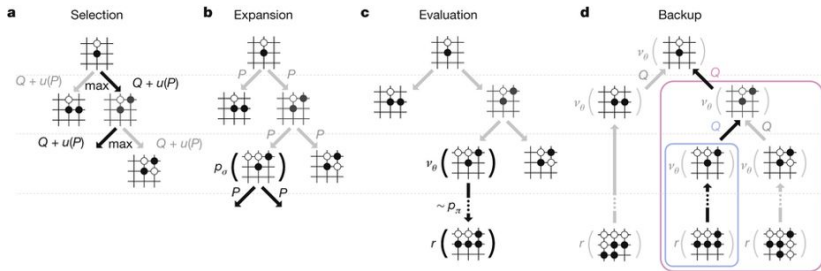


Figure: Monte Carlo tree search in AlphaGo

# Reinforcement Learning Applied to StarCraft

Wender and Watson did some research [10] on applying reinforcement learning (RL) to tiny scale combat in StarCraft, aiming to design an agent performing unsupervised learning in complex environment. The result showed the viability of RL algorithms in SC.

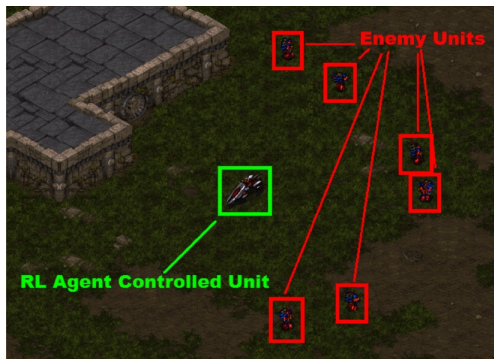


Figure: Initial unit positioning for the experimental evaluation

# Deep Reinforcement Learning Applied to Atari Games

Mnih et.al [11] applied convolutional neural network trained with a variant of Q-learning to seven Atari 2600 games from the Arcade Learning Environment. The AI reached the expert level of human-like player.



Figure: Screen shots from five Atari 2600 Games: (Left-to-right) Pong, Breakout, Space Invaders, Seaquest, Beam Rider

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