

Analysis of Vanilla Rolling Horizon Evolution Parameters in General Video Game Playing

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Introduction

- One of the most promising techniques in General Video Game AI competition ([GVGAI](#)) are the Rolling Horizon Evolutionary Algorithms ([RHEA](#)).
- Analysis of the vanilla version of RHEA on [20](#) GVGAI games
 - Special focus on the [population size](#) and the [individual length](#).
- Comparison with the sample Monte Carlo Tree Search ([MCTS](#))
 - Best sample agent in GVGAI.
 - Base of many winning competition entries.

RHEA in Game AI Literature

- Perez et al: comparison with tree search on the [Physical Travelling Salesman Problem](#)
- Justesen et al: [Hero Academy](#), groups of actions evolved for a single turn, for up to 6 different units, fixed population of 100 individuals (online evolution is able to beat MCTS).
- Wang et al: modified version in [Starcraft micro](#)¹, evolving plans to determine which script each unit should use at each time step.



Hero Academy: https://youtu.be/nox2dk0_aSA



Starcraft micro: <https://youtu.be/Xpjp0sm2reE>

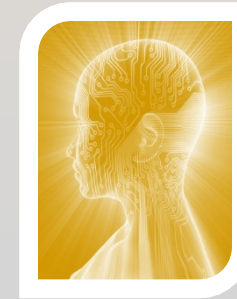
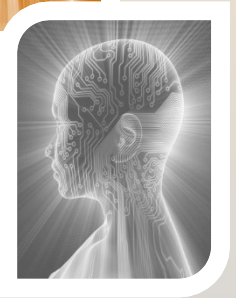
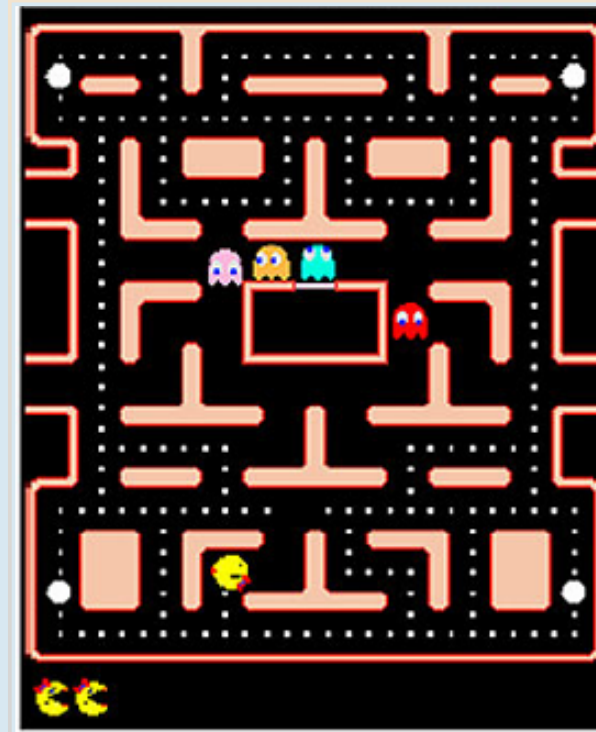
¹ JarCraft: <https://github.com/tbalint/JarCraft> :: SparCraft: <https://github.com/davechurchill/uAlbertabot>

Game AI

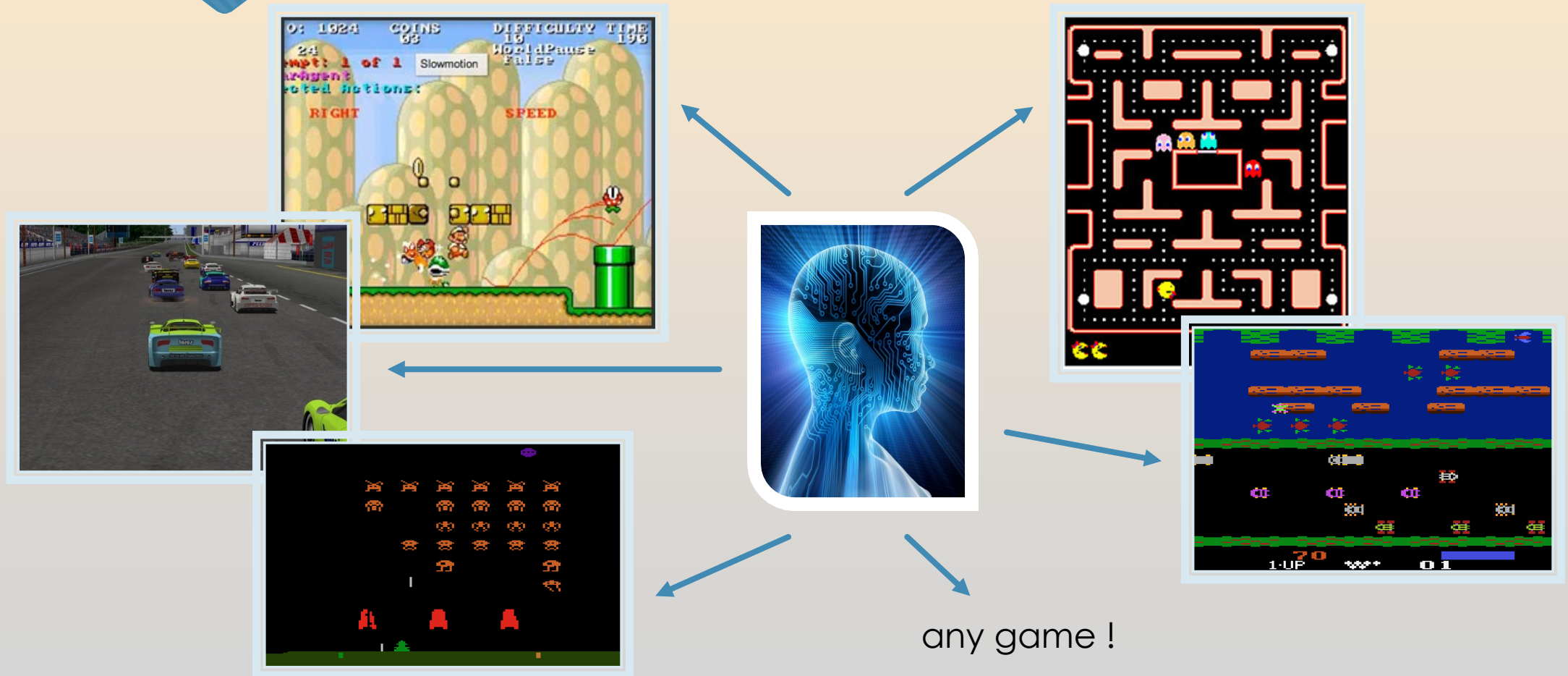


Super Mario AI

Ms. Pacman



General Video Game AI

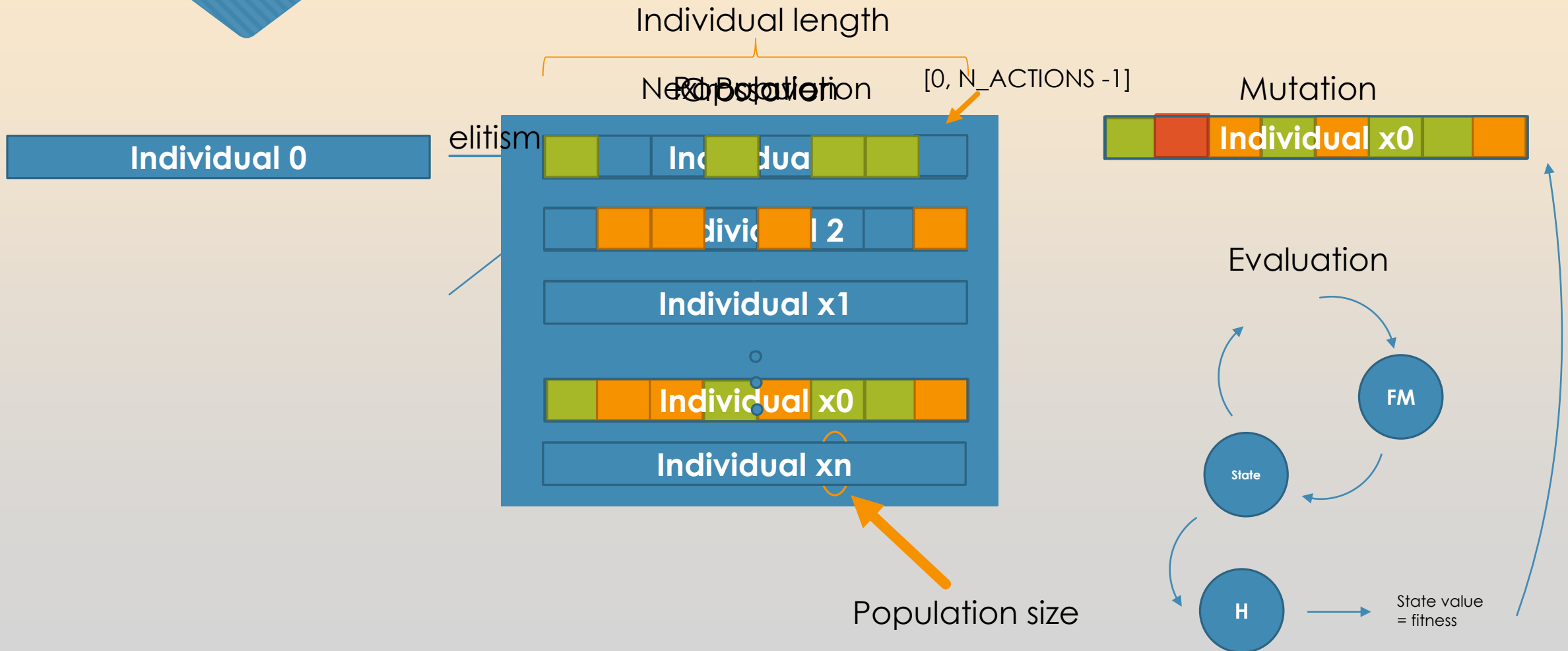


General Video Game AI Competition

- 2D grid-physics games
- Arcade, puzzles, shooters, adventure.
 - Ways to interact with the environment
 - Ways to win
 - Elements in a game
 - Scoring systems
 - Single and two player, cooperative and competitive.

... agents receive only a high-level view of the current game state and must make decisions in real-time (40ms)

Rolling Horizon Evolution



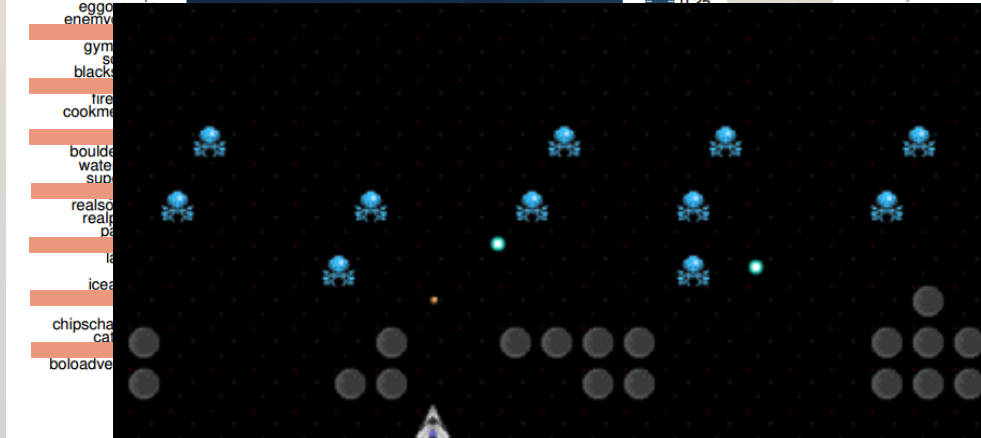
Approach

- Population sizes $P=\{1, 2, 5, 7, 10, 13, 20\}$, individual lengths $L=\{6, 8, 10, 12, 14, 16, 20\}$
- All other parameters fixed to default values
- Budget: 480 Forward Model calls
- Special case tested – Random Search: $P=24, L=20$
 - No evolution.
- Validation
 - Comparison with MCTS.
 - Budget extension.

20 Games from GVGAI corpus

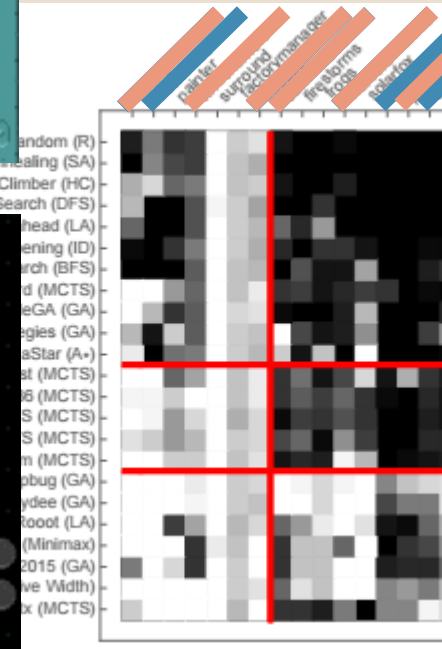


Survive Zombies



Aliens

ormly sampled from two
roller performance in 62
rollers + competition ent
nced set: 10 stochastic,



Sea Quest

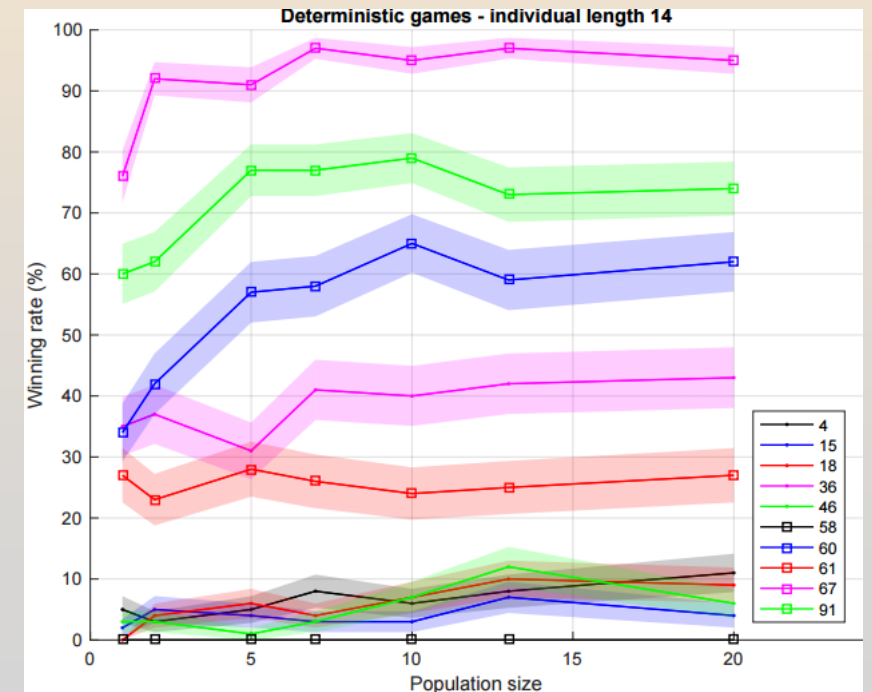
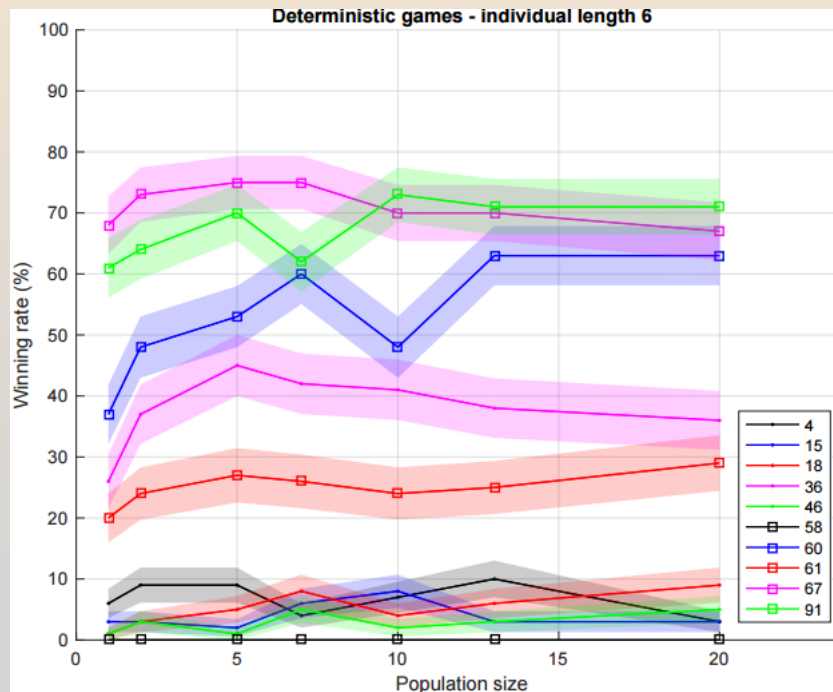
ACTS

Results Overview

- Trend noticed in most of the games: win rate increases, regardless of game type. Overall, performance increases with greater parameter values.
- Exceptions:
 - win rate starts at 100% (room for improvement, *Aliens* and *Intersection*)
 - win rate stays very close to 0% (outstanding difficulty, *Roguelike*).
- Best: **P = 20**, **L = 20**
 - 47.50 (2.33) win rate
- Worst: **P = 1**, **L = 20**
 - 33.15 (2.60)

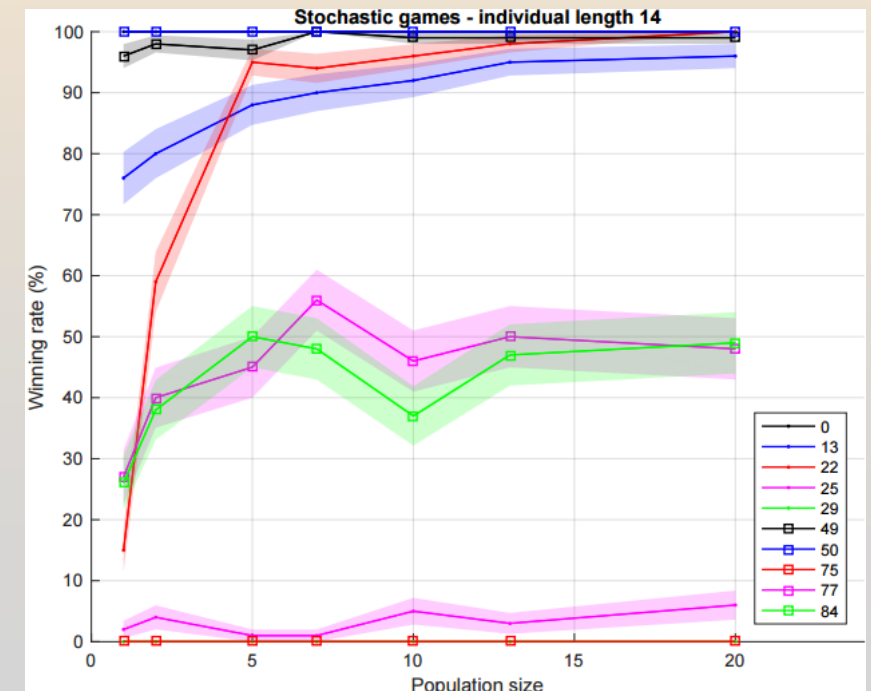
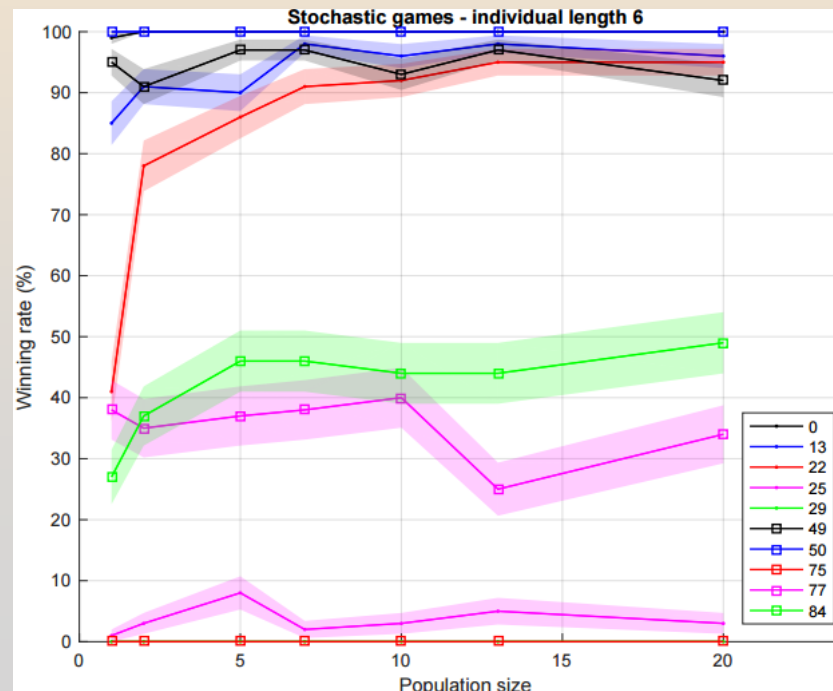
Results – Population Variation (Deterministic)

- Winning rate increases progressively in most games.
- High diversity in performance
- Interesting games (largest performance difference):
 - Game 67 (Plaque Attack)
 - Game 91 (Wait for Breakfast)
 - Game 60 (Missile Command)



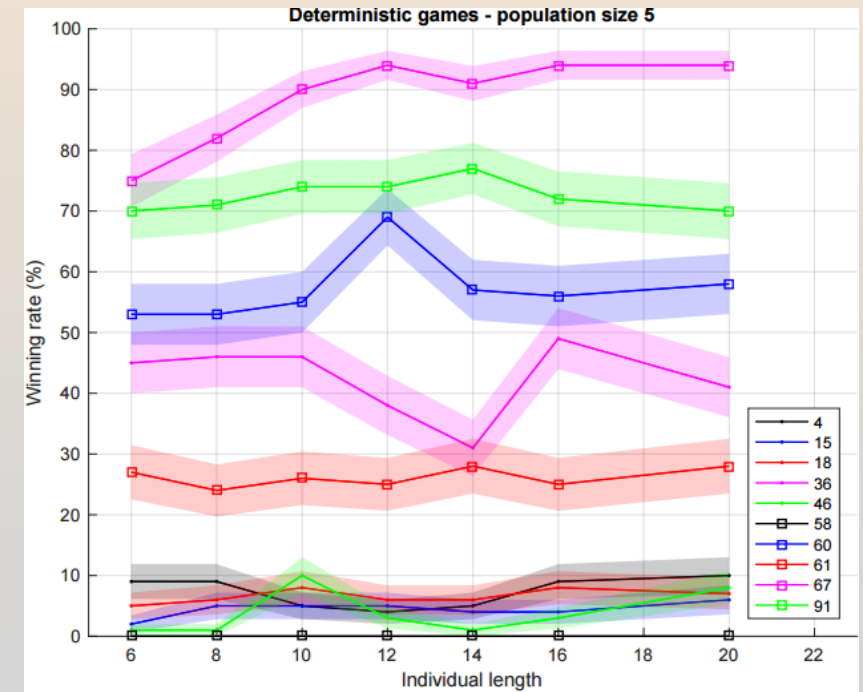
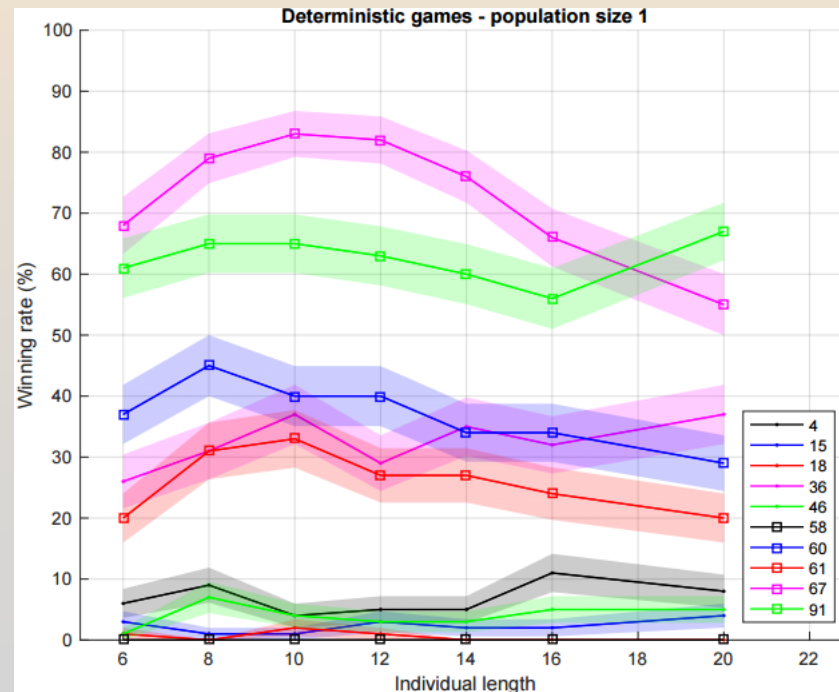
Results – Population Variation (Stochastic)

- If the length of the individual is small, increasing the population size is not beneficial in all cases, sometimes causing a drop in win rate
- Interesting games (largest performance difference):
 - Game 13 (Butterflies)
 - Game 22 (Chopper)
 - Game 25 (Crossfire)
 - Game 77 (Sea Quest)
 - Game 84 (Survive Zombies)



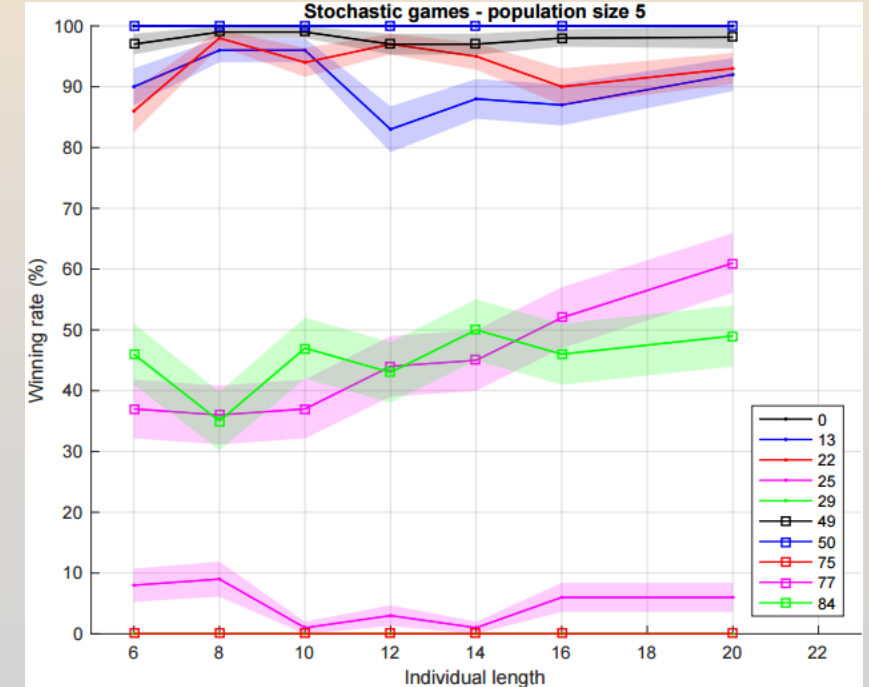
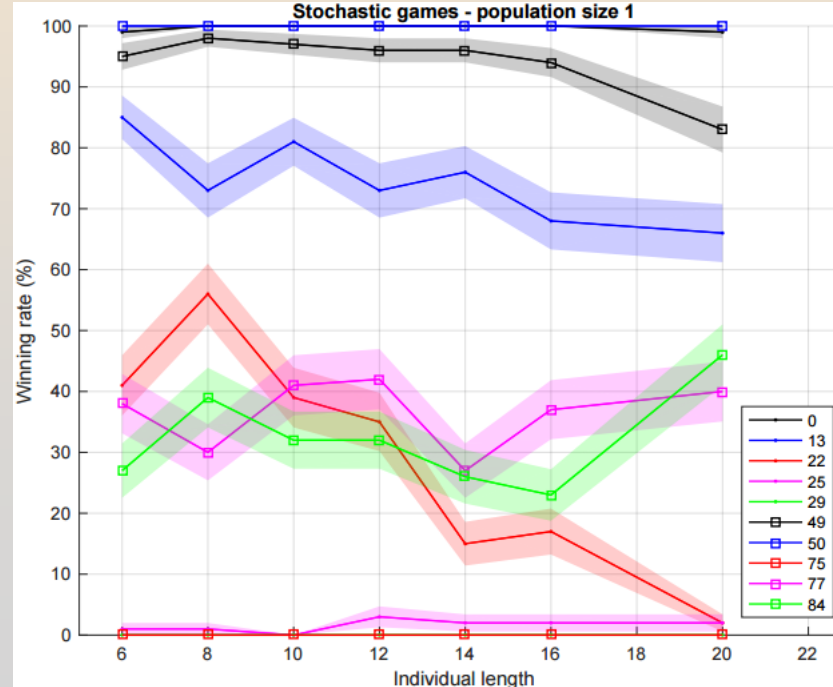
Results – Individual Variation (Deterministic)

- If population size is small, win rate sees a significant increase followed by a drop in large individual lengths; this issue is solved by increasing the population size.
- Interesting games (largest performance difference):
 - Game 67 (Plaque Attack)



Results – Individual Variation (Stochastic)

- Performance highly dependant on game.
- No significant change in win rate can be appreciated in larger population sizes.
- Interesting games (largest performance difference):
 - Game 13 (Butterflies)
 - Game 22 (Chopper)



Results – Random Search & Increased Budget

- Reminder:
 - No evolution.
 - $L = 20$, $P = 24$
- Performance no worse than any other RHEA configuration.
- Budget increase \Rightarrow Performance increase

Algorithm	Average Wins (T)	Points (T)	Average Wins (D)	Points (D)	Average Wins (S)	Points (S)
RHEA-1920	48.25 (2.36)	351	36.30 (2.88)	181	60.20 (1.84)	170
RHEA-1440	48.05 (2.23)	339	35.40 (2.82)	177	60.70 (1.65)	162
RHEA-960	47.85 (2.39)	323	34.60 (2.99)	162	61.10 (1.79)	161
RHEA/RS-480	46.60 (2.40)	271	32.90 (3.04)	131	60.30 (1.76)	140

Results – RHEA vs MCTS

- If $P > 5$, RHEA outperforms MCTS.
- Random Search (RS) outperforms MCTS in terms of win rate, but not in F1 points.
 - MCTS is more general.
- In deterministic games, MCTS performance similar to *worst RHEA* configuration ($P=1$, $L=20$).
- In stochastic games, MCTS and RS performances are similar.

Algorithm	Average Wins (T)	Average Wins (D)	Average Wins (S)
<i>Worst RHEA</i>	33.15 (2.60)	22.50 (2.99)	43.80 (2.22)
RHEA $P=1$	37.95 (2.47)	26.90 (2.93)	49.00 (2.01)
RHEA $P=2$	41.05 (2.62)	27.90 (3.05)	54.20 (2.20)
RHEA $P=5$	44.65 (2.40)	31.90 (3.18)	57.40 (1.61)
RHEA $P=7$	44.65 (2.36)	30.80 (3.09)	58.50 (1.64)
RHEA $P=10$	44.06 (2.26)	29.50 (2.90)	58.60 (1.63)
RHEA $P=13$	45.15 (2.47)	32.10 (3.06)	58.20 (1.88)
RHEA $P=20$	44.75 (2.31)	31.50 (2.87)	58.00 (1.74)
RS	46.60 (2.40)	32.90 (3.04)	60.30 (1.76)
MCTS	41.45 (1.89)	22.20 (2.45)	60.70 (1.34)

Summary

- Analysis of [population size](#) and [individual length](#) of vanilla Rolling Horizon Evolutionary Algorithm ([RHEA](#))
- Win rate measured on [20](#) games of the General Video Game AI corpus (selected based on difficulty for a diverse set, deterministic vs stochastic).
- Special case of [Random Search](#) studied, comparison with [MCTS](#) and increased budget effects.

Conclusions

- RHEA is no better than Random Search, worse in many cases.
- RHEA cannot explore space quickly enough in limited budget (the increased budget results confirm this; so better and faster evolutionary operators and improvements are needed).
- RHEA can outperform MCTS if population size is high.
- Performance increased in most games in **higher** population sizes and **higher** individual lengths, but there are cases where the opposite is true.
- Bigger impact noticed in population size variation than individual length.

Future Work

- Meta-heuristics: devise methods to identify the type of game being played and ...
 - ... employ different parameter settings.
 - ... modify dynamically parameter settings.
- Improvement of vanilla RHEA in this general setting.
 - Seeking bigger improvements of action sequences during the evolution phase, without the need of having too broad an exploration as in the case of RS.
 - Being able to better handle long individual lengths in order for them to not hinder the evolutionary process.
- Consider effects in stochastic games of ...
 - ... More elite members.
 - ... Resampling individuals to reduce noise.

Thank you