



Moving drones for Wireless Coverage in a three-dimensional grid Analyzed via Game Theory

Elena Camuffo, Luca Gorghetto and Leonardo Badia

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Outline

Introduction

- Drone's coordination
- Why Game Theory?
- State of the Art : Two-Dimensional Grid Games

Innovative scenario

- “Three-Dimensional Grid Game”

Game and Algorithm explanation

- Game-theoretical Analysis
- Nash Q - Algorithm

Results

- Convergence and Stationarity
- Average path length

Conclusions



Introduction

- Unmanned Aerial Vehicles (UAVs): aircraft without any human pilot onboard.



logistics

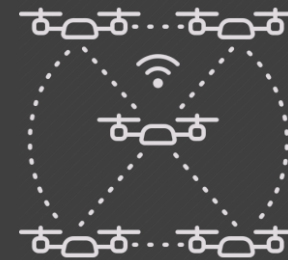


production



*wireless
communication*

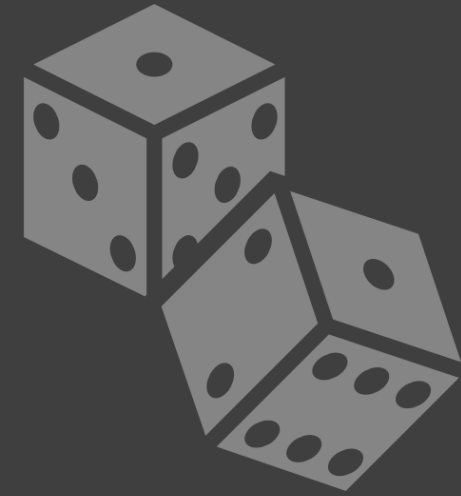
- Future uses of UAVs.



- Problem statement: drones need coordination.

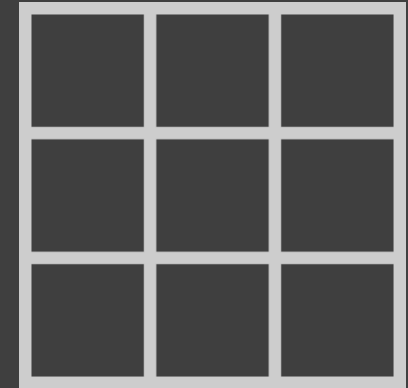
Why Game Theory?

- Motivation for the use of Game Theory
- Stochastic games
 - main idea: allowing to model drone's interactions
- Reinforcement learning (model-free)
 - used when learning problems arise
 - equilibrium learning vs adaptive learning algorithms



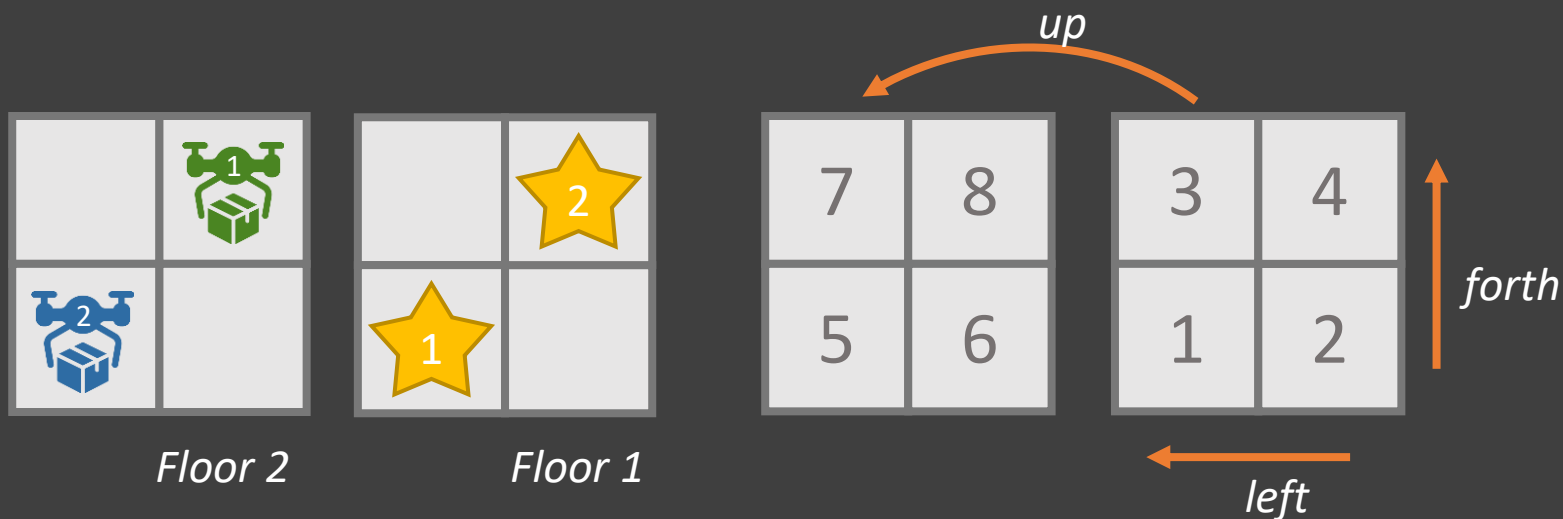
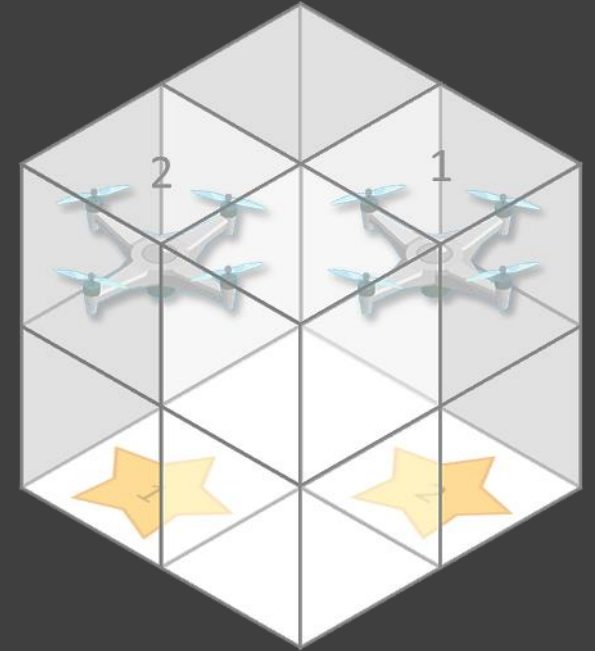
State of the Art

- Two-dimensional Grid games.
- Modeling problem for drones: only 2 dimensions.
- Main idea of the contribution of our work:
 - expansion towards the **third dimension**



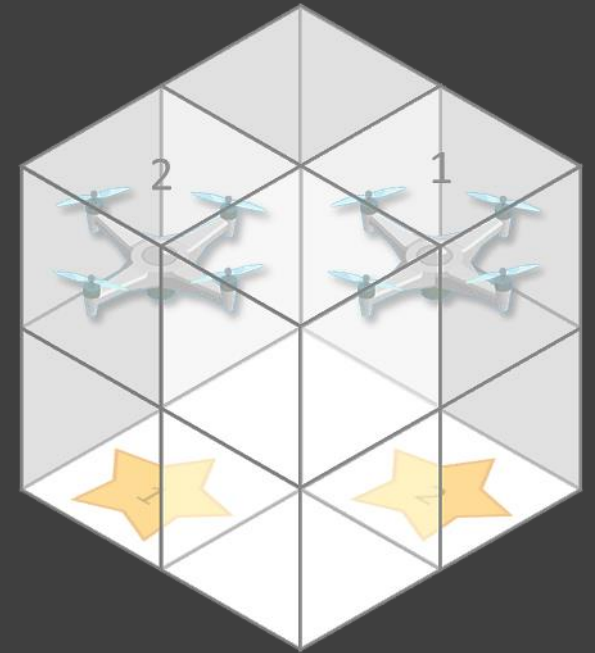
Innovative Scenario (1/2)

- Extended structure: the Three-Dimensional Grid Game
 - New variety of possible settings
- Possible moves (*Up, Down, Left, Right, Forth, Back*)



Innovative Scenario (2/2)

- Game structure/rules: drones
 1. Choose their actions **simultaneously**.
 2. When reaching the goal earn a **positive reward**.
 3. **Game ends** as soon as a drone reaches its goal.
 4. When moving into **the same cell** are **bounced back** to the previous.

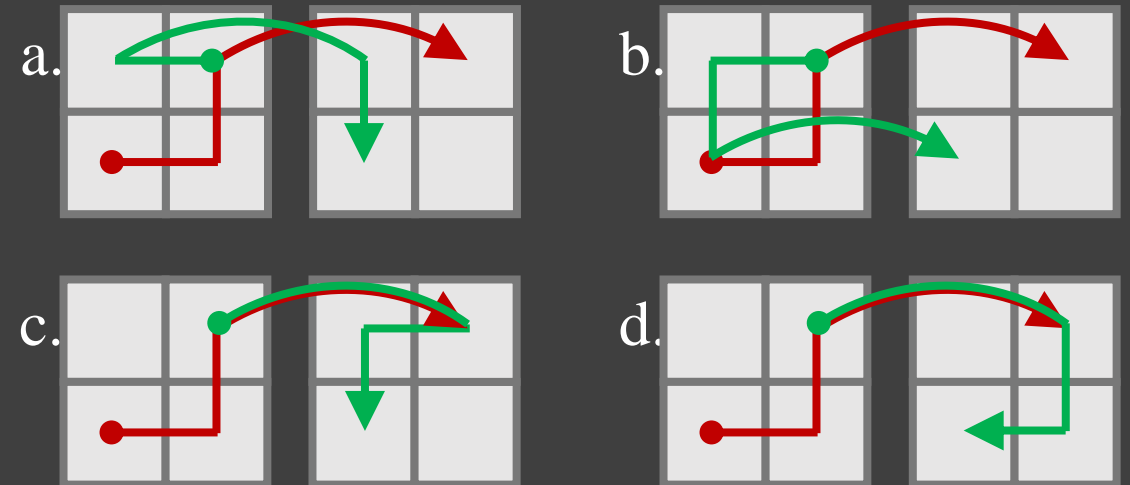
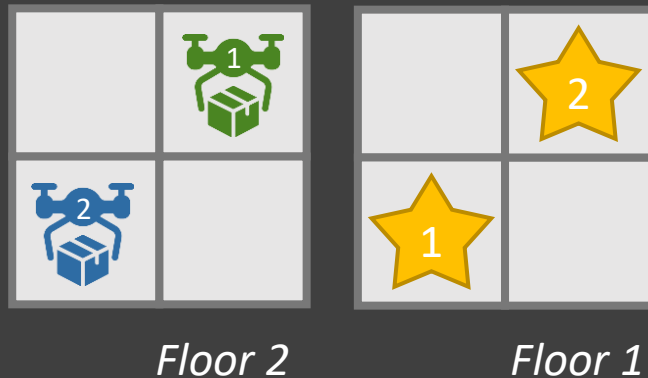


Game Analysis

- Assumptions:

- Rewards that each drone can receive:

- 100 points if it reaches the goal position.
- 1 points if it collides with the other drone.
- 0 points otherwise.



- State transitions are deterministic.

- Game Theoretical Analysis:
7 possible Nash Equilibria obtained.

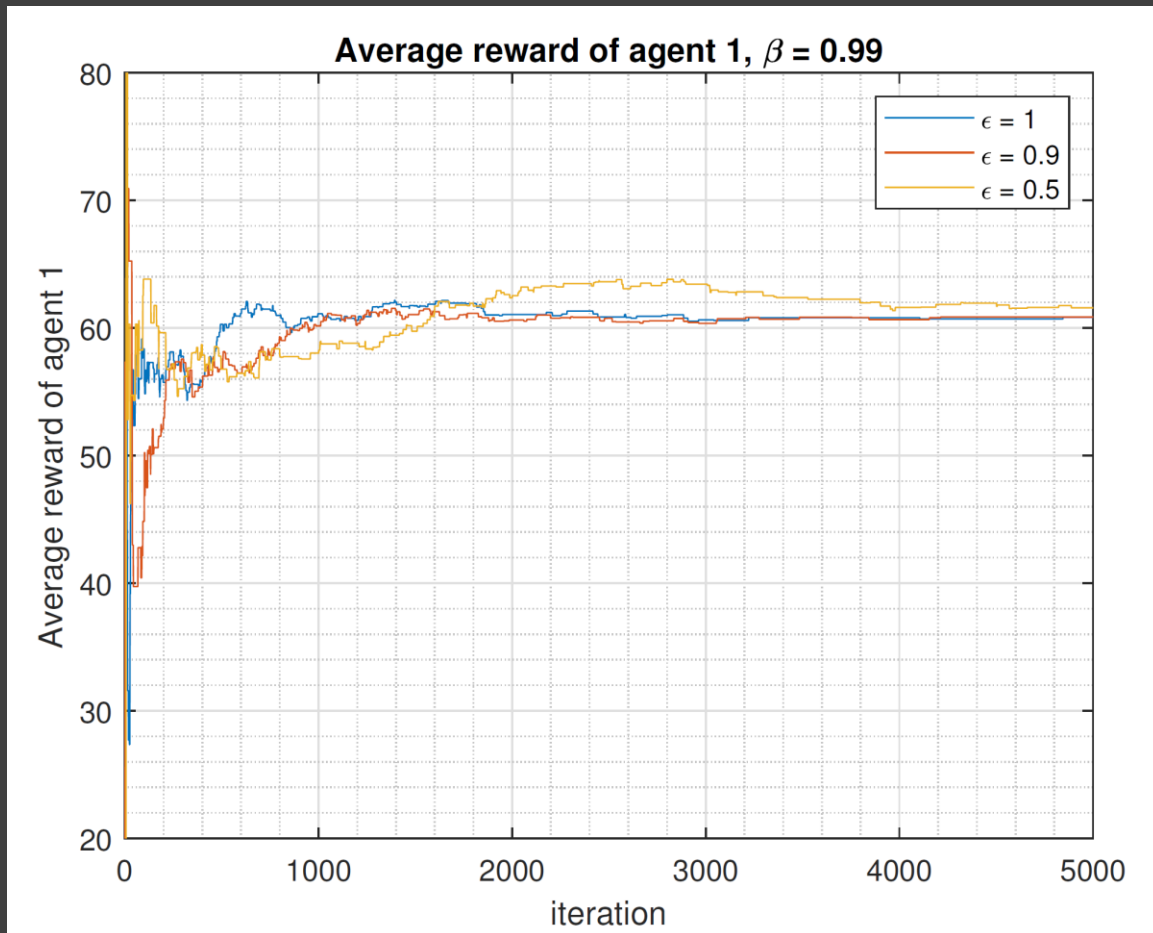


Nash-Q Algorithm

- Nash-Q Learning
 - Our version of Nash – Q algorithm by Hu & Wellman developed in Matlab.
 - Convergence in *self-play*.
 - Takes advantage of the Lemke Howson algorithm to find the NEs.
- Exploits ϵ -greedy exploration strategy
 - strategies to adopt can be explore, exploit, or explore and exploit.



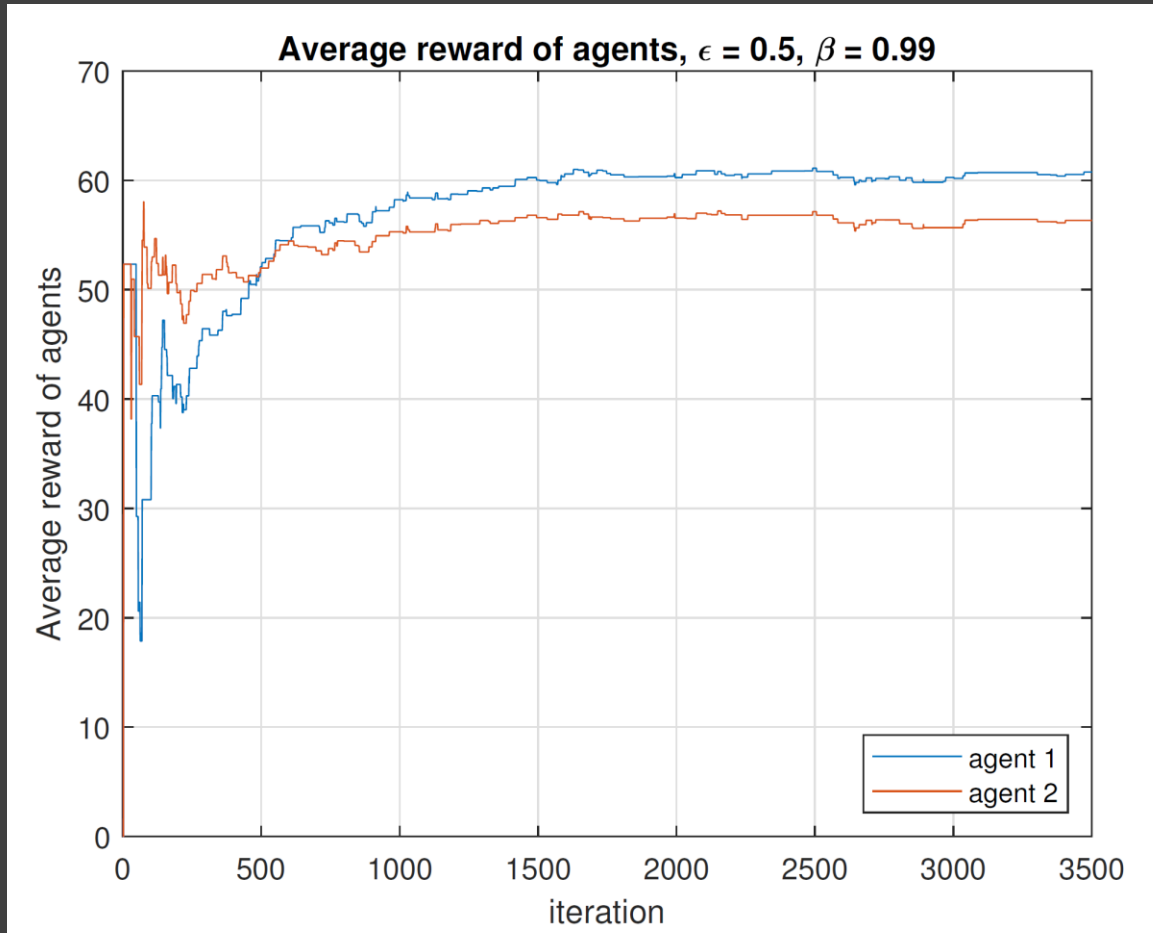
Results (1/3)



- *Convergence* of the algorithm for both players ...
- ... but with **different timing**.
- What does ϵ represent?
 - controls the probability of choosing the exploit strategy
 - How does the variation of ϵ change the final outcome?



Results (2/3)

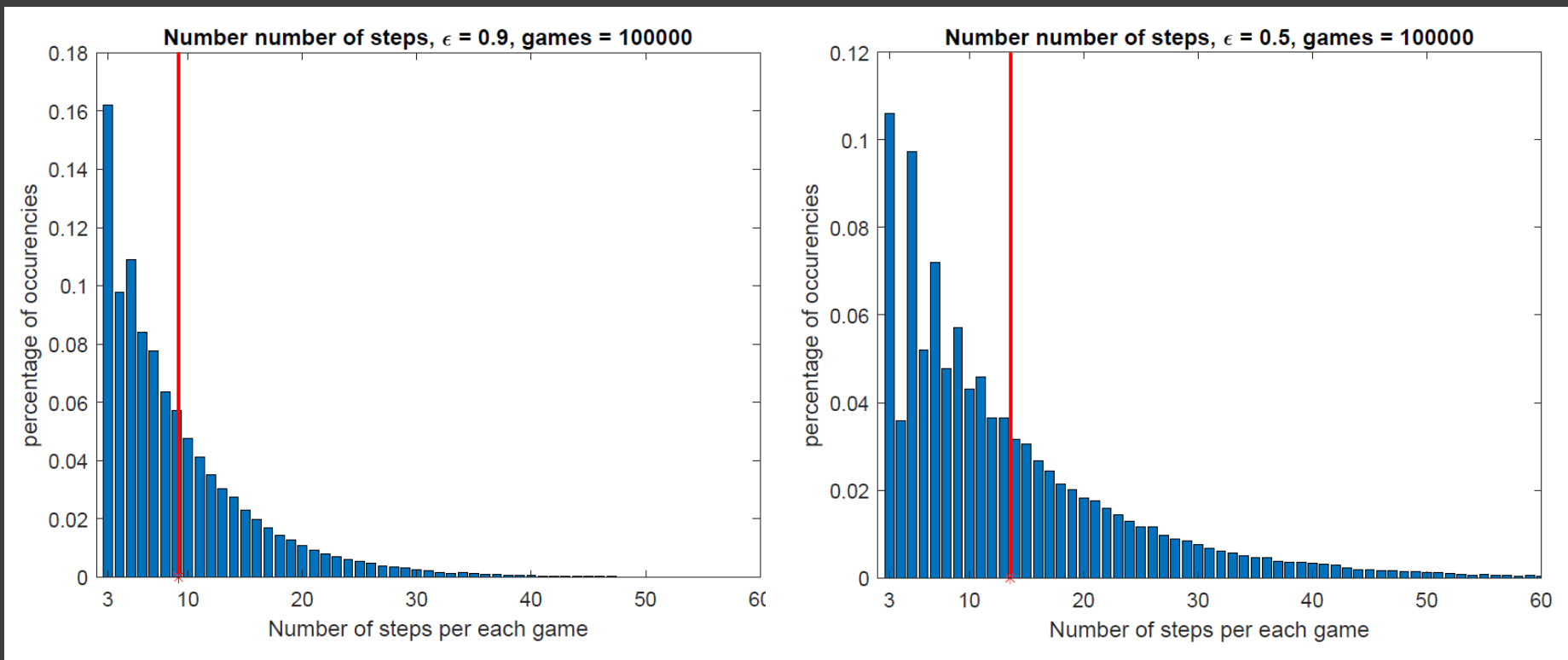


- What does β represent?
 - Discount factor.
- Average reward: similar behavior.
- Stationarity of the algorithm.



Results (3/3)

- New Metric introduced: Average path length.
- Number of steps per path distributed geometrically.



Conclusions

- ✓ Nash Equilibria verified.
 - ✓ Convergence of Nash-Q algorithm.
 - ✓ Average Path length evaluated.
- Possible extensions and Future works:
 - Include **more players/obstacles**.
 - Verify the solution with **other learning algorithms**.



Thank you!



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