# Multiagent Reinforcement Learning: Rollout and Policy Iteration (Bertsekas, 2020)

Mikalai Korbit

IMT School for Advanced Studies Lucca

January 20, 2021

#### Outline

Introduction

Brief Overveiw of MARL

Multiagent Rollout

Extensions

Conclusion

# Sidenote – Path Integral Control

**TODO** 

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#### Motivation

**TODO** TODO.

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## Outline

TODO TODO

## Reinforcement Learning Problem

- Learning how to map situations to actions
- Trial-and-error search
- Delayed feedback
- Trade-off between exploration and exploitation
- Sequential decision making
- Agent's actions affect the subsequent data it receives

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#### Conclusion

- RL methods can be applicable to a wide variety of problems
- Out-of-the-box models work but require fine-tuning and take longer to converge
- Simple methods like state discretization are worth exploring when training speed and solution complexity are of the essence

#### References



Reinforcement learning lectures by David Silver. UCL. http://www0.cs.ucl.ac.uk/staff/d.silver/web/Teaching.html

Playing Atari with Deep Reinforcement Learning. Mnih et al. https://arxiv.org/abs/1312.5602

# Thanks for your attention!