Atari Breakout with LTL_f/LDL_f Goals

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Introduction

Intro.

Q-Learning

Q-Learning brief description.

$$Q(S_t, A_t) \leftarrow Q(S_t, A_t) + \alpha \left[R_{t+1} + \gamma \max_{a} Q(S_{t+1}, a) - Q(S_t, A_t) \right]$$

SARSA

SARSA brief description.

$$Q(S_t, A_t) \leftarrow Q(S_t, A_t) + \alpha \Big[R_{t+1} + \gamma Q(S_{t+1}, A_{t+1}) - Q(S_t, A_t) \Big]$$

LTL_f/LDL_f Non-Markovian Rewards

 LTL_f/LDL_f non-Markovian rewards + integration in our project.

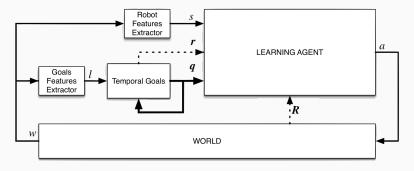


Figure 1: Pipeline describing how the agent is interacting with the world and how the robot features extractor and the goal features extractor are used in order to handle non-Markovian rewards.

PyGame Breakout

Results of the paper and our starting point.

PyGame Breakout (6×18)

Our results on 6×18 non-Atari Breakout + video.

Atari Breakout

Introduction to Gym + ALE
Differences with non-Atari Breakout (initial hypotheses)



Implementation

Robot Features Extraction

Algorithms.

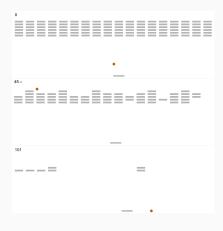
Goal Features Extraction

Algorithms.

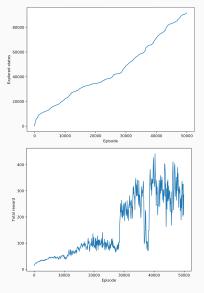
Temporal Goals

Algorithms.

All the experiments.



Description.



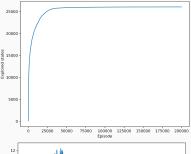
PyGame SARSA.

Atari frames.

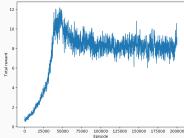


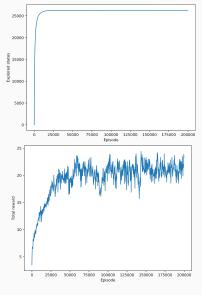






Atari Q-Learning.





Atari SARSA.

Conclusion

Conclusion.



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