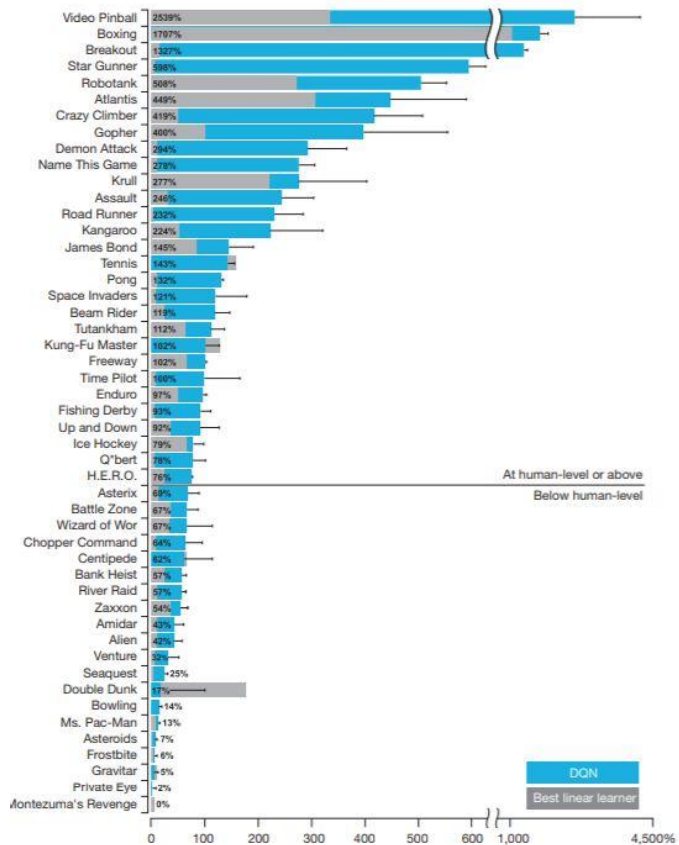


Sparse Reward and Temporal Abstraction: Hierarchical DQN

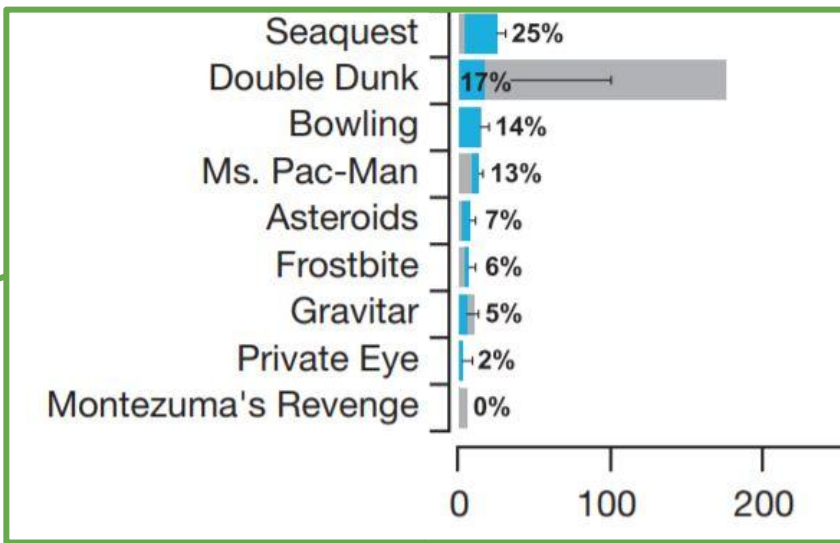
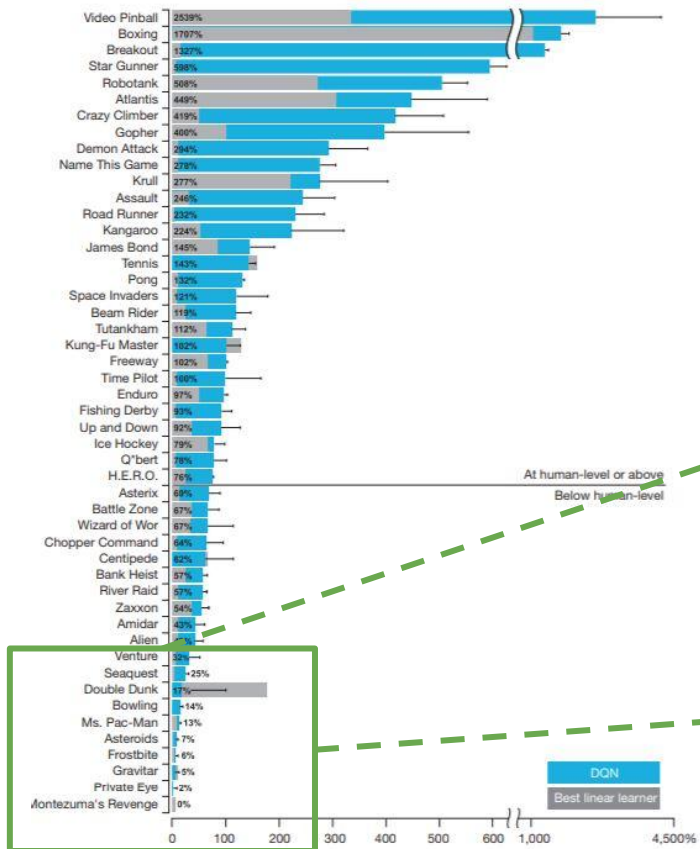
Gabe Margolis

Motivation: DQN and Atari



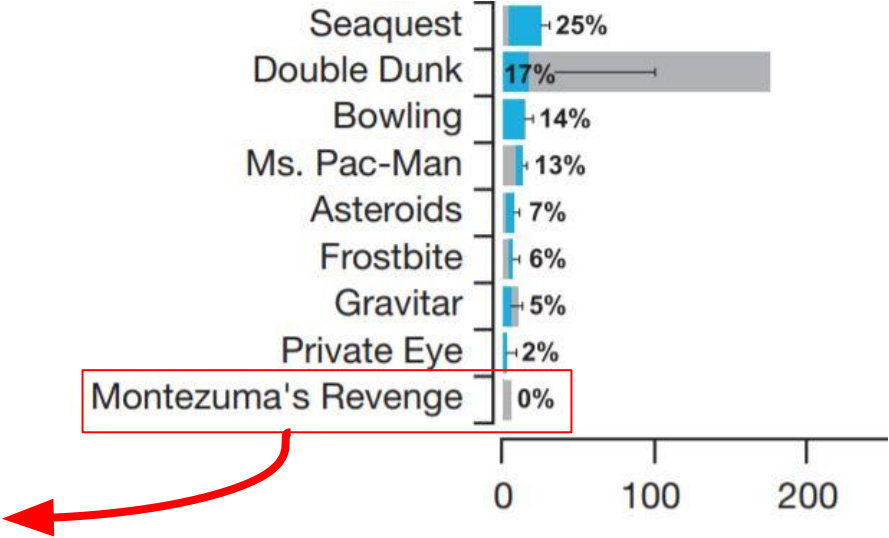
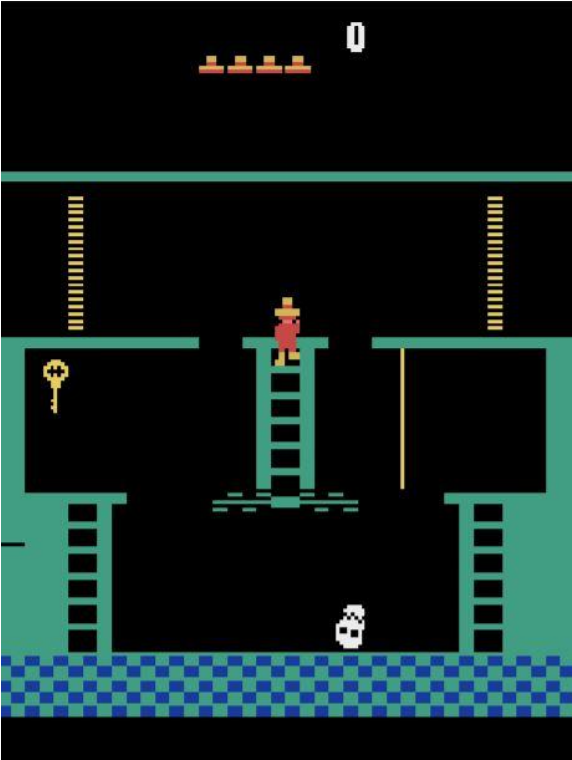
Mnih et al. (2015)

Motivation: DQN and Atari

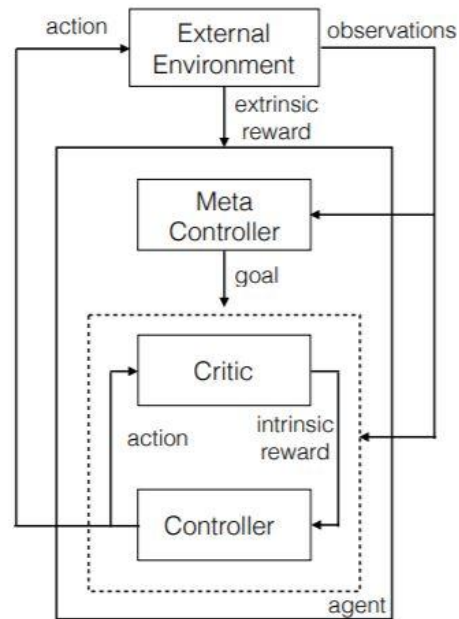
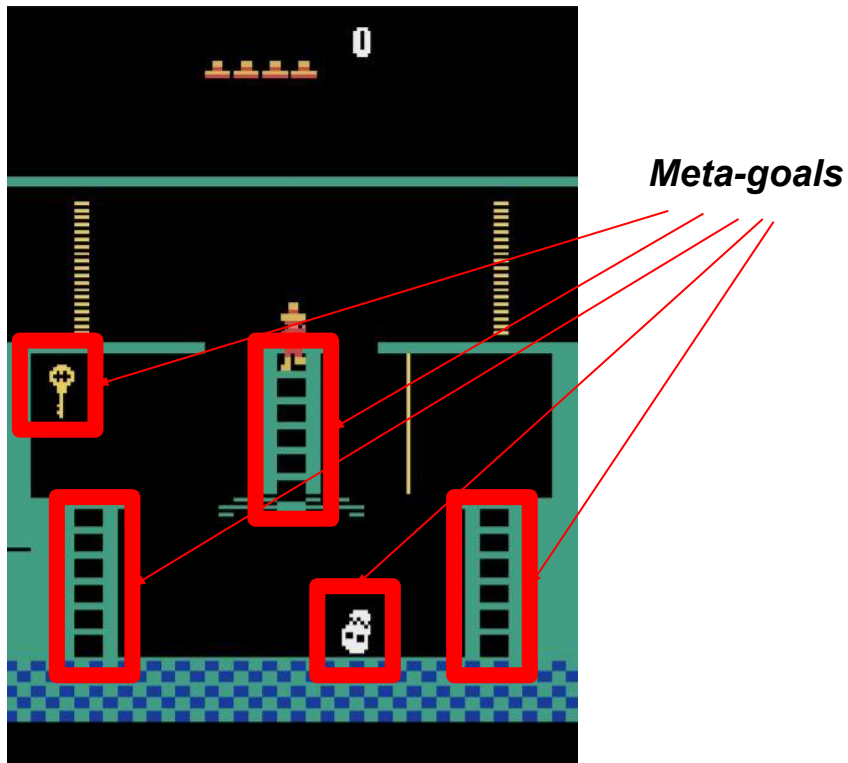


DQN Fails to Learn from Sparse, Delayed Reward

Montezuma's Revenge (Atari)

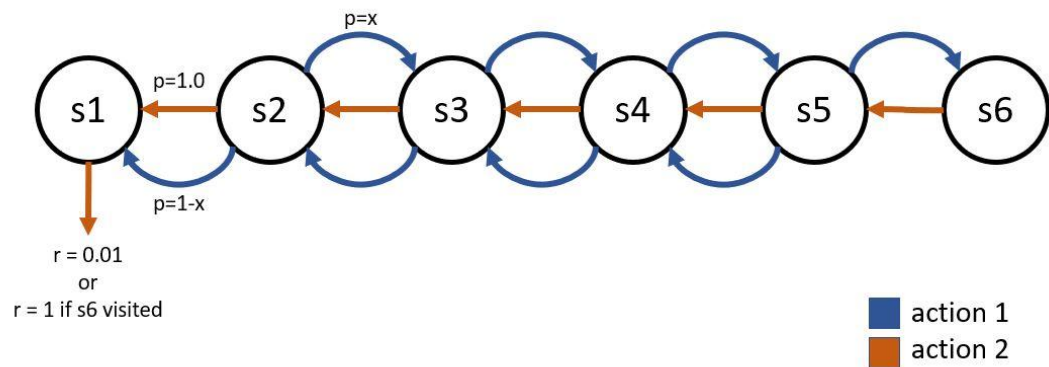


Hierarchical Deep Reinforcement Learning (hDQN)



hDQN Architecture (Kulkarni et al. 2016)

Sparsity Simplified: Discrete Stochastic Decision Process



Replication of Results

Kulkarni et al.

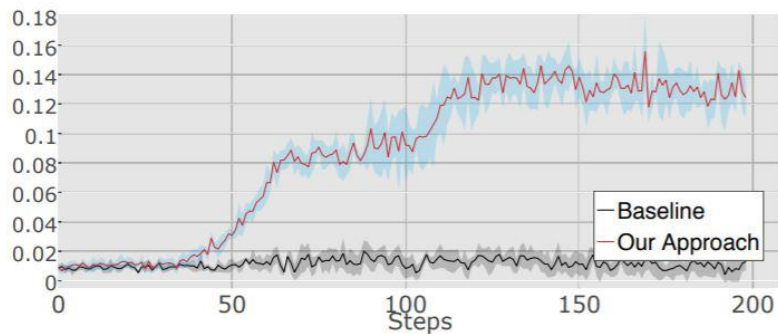
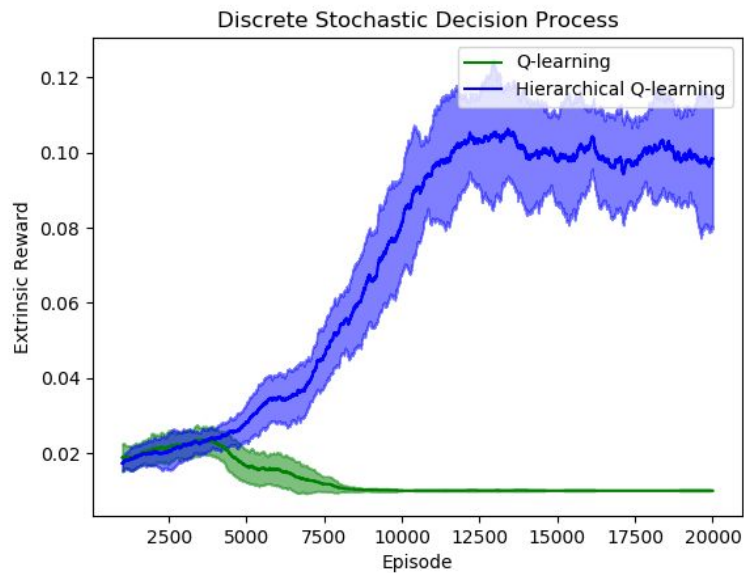


Figure 3: Average reward for 10 runs of our approach compared to Q-learning.

Ours



Replication of Results

Kulkarni et al.

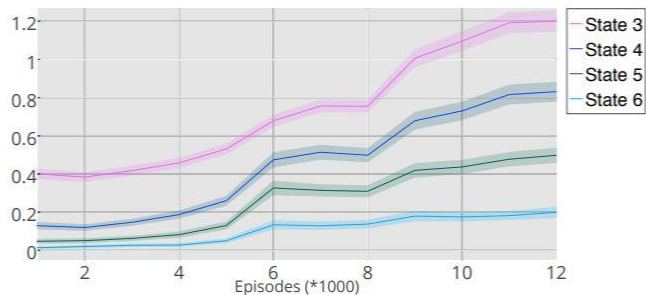
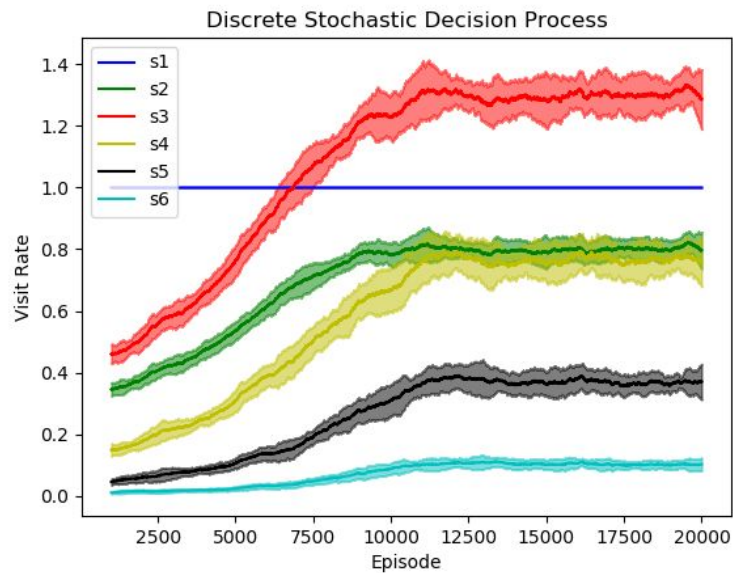
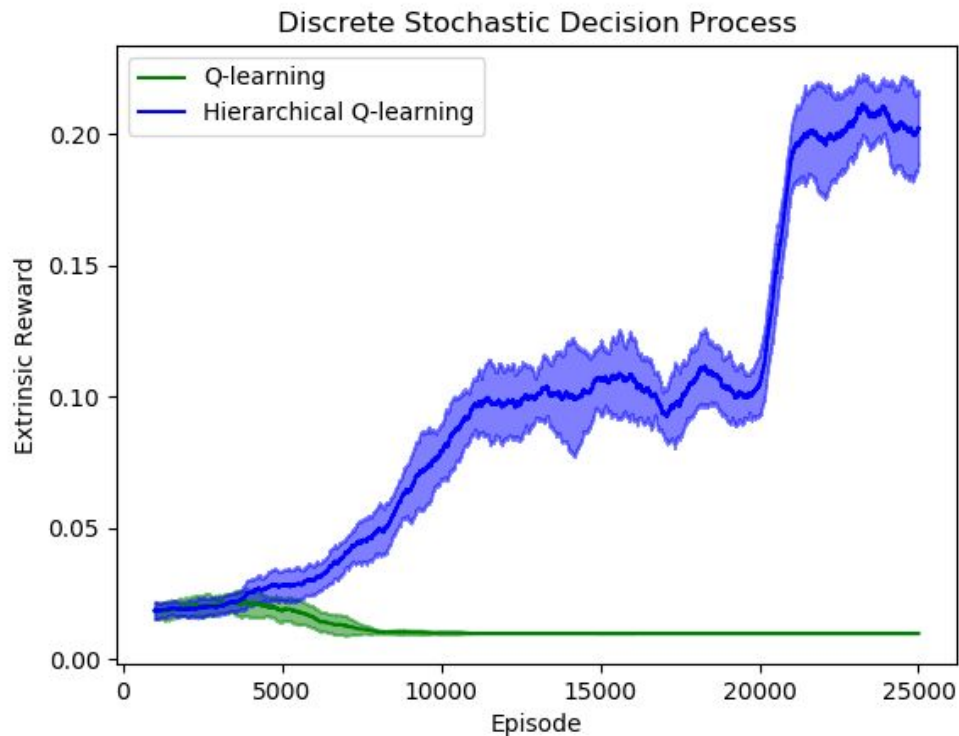


Figure 4: Number of visits (for states s_3 to s_6) averaged over 1000 episodes. The initial state is s_2 and the terminal state is s_1 .

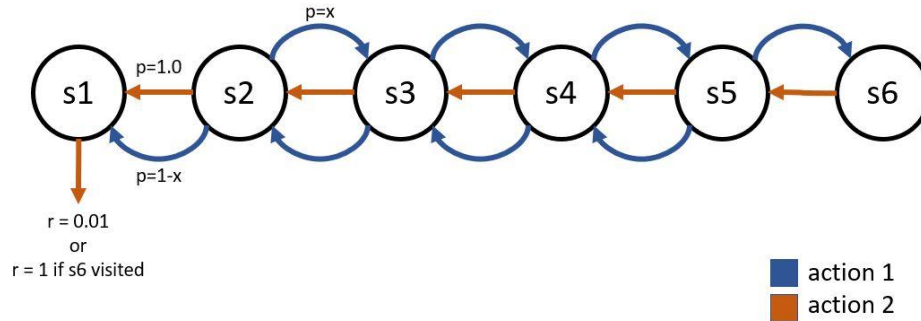
Ours



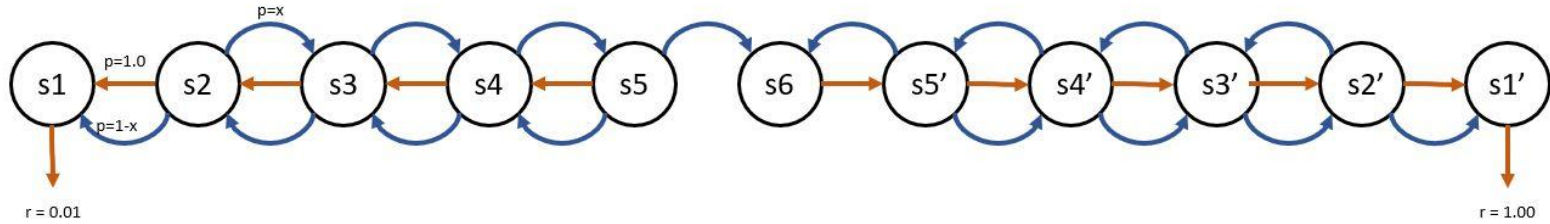
Optimal Reward Achieved under Pure Exploitation



Analysis: Sparsity vs Observability

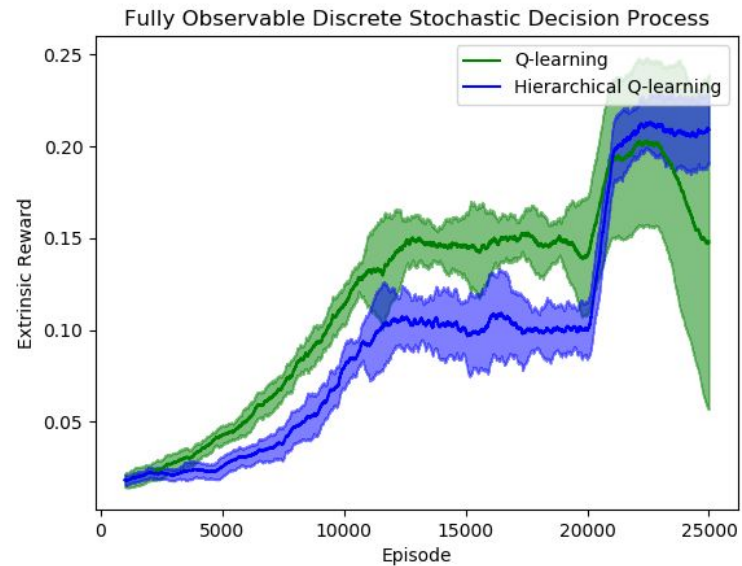
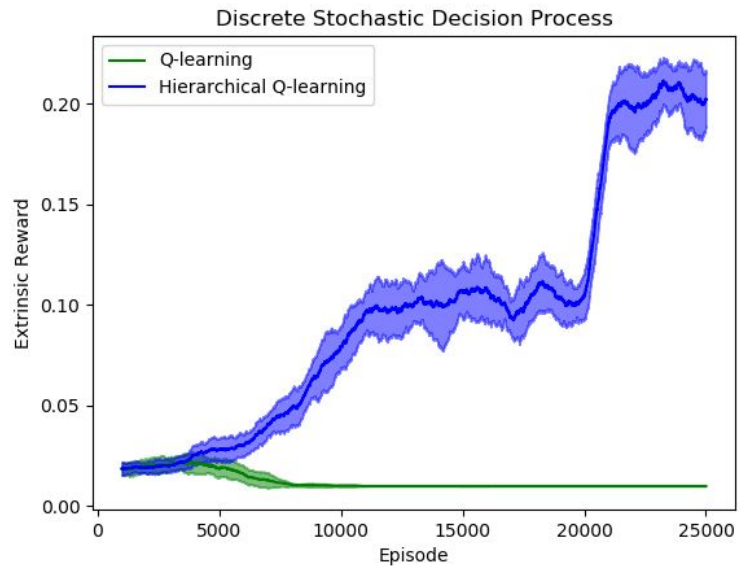


Analysis: Sparsity vs Observability



■ action 1
■ action 2

Experiments in Fully Observable Environment



Summary

- Successfully replicated h-DQN paper results
- Separated impacts of sparsity and partial observability/temporal abstraction
- Beat Montezuma's Revenge! (If I were Google)