
Deep Reinforcement Learning with Double Q-learning

Flappy Bird Hacking using Deep Reinforcement Learning

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The best agent

Introduction

Paper: *Deep Reinforcement Learning with Double Q-learning*

About this project

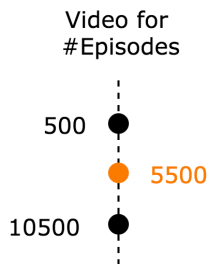
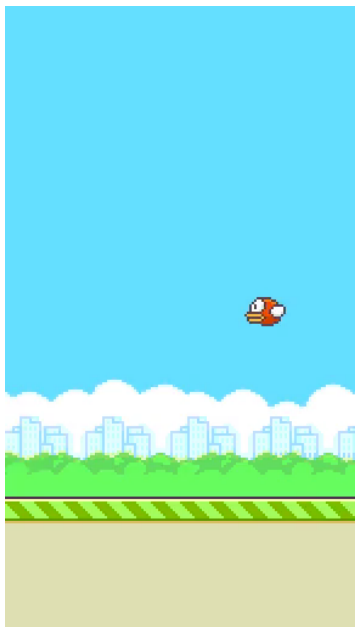
We implement DDQN on PLE FlappyBird environment in PyTorch

Why DDQN?

DDQN is proposed to solve the overestimation issue of Deep Q Learning (DQN).

Apply separate target network to choose action,
reducing the correlation of action selection and value evaluation.





Implementation: Environment

Valid Actions

Up causes the bird to accelerate upwards.

Terminal states (`game_over`)

If the bird makes contact with the ground, pipes or goes above the top of the screen the game is over.

Rewards

For each pipe it passes through it gains a positive reward of `+1`. For each frame the live bird receives reward of `+0.1`. Each time a terminal state is reached it receives a negative reward of `-1`.

Experimental Setup: Hyperparameters

Optimizer

Learning Rate

Discount Factor

Replay Buffer Size

Total Episodes

Target Frequency

Epsilon

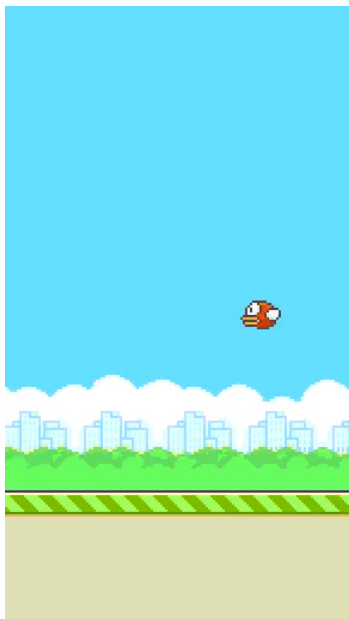
Batch Size

Initial Observed
Episodes

Epsilon Discount

Screen Width

Screen Height



Video for
#Episodes

500

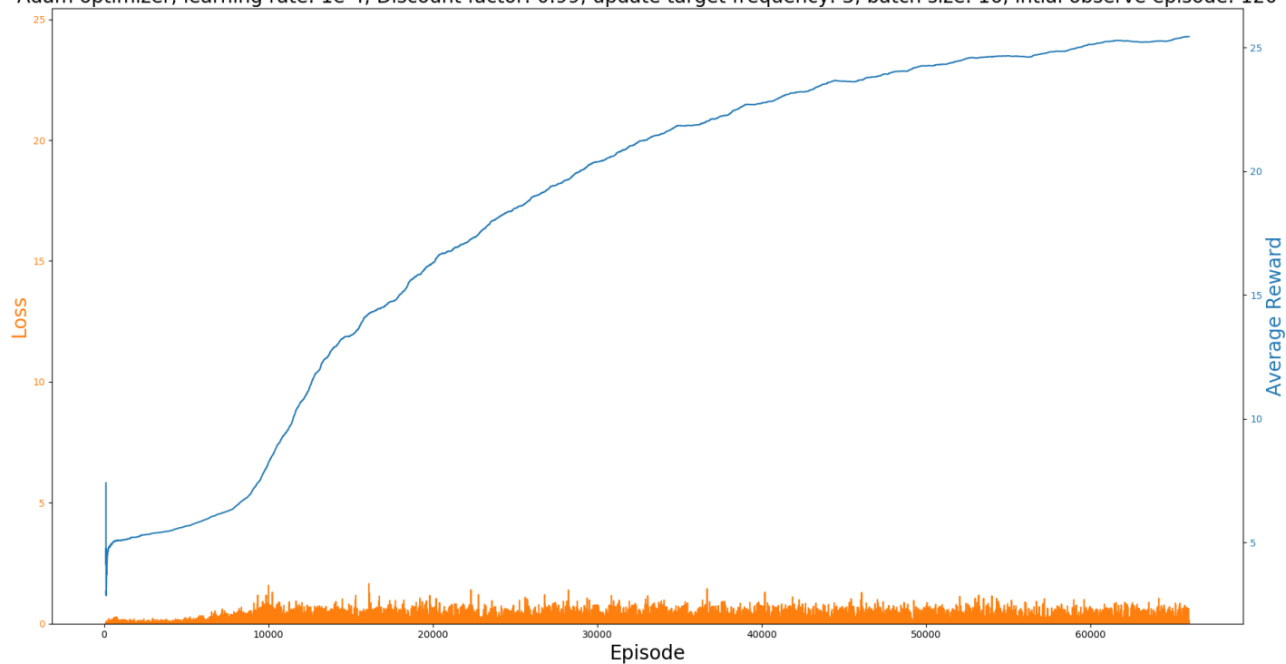


5500

10500

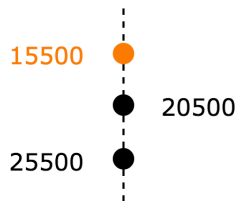
Experimental Results: Best

Adam optimizer, learning rate: $1e-4$, Discount factor: 0.99, update target frequency: 3, batch size: 16, initial observe episode: 120



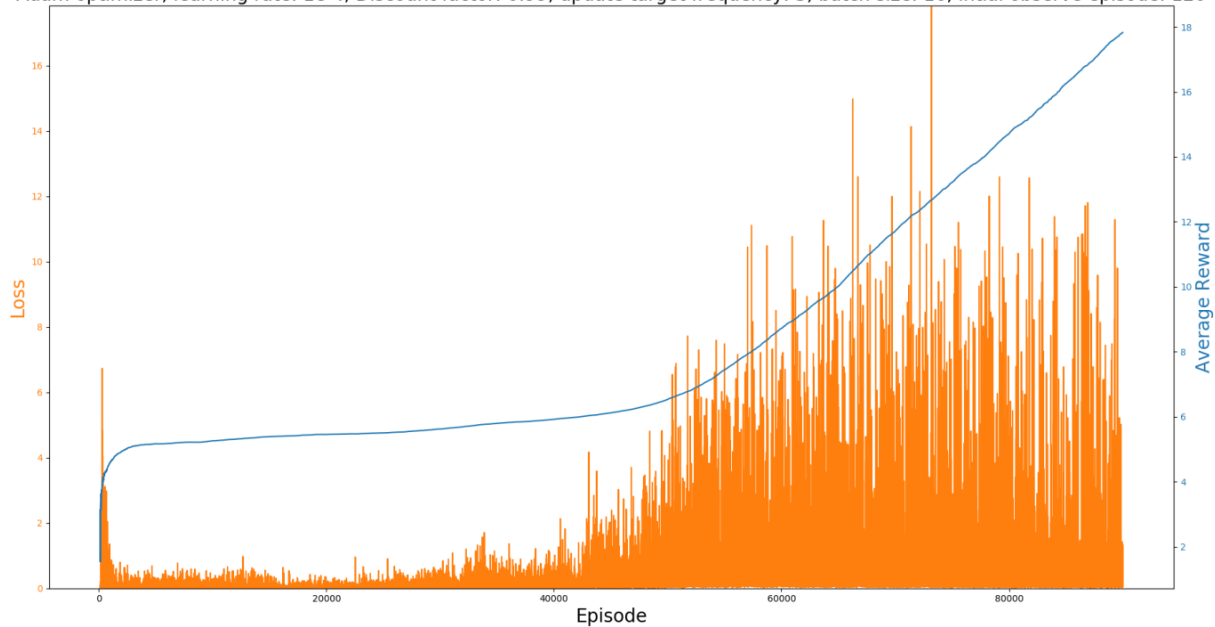


Video for
#Episodes



Experimental Results: Worst

Adam optimizer, learning rate: $1e-4$, Discount factor: 0.99, update target frequency: 3, batch size: 16, initial observe episode: 120





Video for
#Episodes

15500



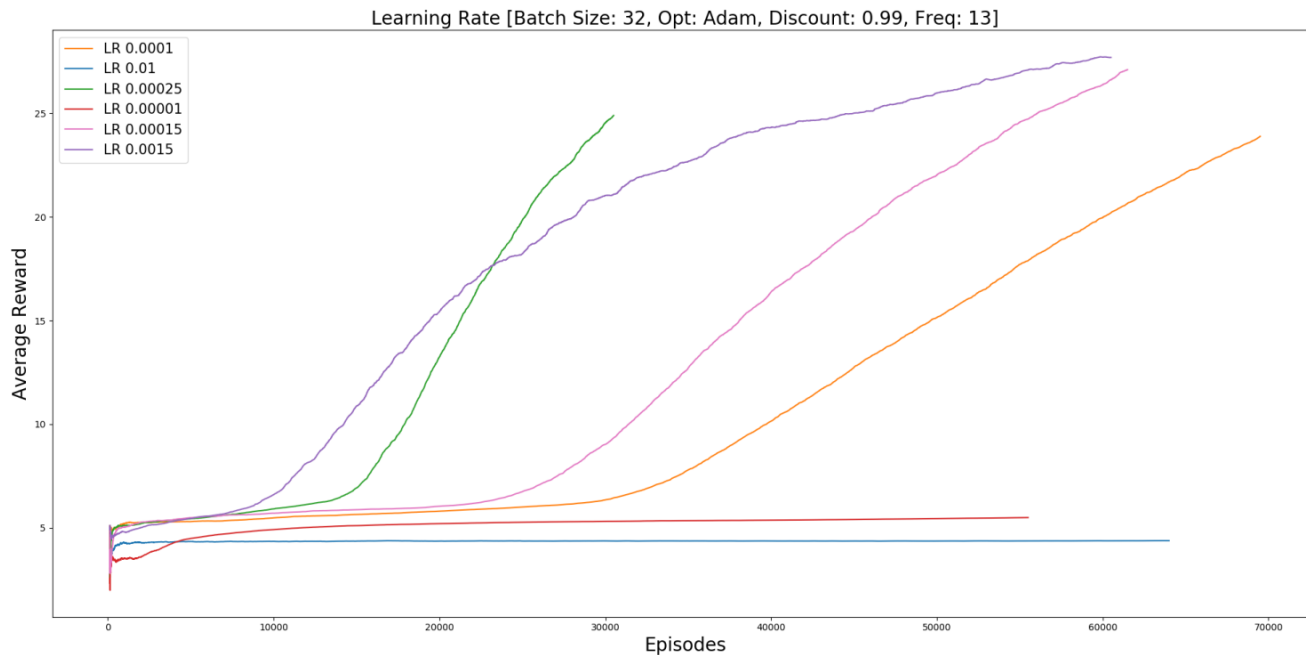
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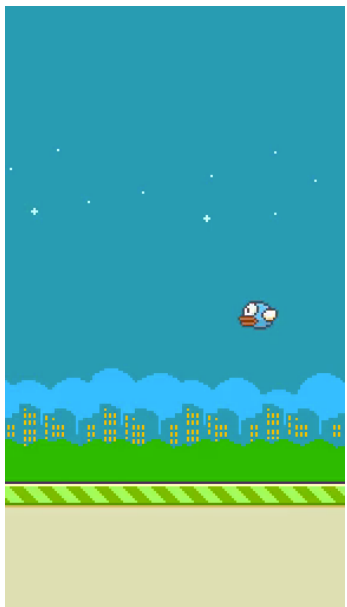


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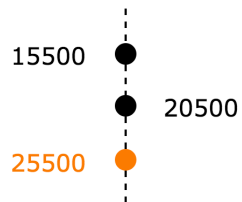


Learning Rate Analysis

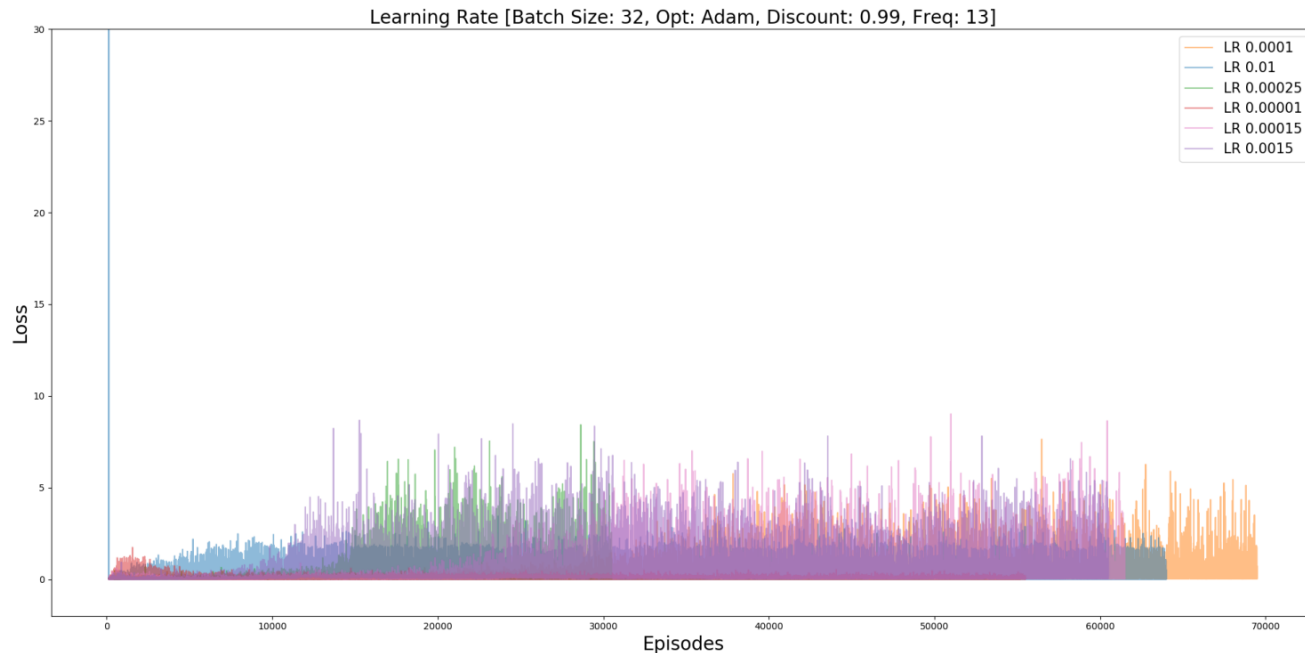


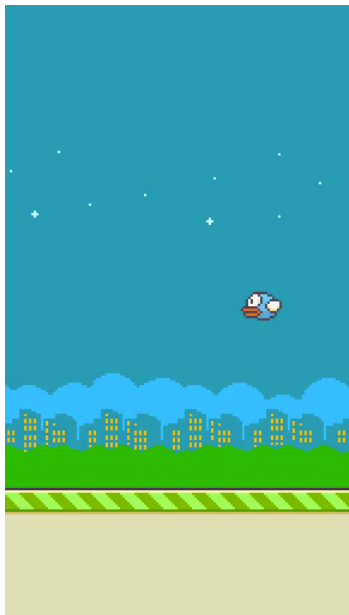


Video for
#Episodes



Learning Rate Analysis





Video for
#Episodes

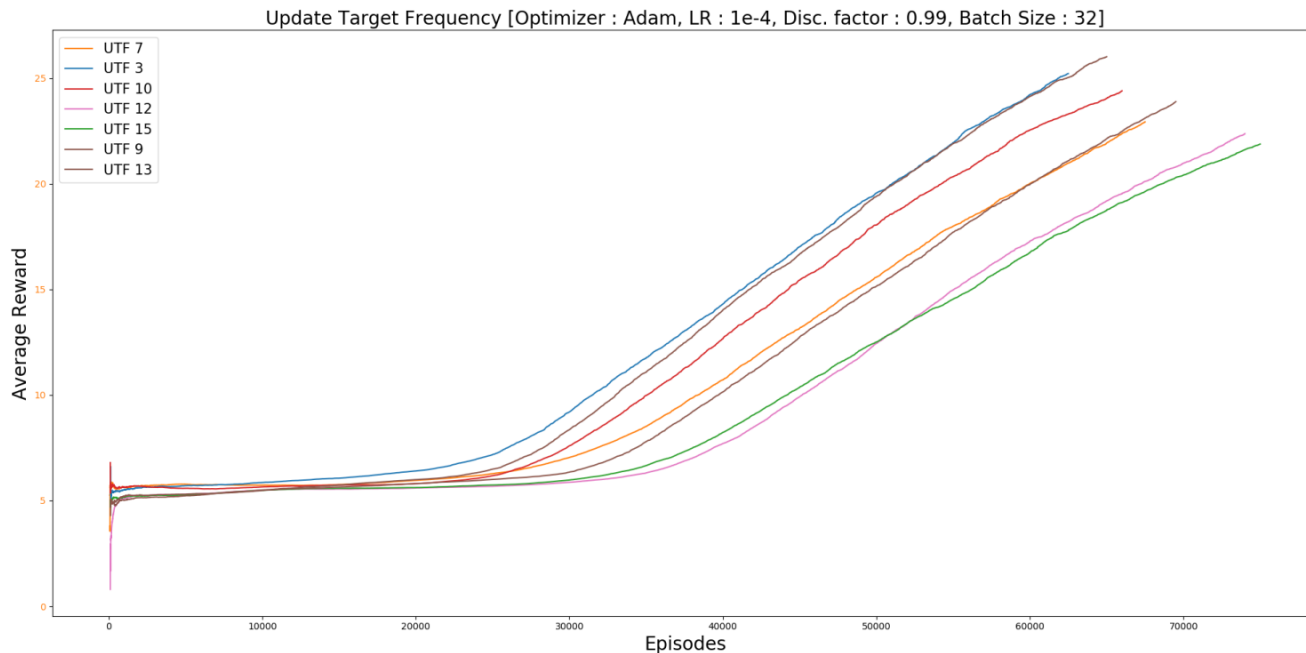
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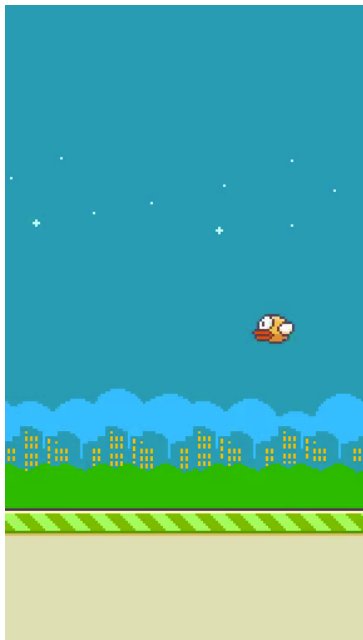


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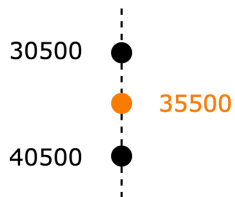
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Update Target Frequency Analysis

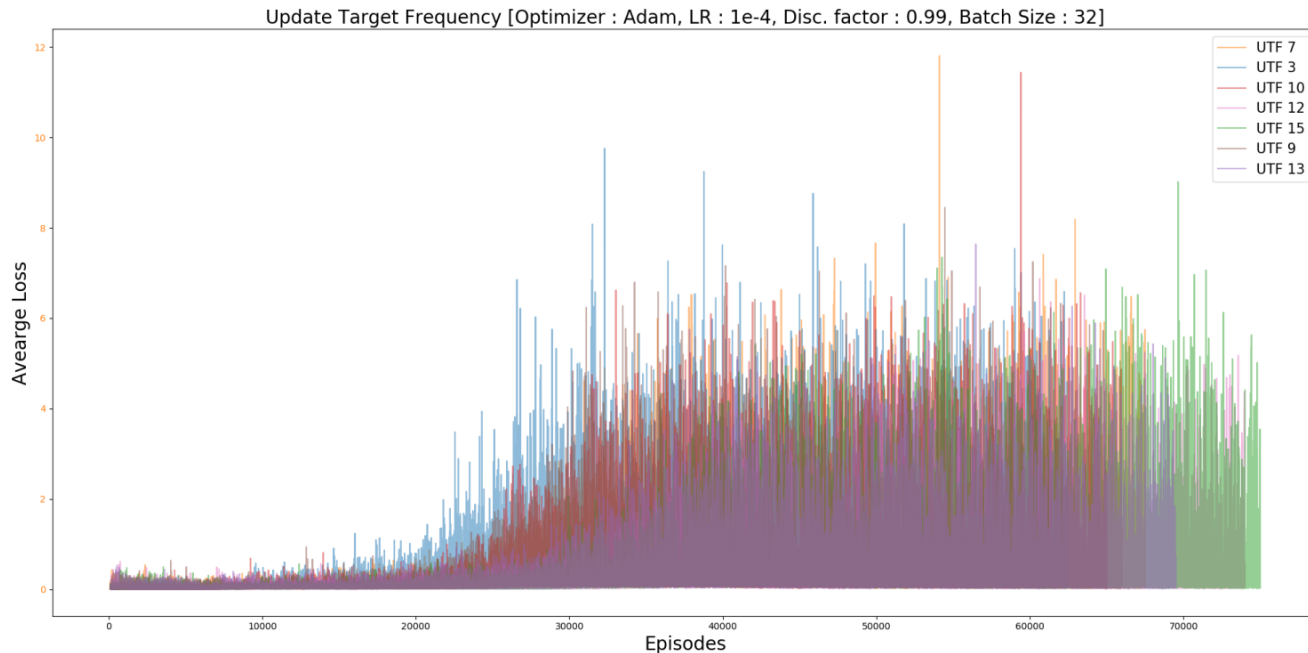


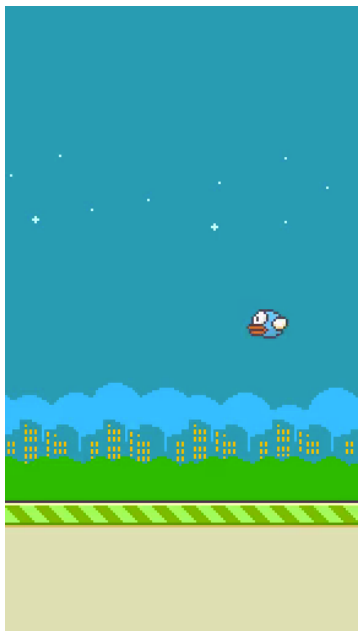


Video for
#Episodes



Update Target Frequency Analysis





Video for
#Episodes

30500



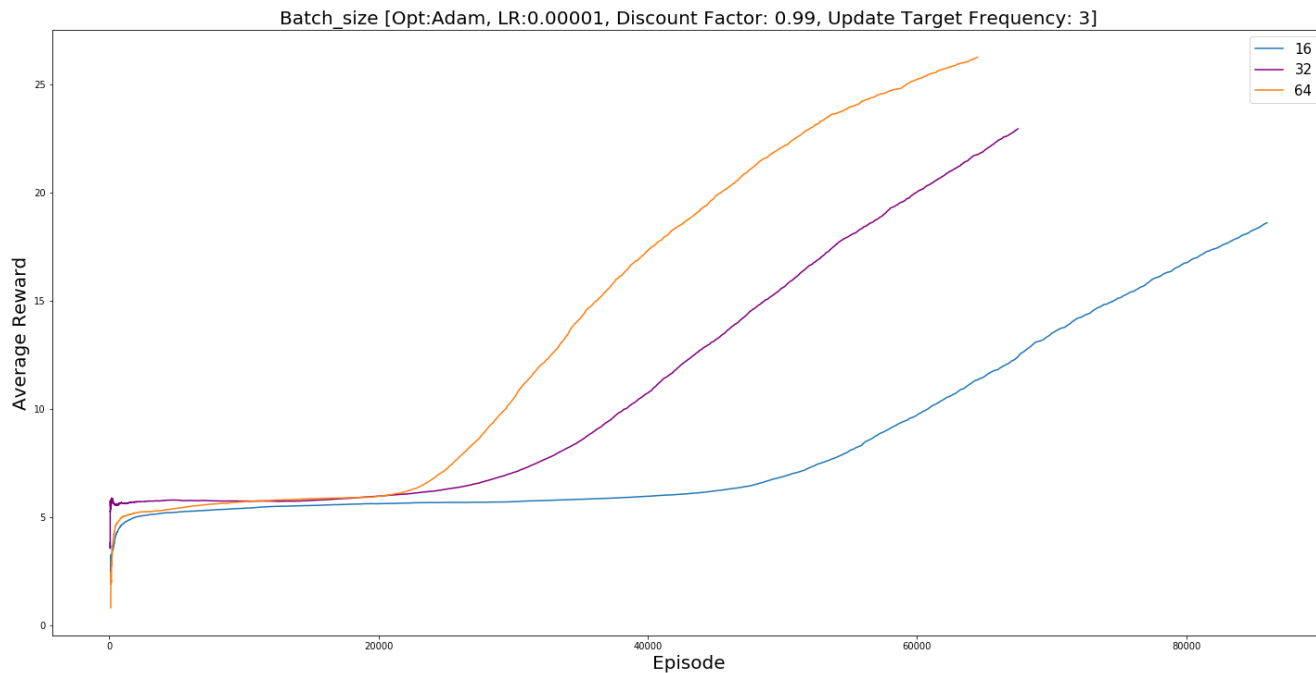
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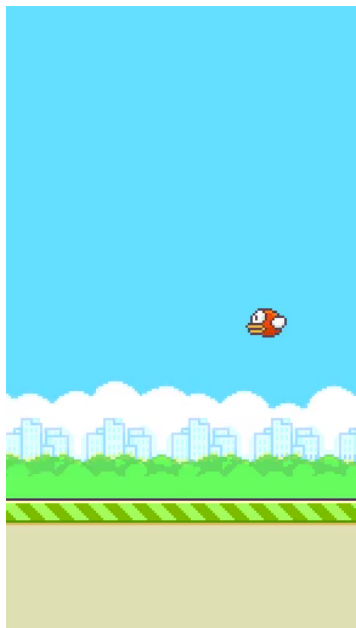


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Batch Size Analysis





Video for
#Episodes

45500

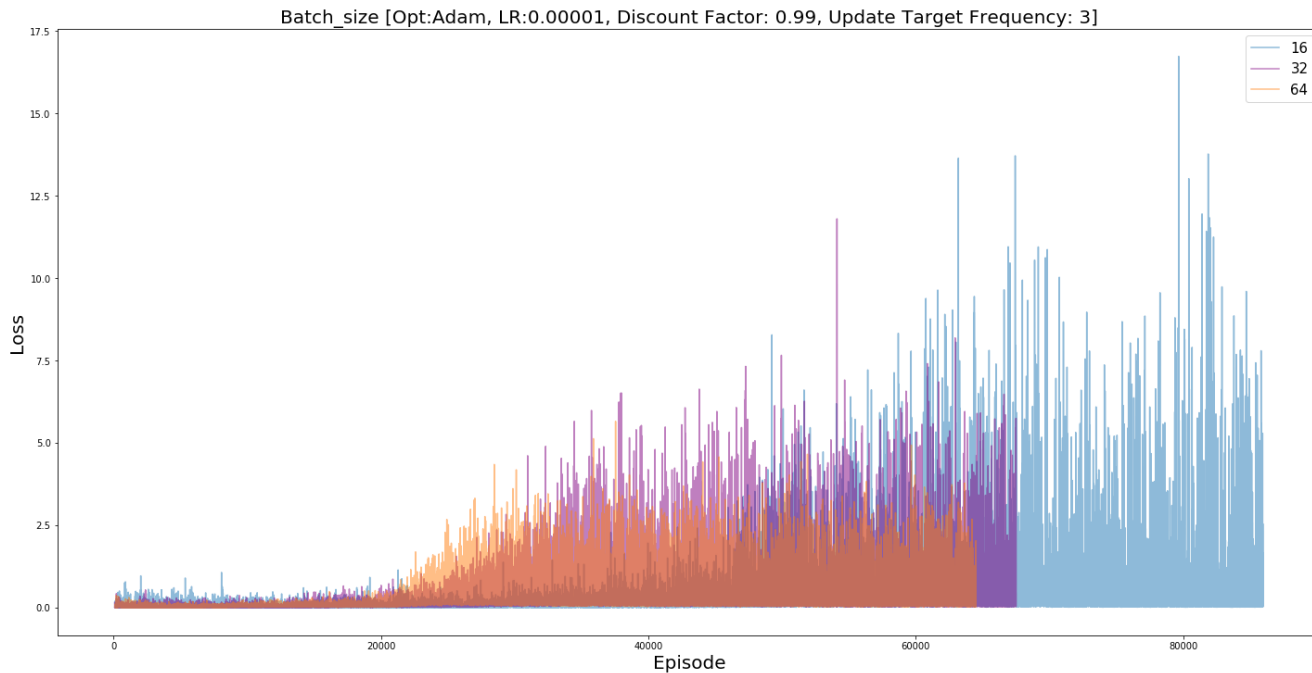


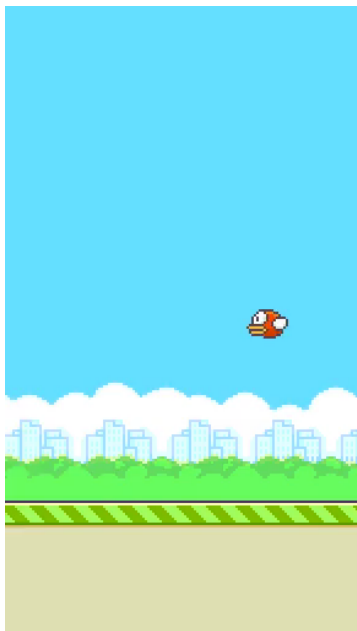
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55500



Batch Size Analysis





Video for
#Episodes

45500



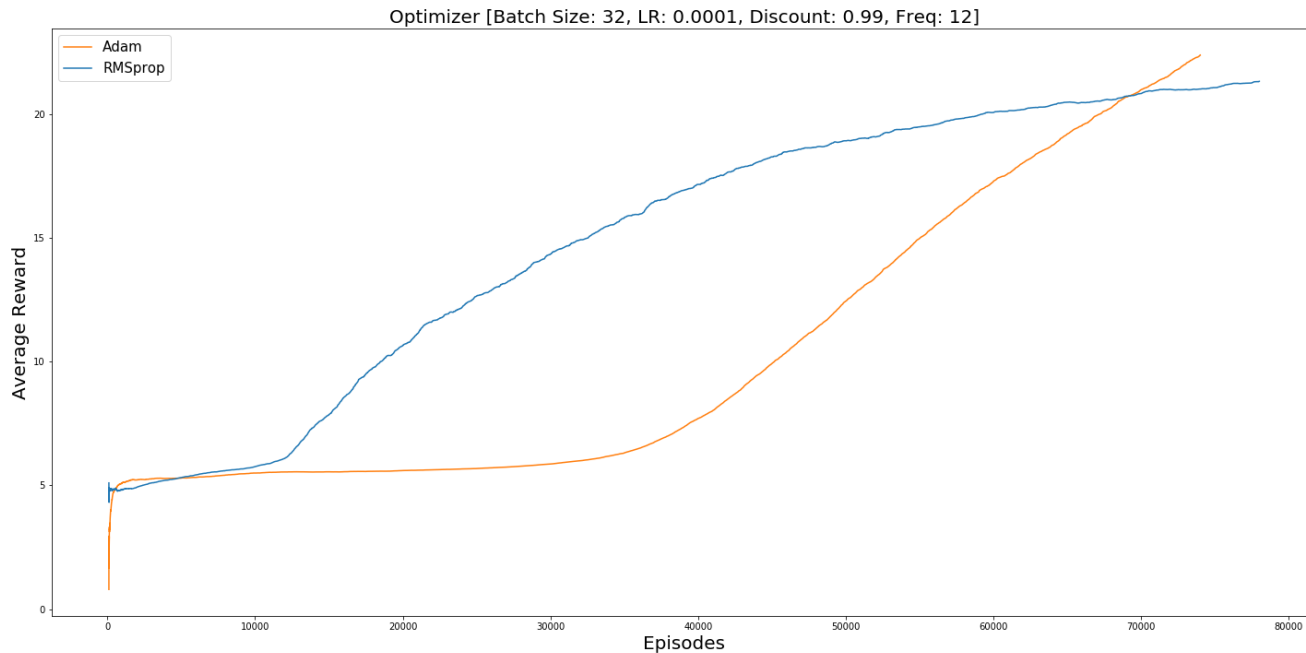
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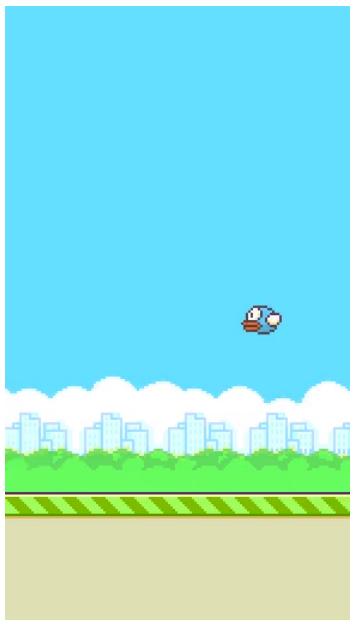


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Optimizer Analysis





Video for
#Episodes

45500



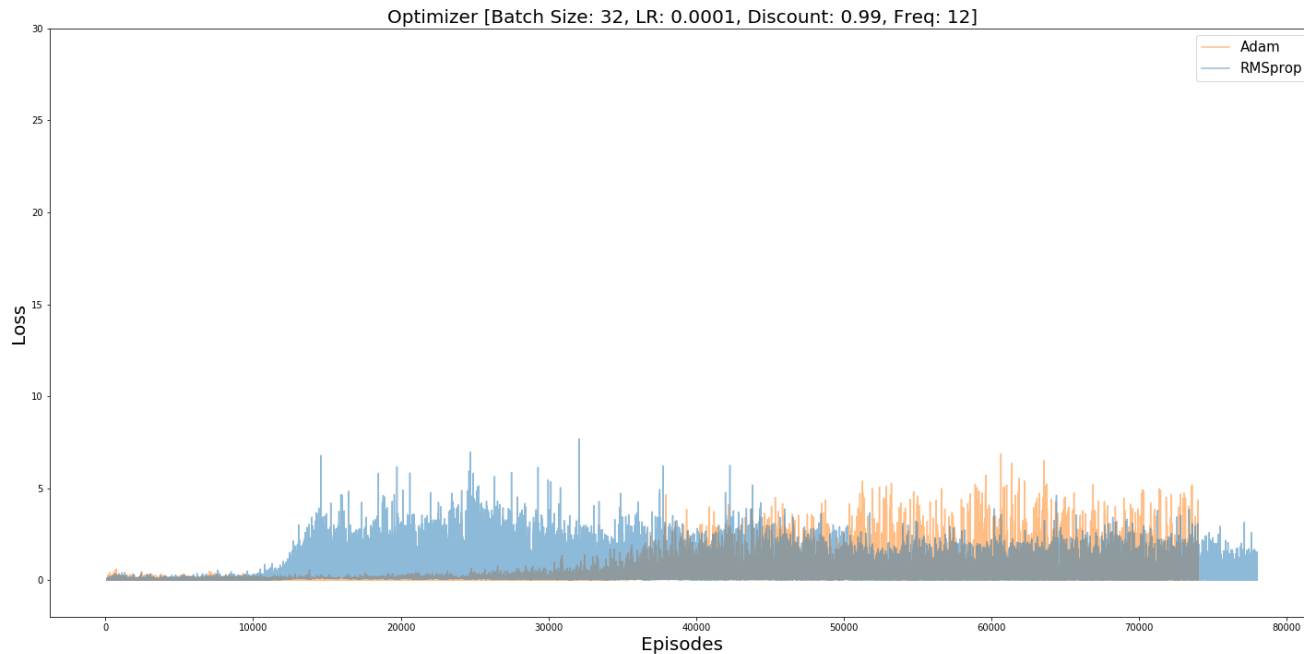
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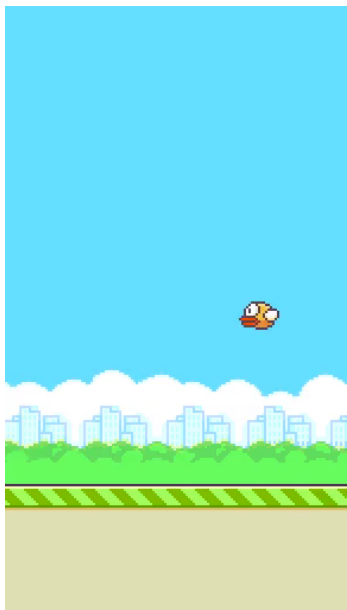


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Optimizer Analysis





Video for
#Episodes

60500

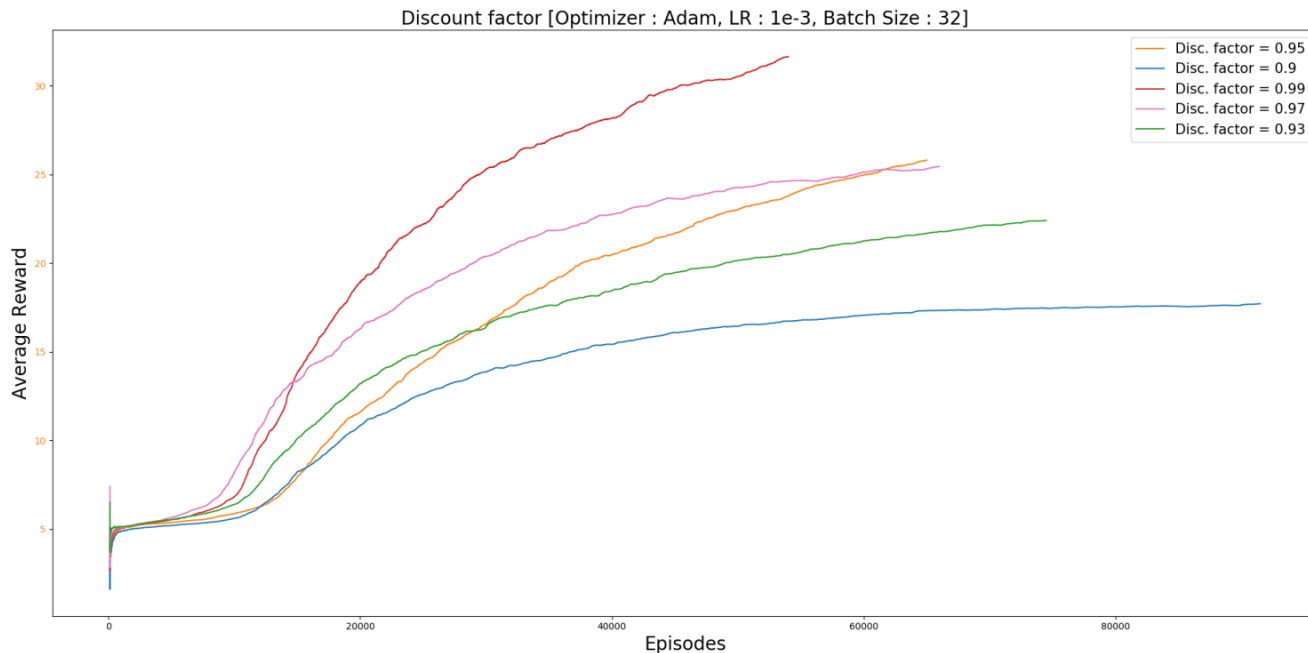


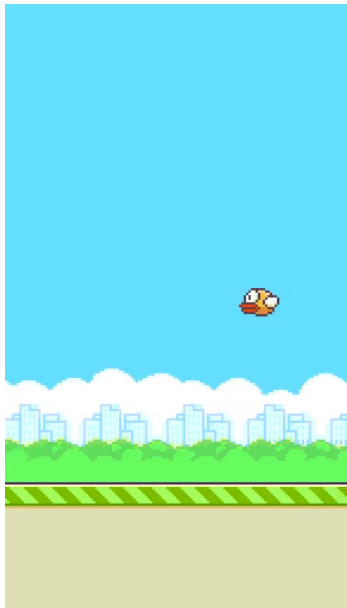
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70500



Discount factor Analysis





Video for
#Episodes

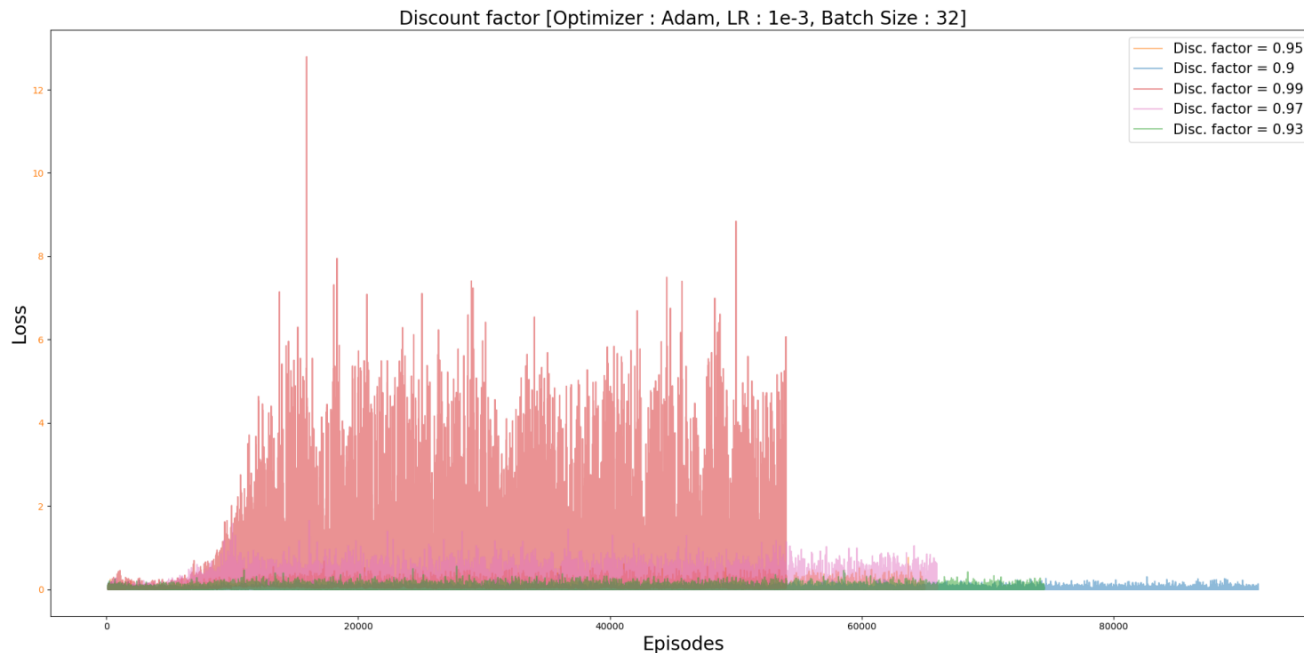
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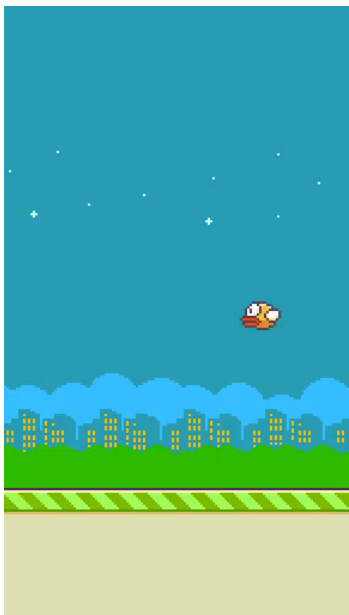


65500

70500

Discount factor Analysis





Questions?

Resources Used

1. Blue Waters ~ 1300 Computation Hours
2. Google Colab (Free GPU) ~ 30 Hours

