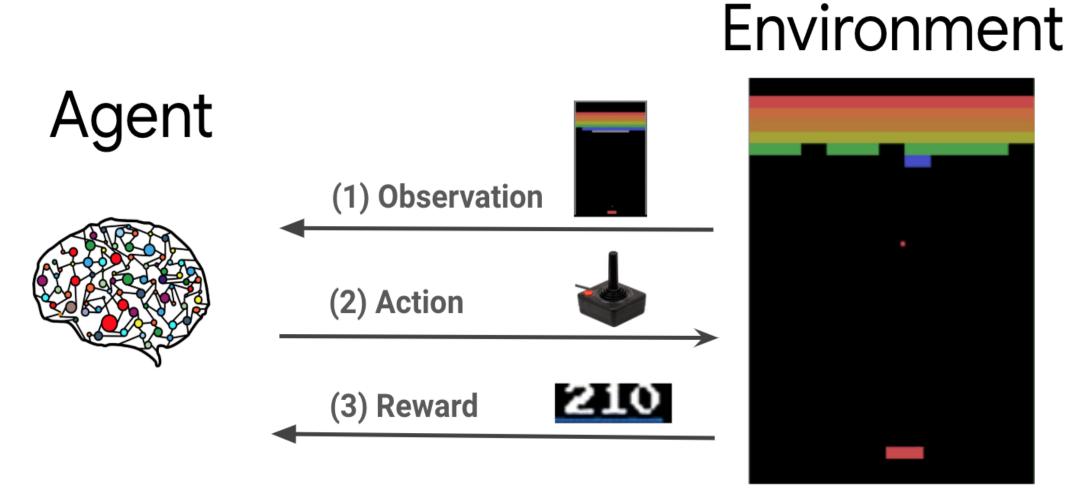




# MinAtar

MinAtar, introduced by researchers at the University of Alberta, Google Brain and Deepmind simplified 5 games representational complexity to focus more on the behavioral challenges.

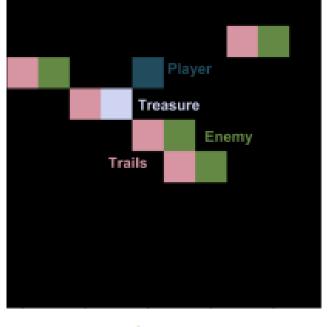


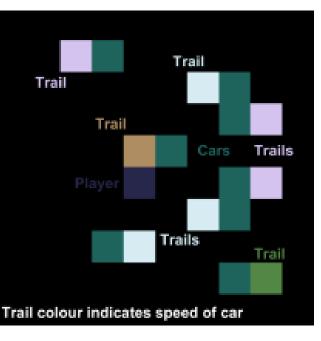
### Overview

Current implementation of these games are slow hence we optimized and refactored the whole codebase by reducing the overall training time. We also propose a new hyperparameter approach that computes N average returns for each hypersetting and records the highest average performance. We then proceed to perform additional K runs with different seeds on these best hyper-settings and then construct and report the mean learning curves > Testbed for researchers to test their Reinforcement Learning and Artificial Intelligence agents.

### Environments

Five Atari games which play out on a 10x10 grid. State representation is 10x10xn where each of the n channels correspond to a game-specific object.





Space Invader

Asterix

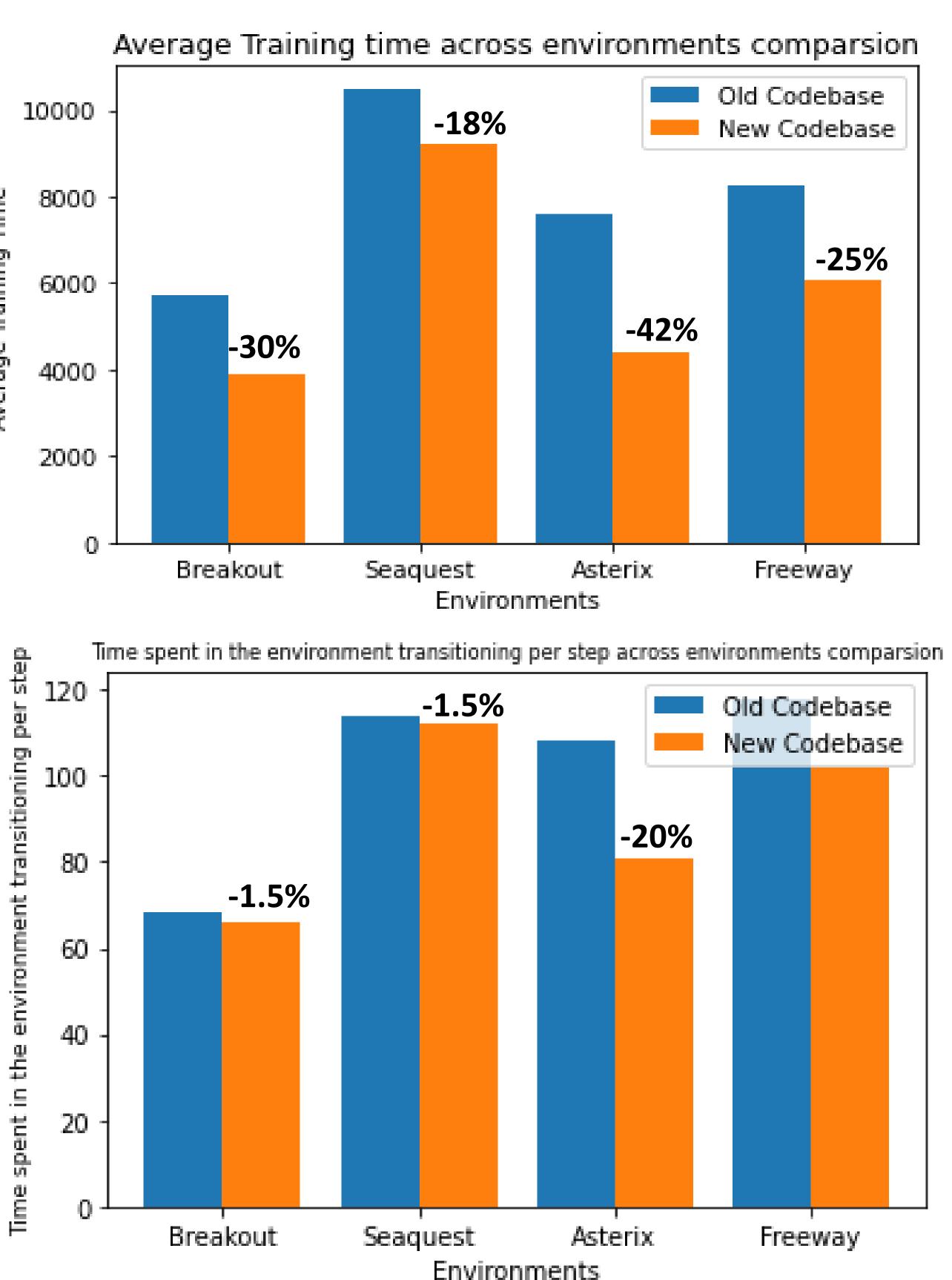
Freeway

### **Making Reinforcement Learning Experiments** UNIVERSITY OF more Reproducible and Computationally Efficient ALBERTA **Reinforcement Learning and Artificial Intelligence Lab** Robert Joseph, Martha White, Adam White, Samuel Neumann {rjoseph1, whitem, amw8, sfneuman}@ualberta.ca

## Results

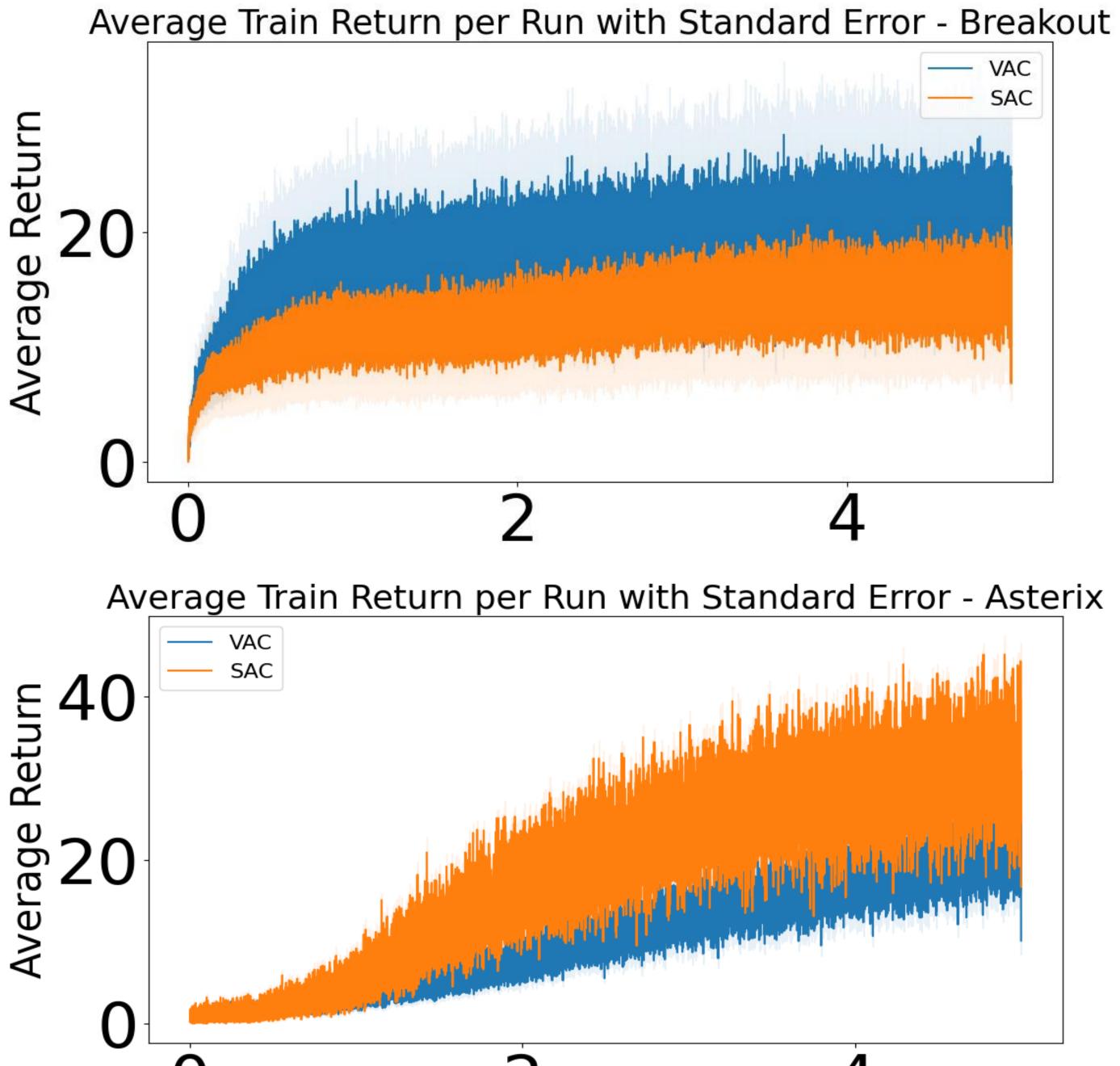


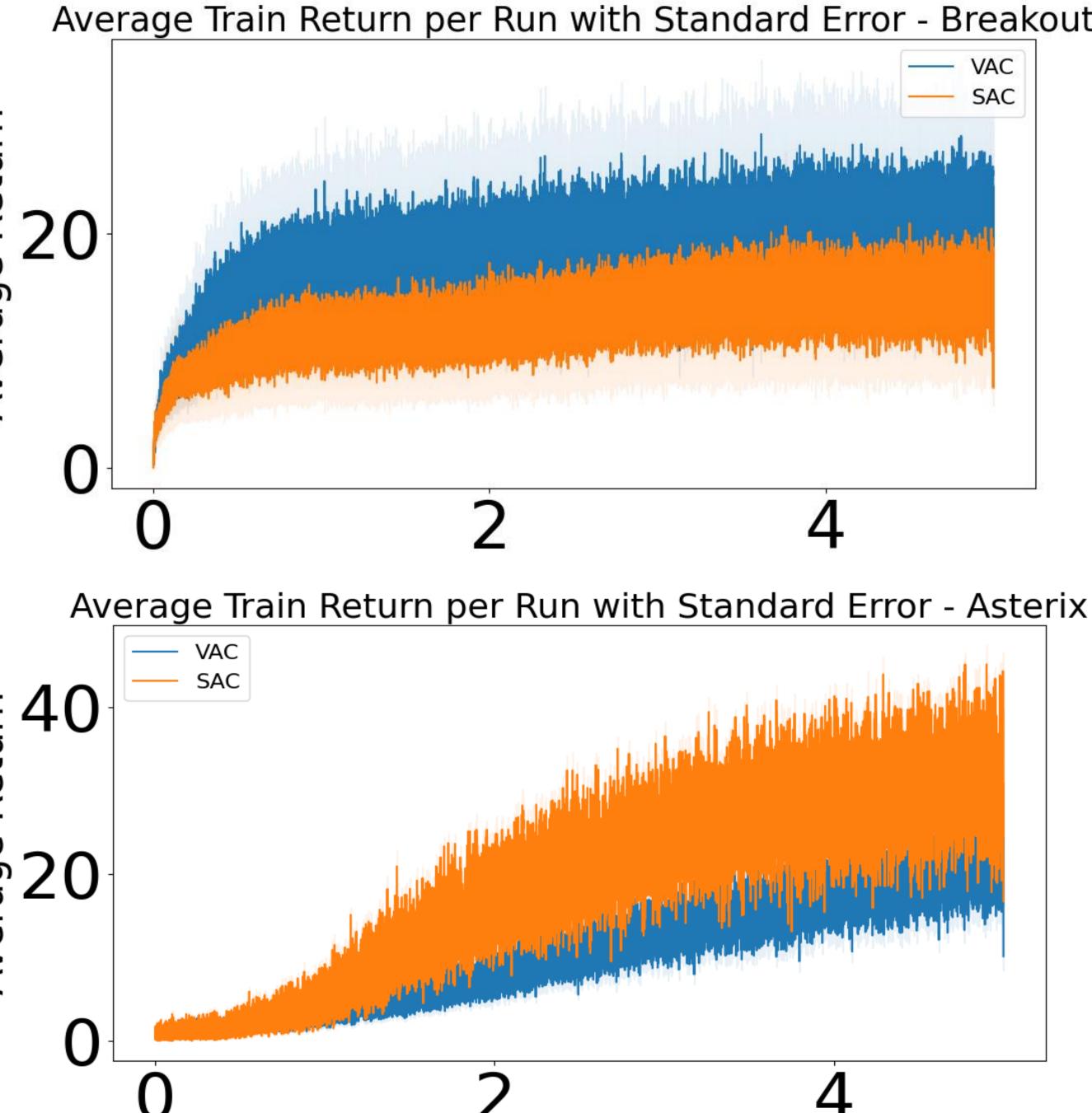
Environments were trained on the Soft Actor Critic agent. > Fixed hyperparameter setting – Chosen randomly. > 1.5 Million frames and 2500 steps per episode. Added Just in Time Compilation (JIT) too for the Python implementation using Numba. Added more agents to the codebase such as Vanilla Actor E Critic (VAC), Deep Q-network (DQN), SAC etc.



# References

Young, K. Tian, T. (2019). MinAtar: An Atari-Inspired Testbed for Thorough and Reproducible Reinforcement Learning Experiments.





- hyperparameter approach. hyperparameter sensitivity.

- Implement this in other languages.

### Benchmark

Comparison between the SAC and VAC agents. 5 Million frames and 2500 steps per episode to compare against the original benchmark in MinAtar

Hyperparameter sweep using the new proposed

This method is more robust and takes into account

### **Future work**

> Compare various other algorithms such as DQN to Double DQN and produce more benchmarks.

Prove theoretical guarantees on the new proposed Hyperparameter approach.

Add more environments to MinAtar.

Scan QR code for more information.

