



Autonomous Vehicle Simulator Trained with Lumina RCL vs. Artificial Neural Network

Advancements in Autonomous Driving with Lumina AI's Random Contrast Learning

Lumina AI has demonstrated its potential for groundbreaking advancement in the domain of autonomous driving. To evaluate our performance, we have rigorously compared Lumina's CPU-based application PrismRCL with traditional transfer learning neural networks in autonomous simulators.

The Distinctive Edge of Lumina RCL

Our experiments demonstrate the efficacy of Lumina RCL in a controlled simulation environment. Equipped with an array of eight sensors, the autonomous vehicle, guided by Lumina RCL, learned to navigate the track with precision. The system gradually enhanced its performance, achieving higher speeds while maintaining its trajectory. We ran a one-to-one comparison with an artificial neural network (ANN), ensuring the simulator applied positive and negative reinforcement equally.

The data presents a compelling picture: in 100 simulated trials, PrismRCL - our CPU-based application of Lumina RCL - completed a full lap, equivalent to 4,500 simulation length units, with an impressive 81% success rate after only 15,000 training steps. This beat the 23% success rate achieved by the neural network model, which required ten times the number of training steps.

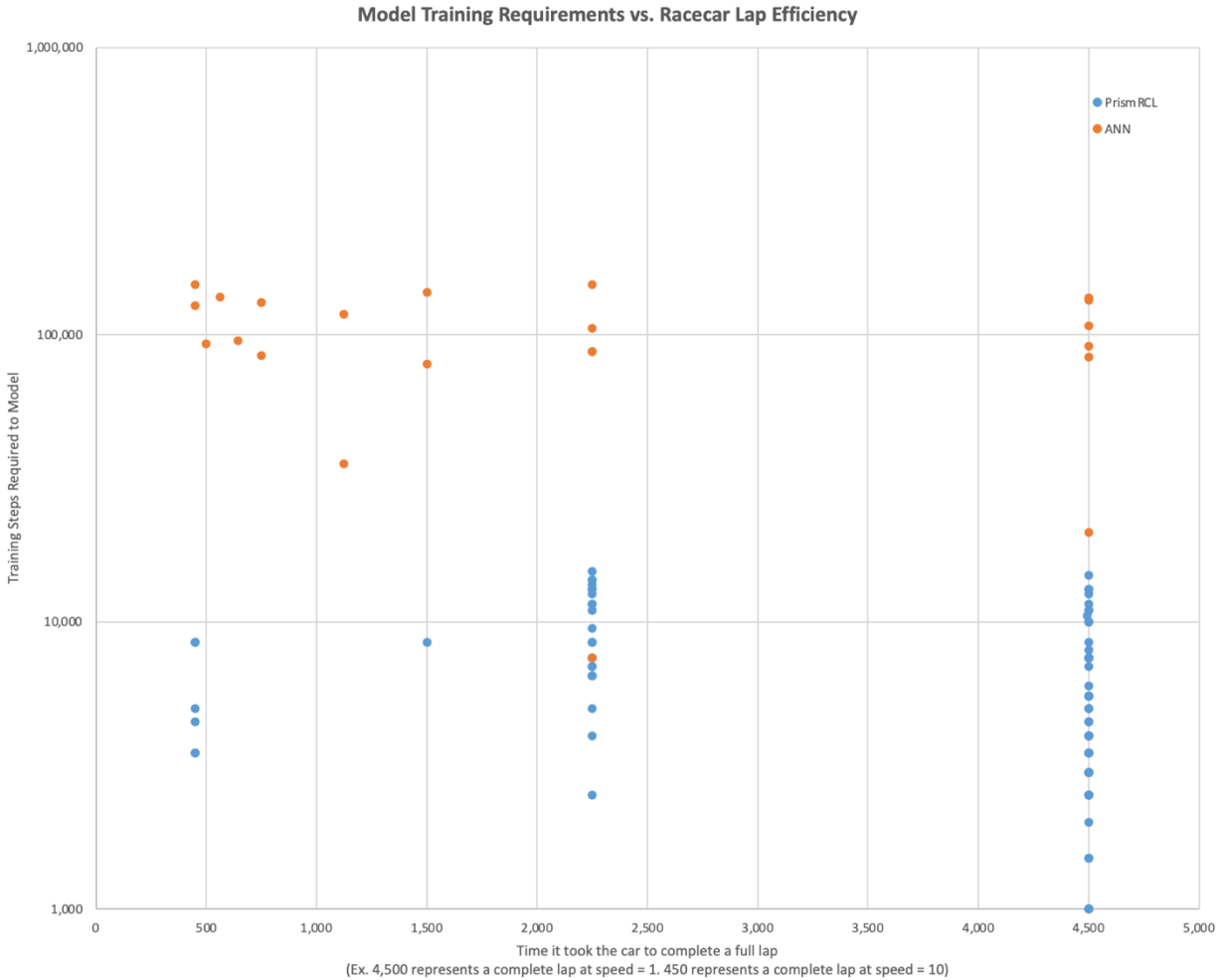
High Speed Trials: PrismRCL vs. Artificial Neural Network

In high speed trials, conducted at a maximum speed of ten, vehicles utilizing PrismRCL completed a lap 7 times, with an average of 5,583 training steps. In contrast, the artificial neural network model was only able to achieve a full lap once, with an average of 126,500 training steps required. This demonstrates PrismRCL's efficiency and success under demanding conditions.

The Efficiency of Random Contrast Learning

The PrismRCL model both achieved a higher rate of success and did so requiring ten times fewer training steps than neural network models.

Moreover, PrismRCL utilizes CPU, in contrast to more expensive AI chipsets employed by neural networks. The combination of superior performance, lower requirements of data and requirements of cheaper and more available hardware positions Lumina RCL to transform the field of machine learning.



This chart demonstrates the success of our racecar models in achieving the goal of completing laps, while also showcasing their ability to reach high speed (max speed=10)

The graph above illustrates the comparative study, indicating the training steps required by both PrismRCL (in blue) and the neural network (in orange). PrismRCL not only requires fewer training steps to complete a lap, but also is capable of achieving faster lap completion times, highlighting its efficiency and effectiveness in comparison to the neural network.

Discover Random Contrast Learning with Lumina AI

Our technology is not just a step forward; it is a leap into the future of machine learning. We are excited to offer a glimpse into the potential that Random Contrast Learning (Lumina RCL) holds for autonomous driving and beyond.

We encourage you to join us in this journey with our 30-day free trial and help us shape the future of machine learning together.