

Dissipating Traffic Jams with Piecewise Constant Control

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Background

- Carbon dioxide makes up 80% of US GHG emissions in 2019
- Transportation sector causes ~35% of total CO₂ emissions and 29% of GHG emissions [5]
- Previous studies show simple traffic management (speed management, shock wave suppression etc.) techniques can reduce emissions on the road by almost 20% [1]

Background

- Wu et al. shows that we can increase average speeds by up to 57% in idealized traffic settings with a small fraction of RL-controlled vehicles [6]
- Sridhar et al. shows that we can reproduce traffic smoothing results with piecewise constant policies (up to $\Delta=140$) [3]
- This suggests we can use RL advice to empower human drivers to mitigate congestion for a near-term deployment

Overview

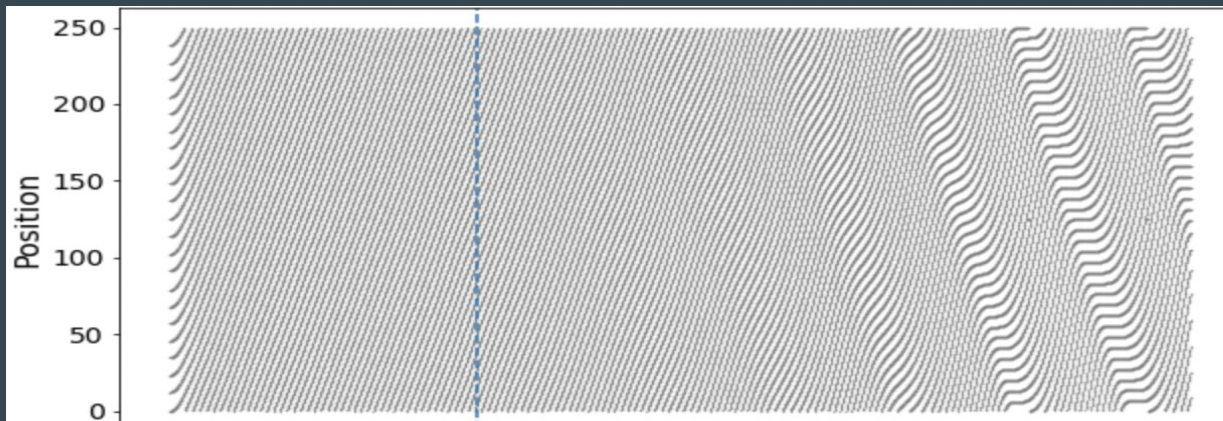
- We extend previous work to more difficult (and realistic) settings:
 - We show that piecewise constant policies (up to $\Delta=100$) can dissipate shock waves even after the formation of a severe traffic jam
 - We show that piecewise constant policies are effective with local reward functions

Experimental Setup

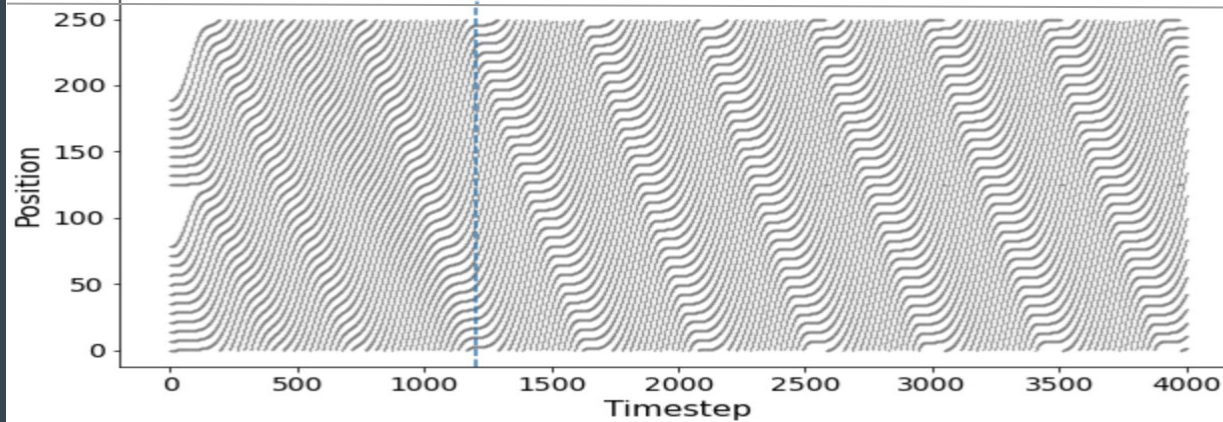
- Single-lane ring road setup
 - 250m circumference
 - 22 vehicles
- Shown to develop traffic waves without the presence of AVs [4]
- Introduce a single agent trained with TRPO with reward function of their own velocity [2]

Traffic Jam Severity

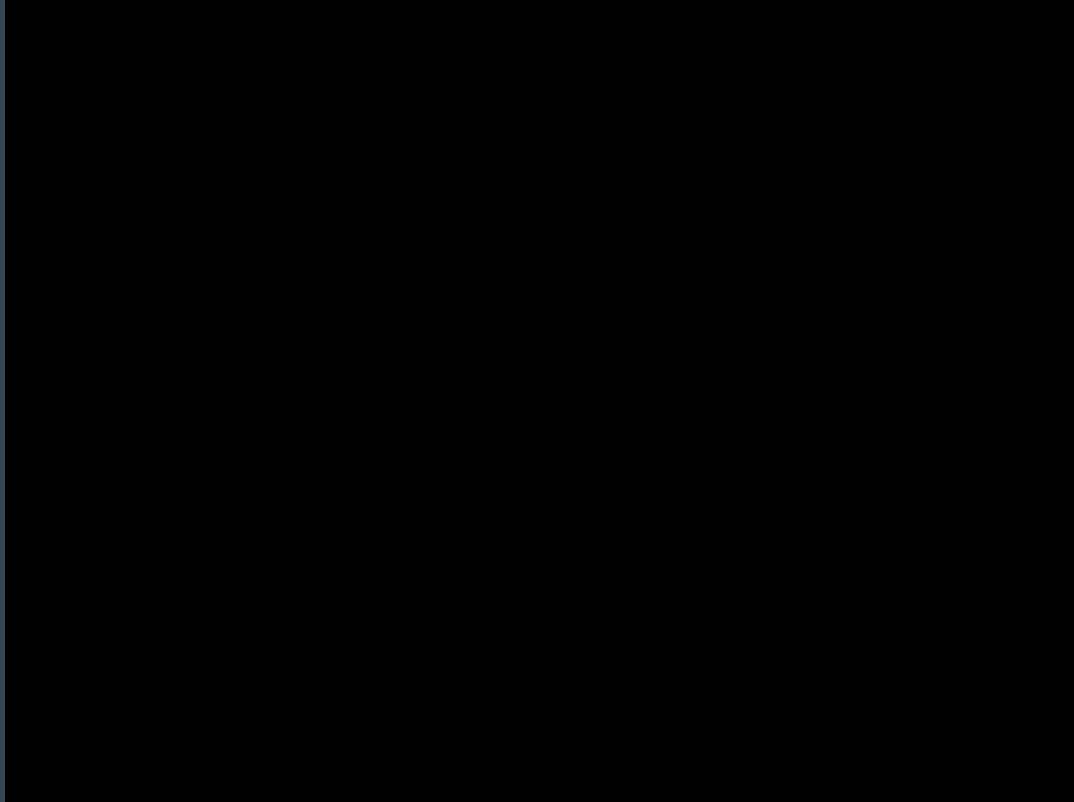
Mild Traffic:



Heavy Traffic:

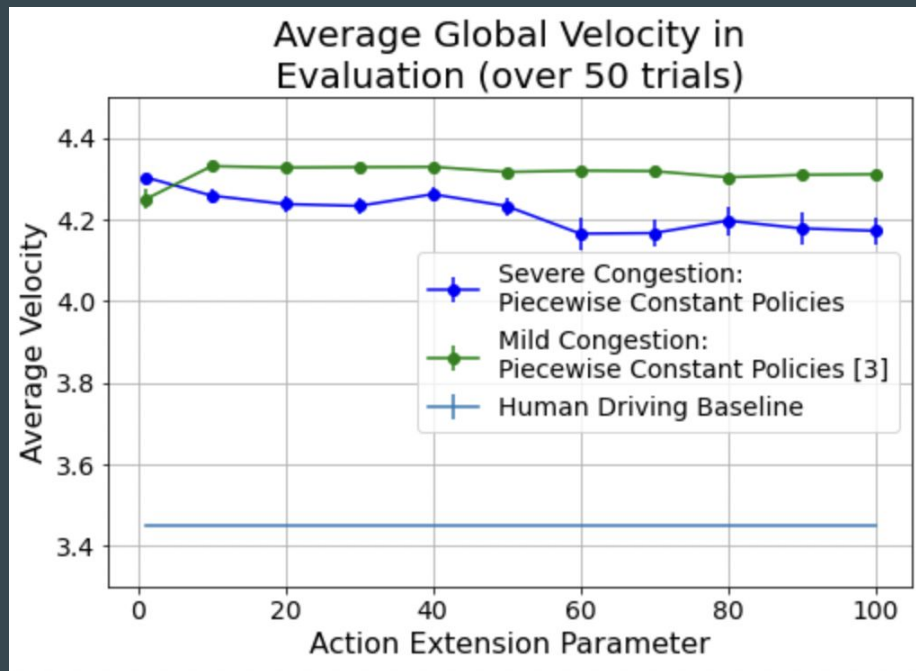


Traffic Mitigation



Results

- Congestion mitigation effects up to $\Delta=100$
- Consistently beat human baseline velocity
- Lower velocity than Sridhar et al. due to traffic severity [3]



Contributions and next steps

- Our work suggests that piecewise constant policies work well in difficult settings, indicating potential for deployment
- We propose pursuing user studies in the future to see how well users can follow policies in practice to measure real-world congestion impacts

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