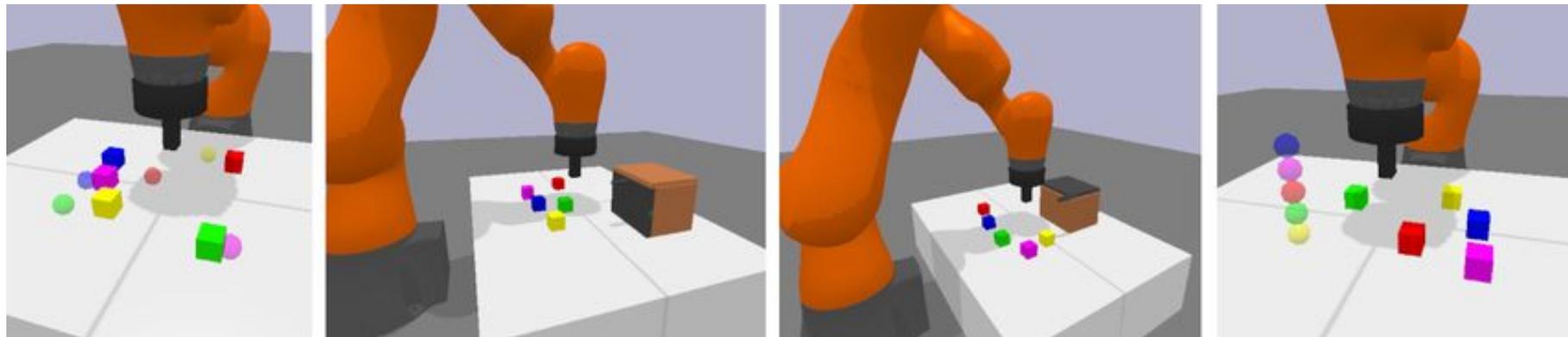


CSC2626 Final Project

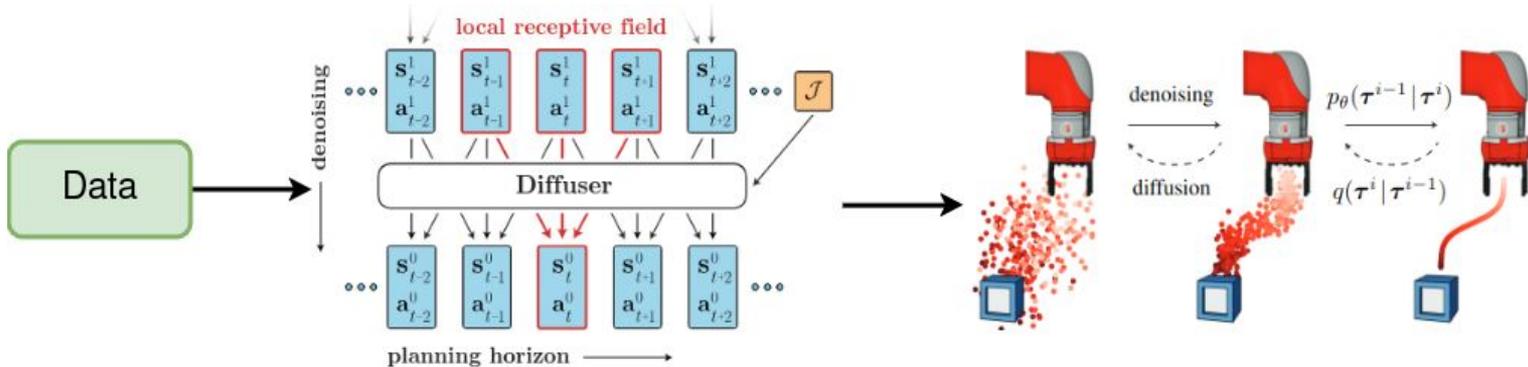
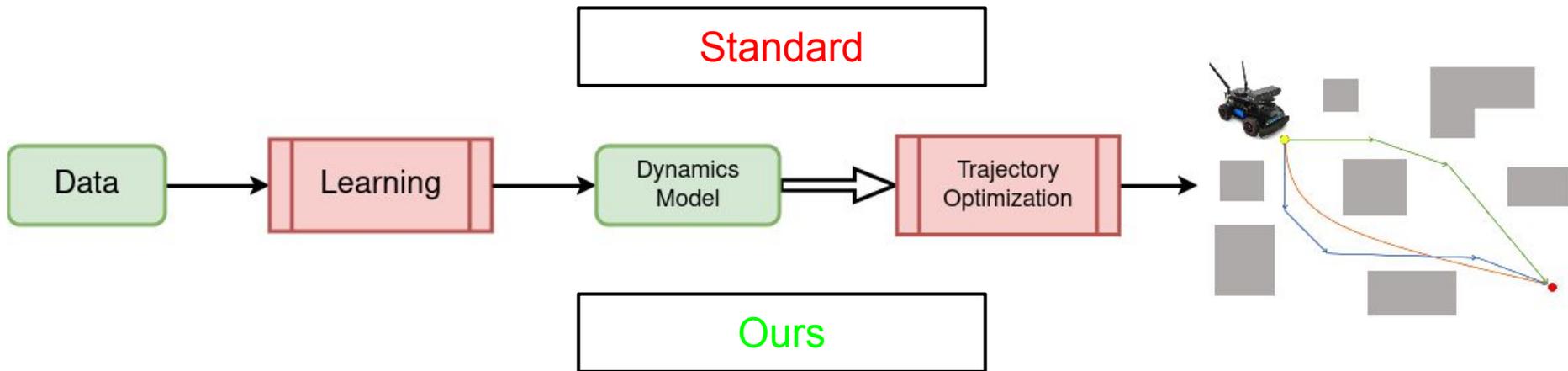
Planning with Diffusion Models



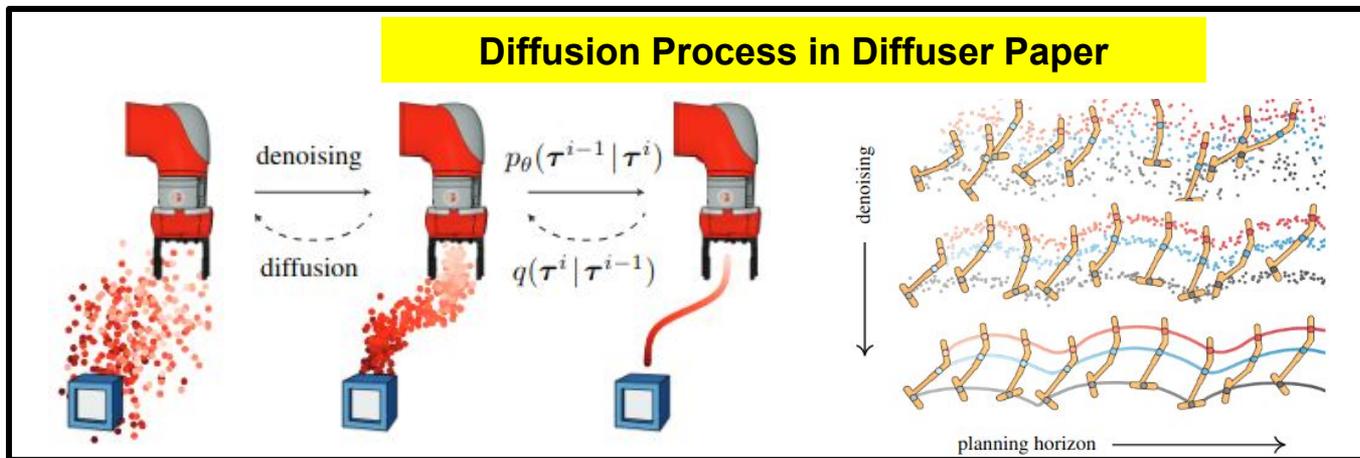
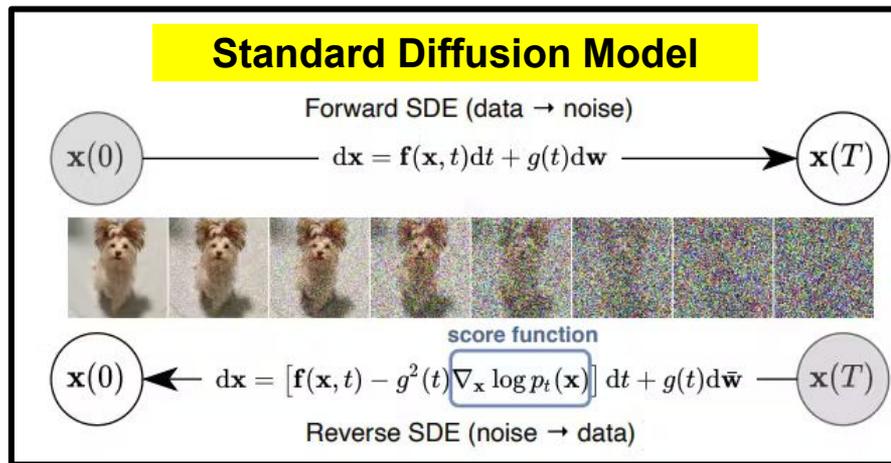
Planning with Diffusion for Flexible Behavior Synthesis

Michael Janner^{*1} Yilun Du^{*2} Joshua B. Tenenbaum² Sergey Levine¹

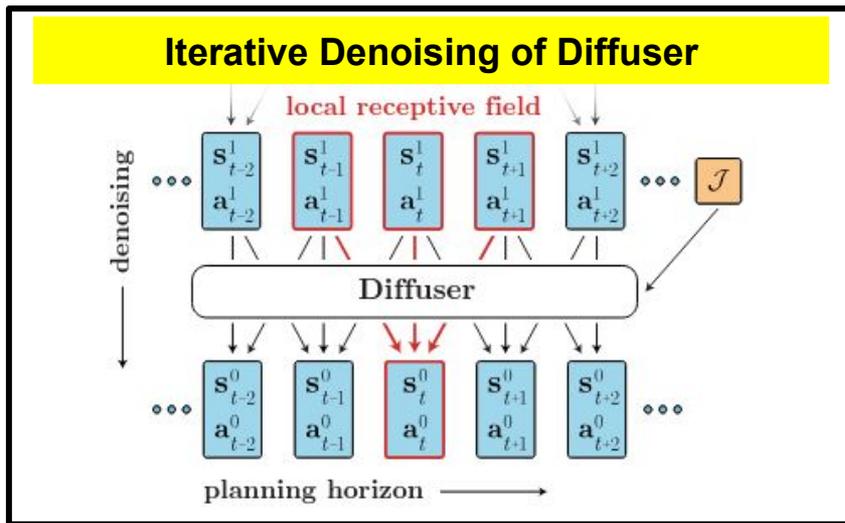
Motivation and Definition of the Problem



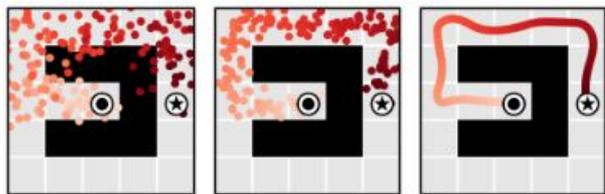
Putting Prior Work into Context - Diffusion Models



Putting Prior Work into Context - Diffuser for trajectory planning

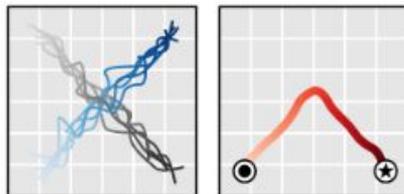


Long-horizon planning



denoising \longrightarrow

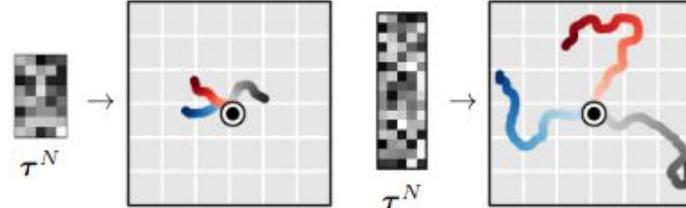
Temporal compositionality



data

plan

Variable-length plans

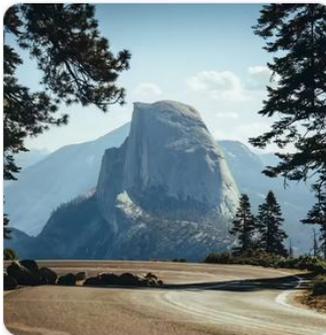
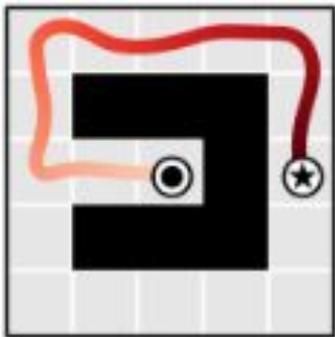


τ^N

τ^N

Scoping - Goal Conditioned RL

Goal Conditioned RL is similar to Image Inpainting

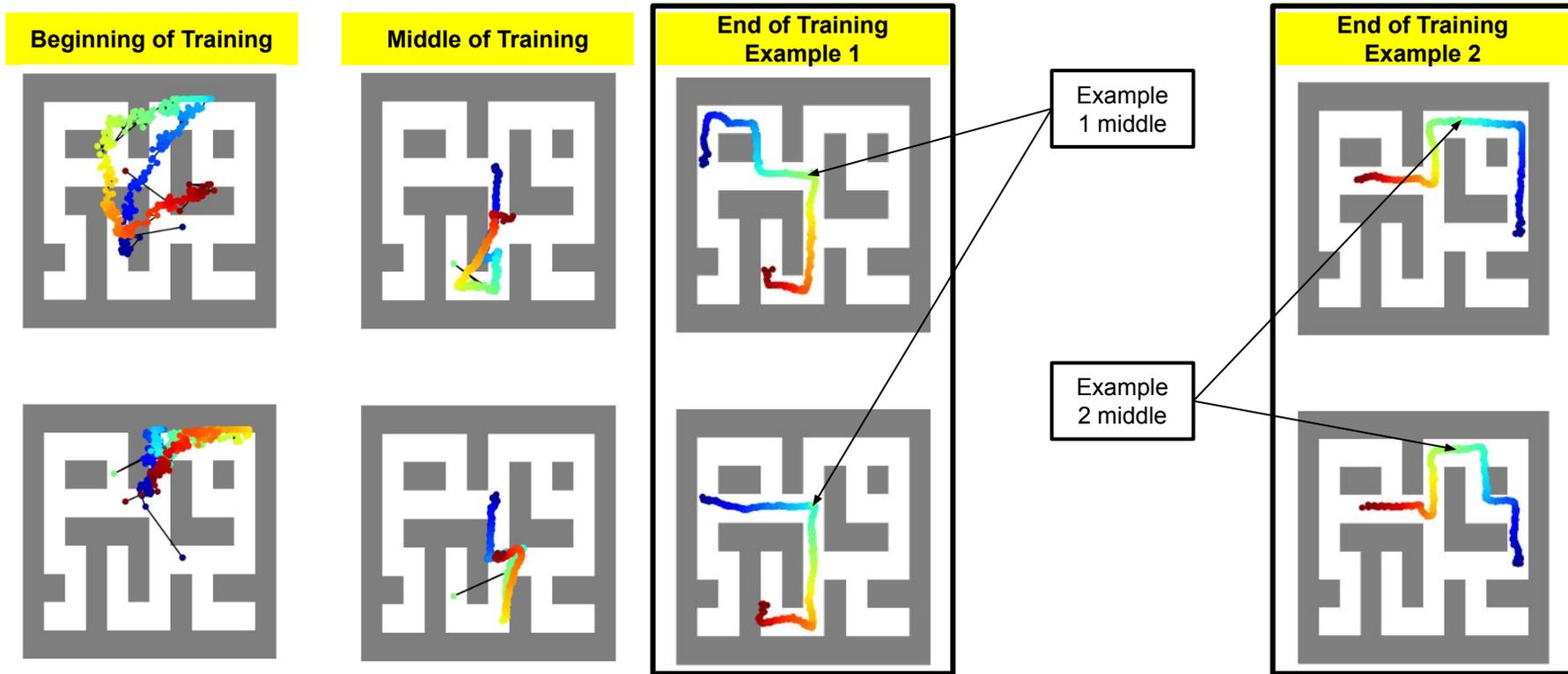


Current Method of Conditioning simply replaces predicted with correct state

Algorithm 1 Guided Diffusion Planning

- 1: **Require** Diffuser μ_θ , guide \mathcal{J} , scale α , covariances Σ^i
 - 2: **while** not done **do**
 - 3: Observe state \mathbf{s} ; initialize plan $\tau^N \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$
 - 4: **for** $i = N, \dots, 1$ **do**
 - 5: // parameters of reverse transition
 - 6: $\mu \leftarrow \mu_\theta(\tau^i)$
 - 7: // guide using gradients of return
 - 8: $\tau^{i-1} \sim \mathcal{N}(\mu + \alpha \Sigma \nabla \mathcal{J}(\mu), \Sigma^i)$
 - 9: // constrain first state of plan
 - 10: $\tau_{\mathbf{s}_0}^{i-1} \leftarrow \mathbf{s}$
 - 11: Execute first action of plan $\tau_{\mathbf{a}_0}^0$
-

Experiment: Train Model to also condition on middle of trajectory



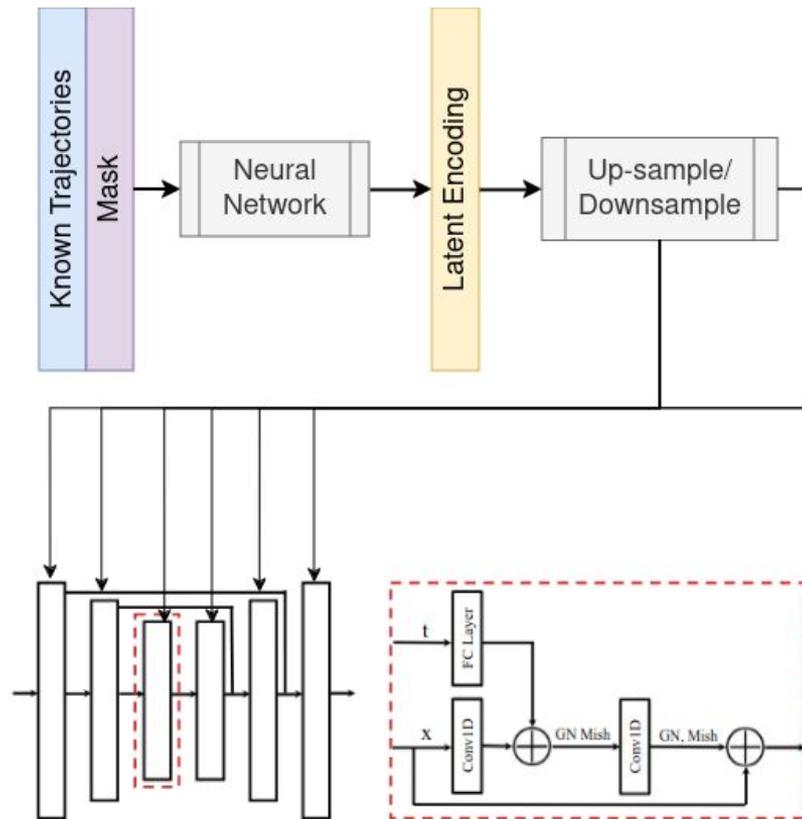
Method: Changing conditioning mechanism

Before

Algorithm 1 Guided Diffusion Planning

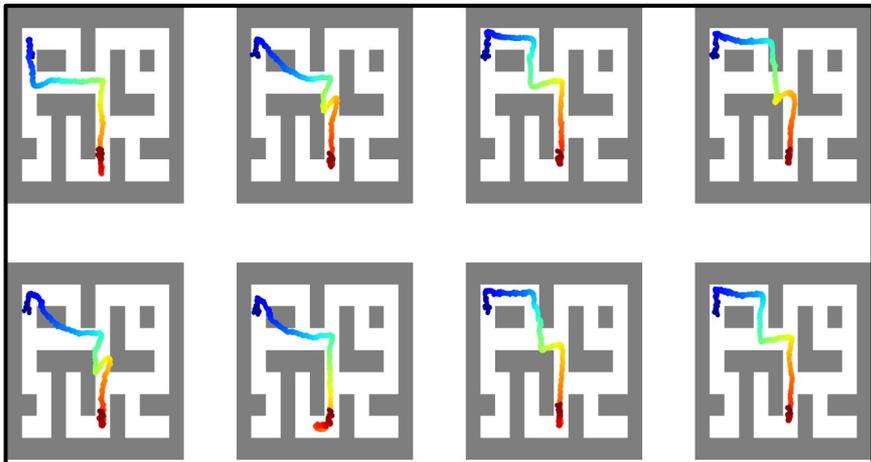
```
1: Require Diffuser  $\mu_\theta$ , guide  $\mathcal{J}$ , scale  $\alpha$ , covariances  $\Sigma^i$ 
2: while not done do
3:   Observe state  $s$ ; initialize plan  $\tau^N \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$ 
4:   for  $i = N, \dots, 1$  do
5:     // parameters of reverse transition
6:      $\mu \leftarrow \mu_\theta(\tau^i)$ 
7:     // guide using gradients of return
8:      $\tau^{i-1} \sim \mathcal{N}(\mu + \alpha \Sigma \nabla \mathcal{J}(\mu), \Sigma^i)$ 
9:     // constrain first state of plan
10:     $\tau_{s_0}^{i-1} \leftarrow s$ 
11:   Execute first action of plan  $\tau_{a_0}^0$ 
```

After

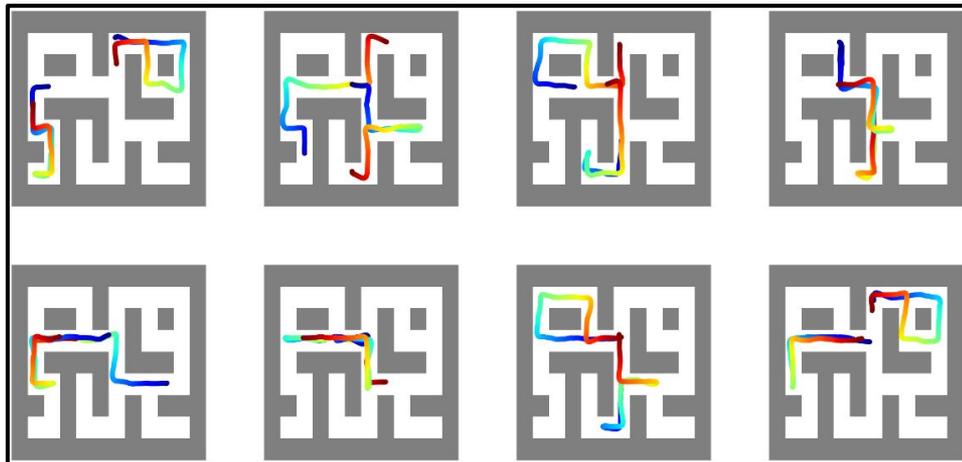


Experiments

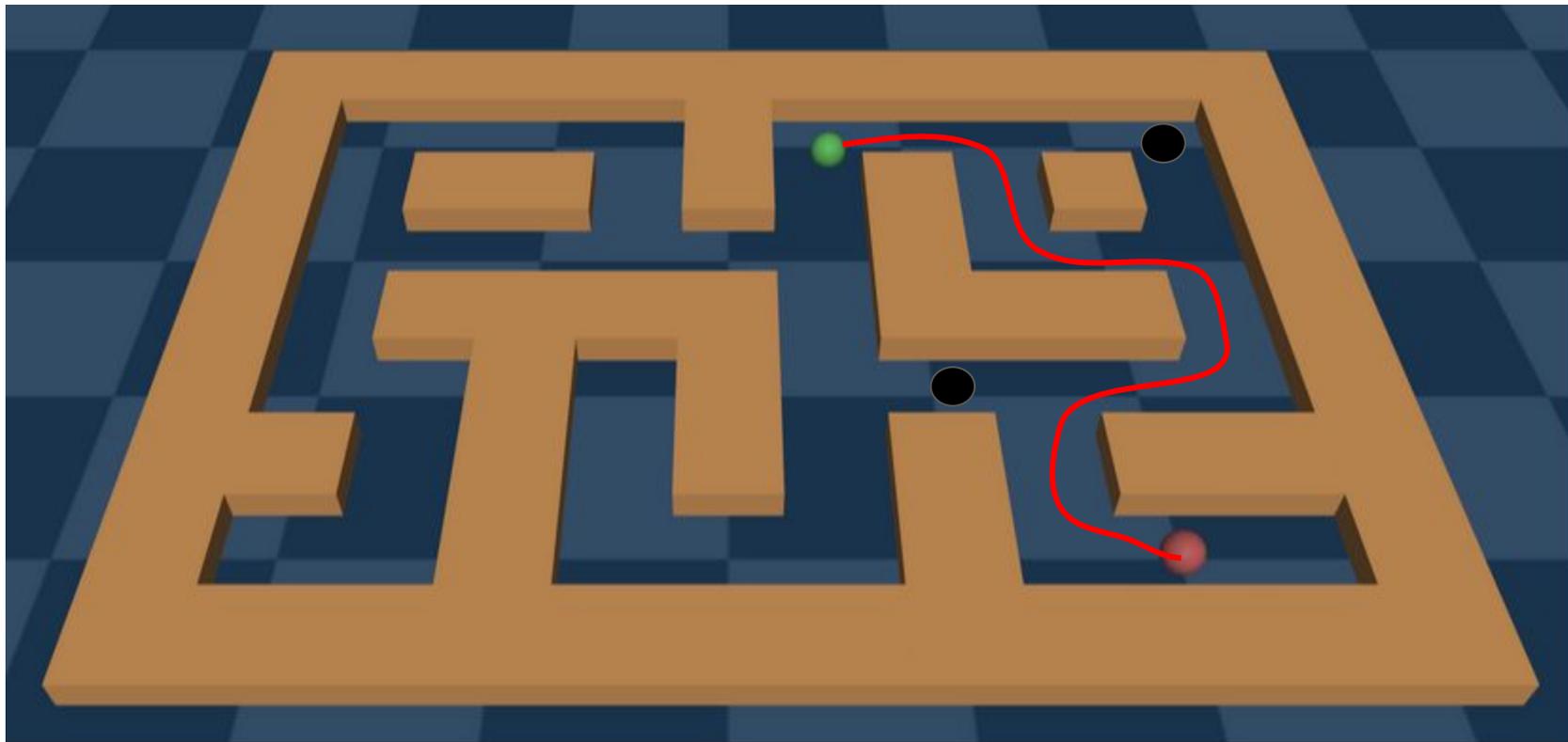
Single Agent with new Conditioning Strategy



Multi Agent with new Conditioning Strategy

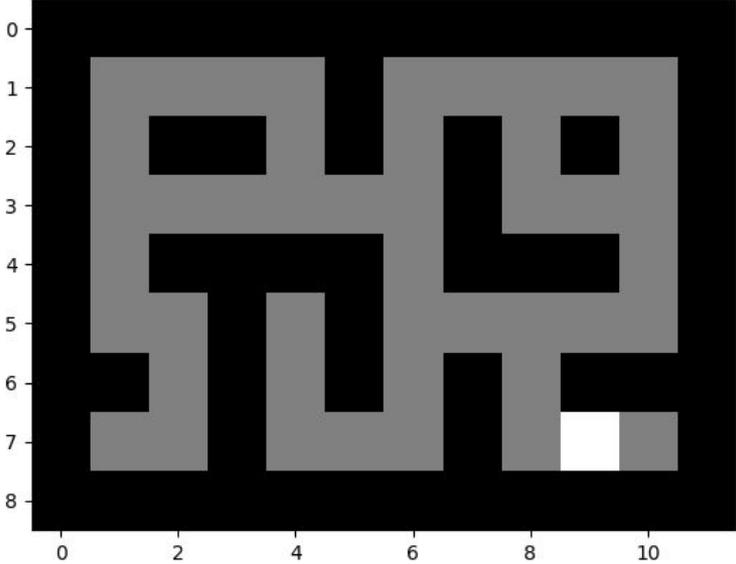


In Progress - Obstacle Avoidance!



Limitations

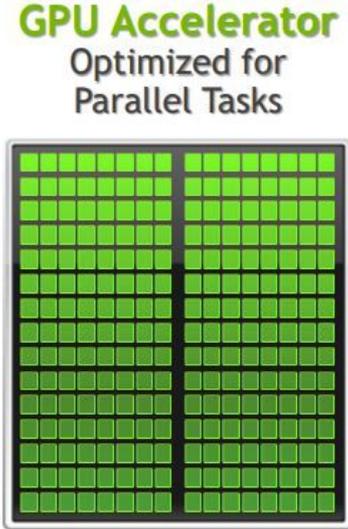
Maze2D-Large Layout



Static maze - will not generalize



+



GPU Compute Limited (NVIDIA T1200 Laptop GPU)