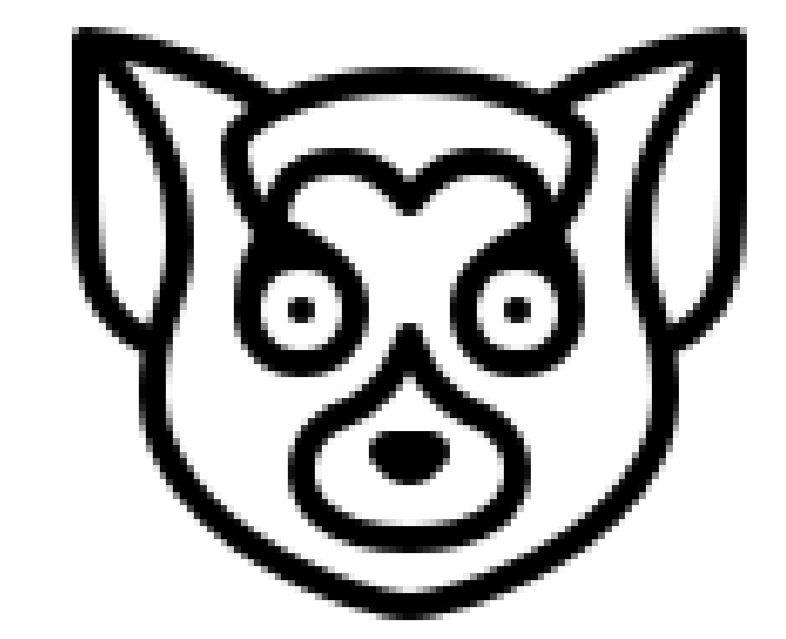


LEMURS: LEarning distributed MUlti-Robot interactions



SCAN ME

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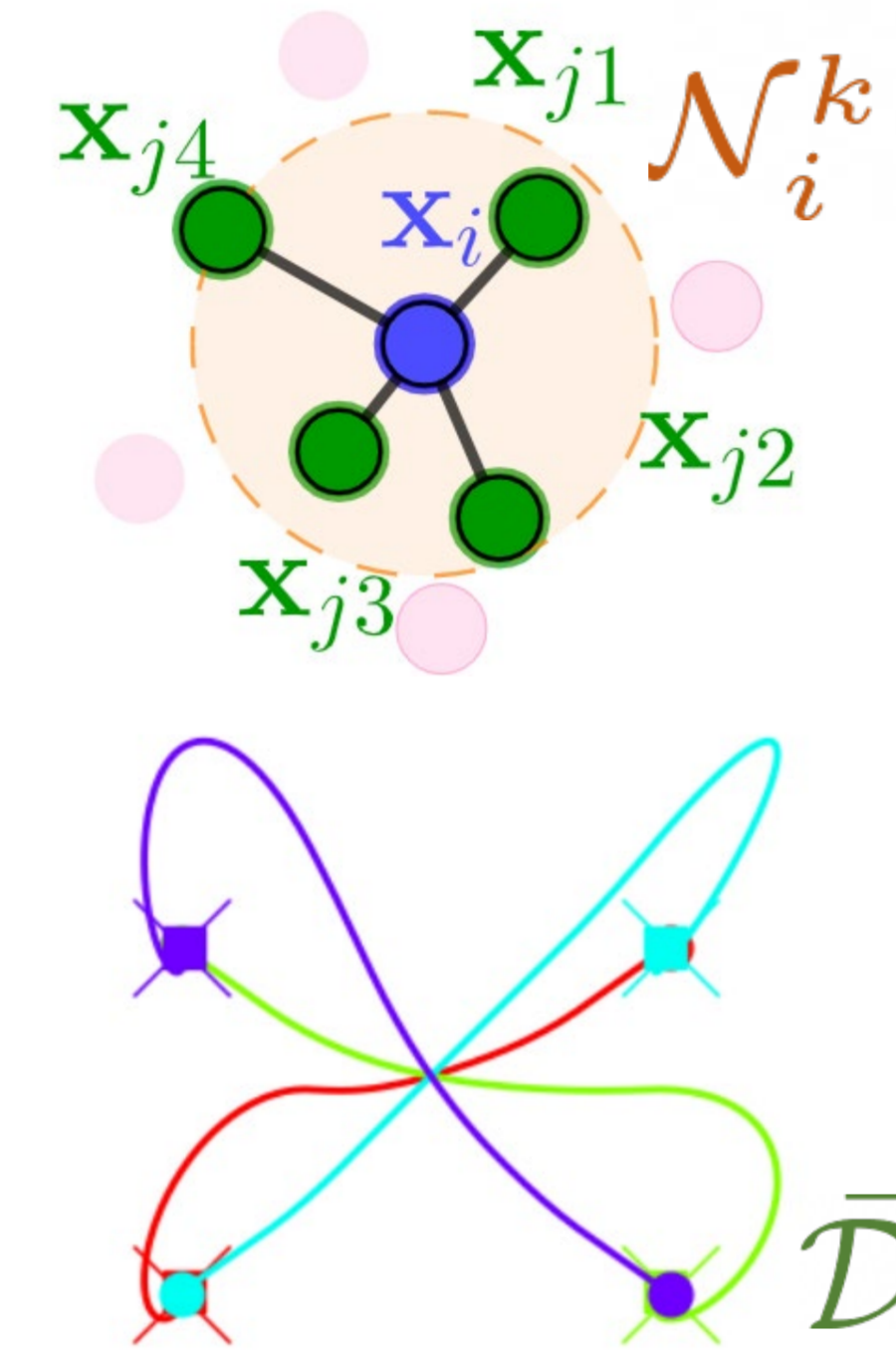
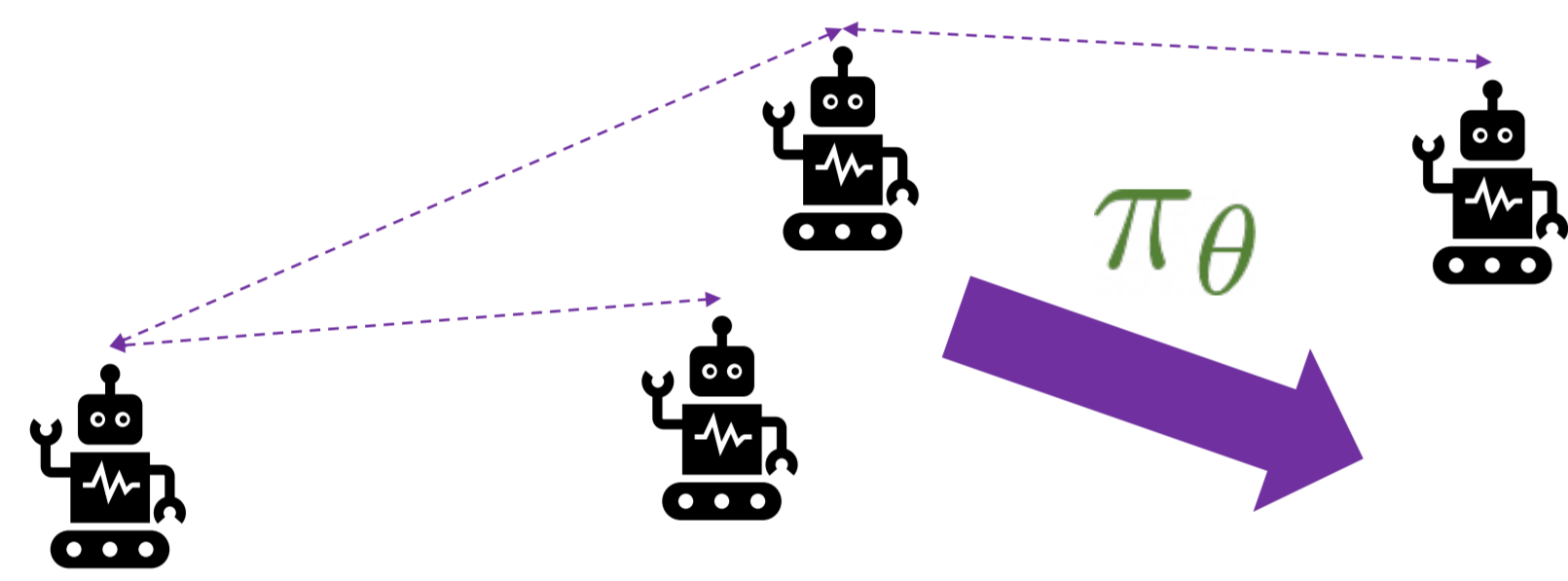
LEMURS learns **scalable, distributed and stable multi-robot control policies** that replicate **cooperative tasks from demonstrations**

I. Problem Statement

from task demonstrations



to multi-robot policies



known robot dynamics

$$\dot{\mathbf{x}}_i(t) = \mathbf{f}_i(\mathbf{x}_i(t), \mathbf{u}_i(t))$$

distributed control policy

$$\mathbf{u}_i(t) = \pi_\theta(\mathbf{x}_{\mathcal{N}_i^k}(t))$$

loss to optimize

$$\min_\theta \mathcal{L}(\mathcal{D}, \bar{\mathcal{D}})$$

state and **input** of robot i

neighbors of i
control policy

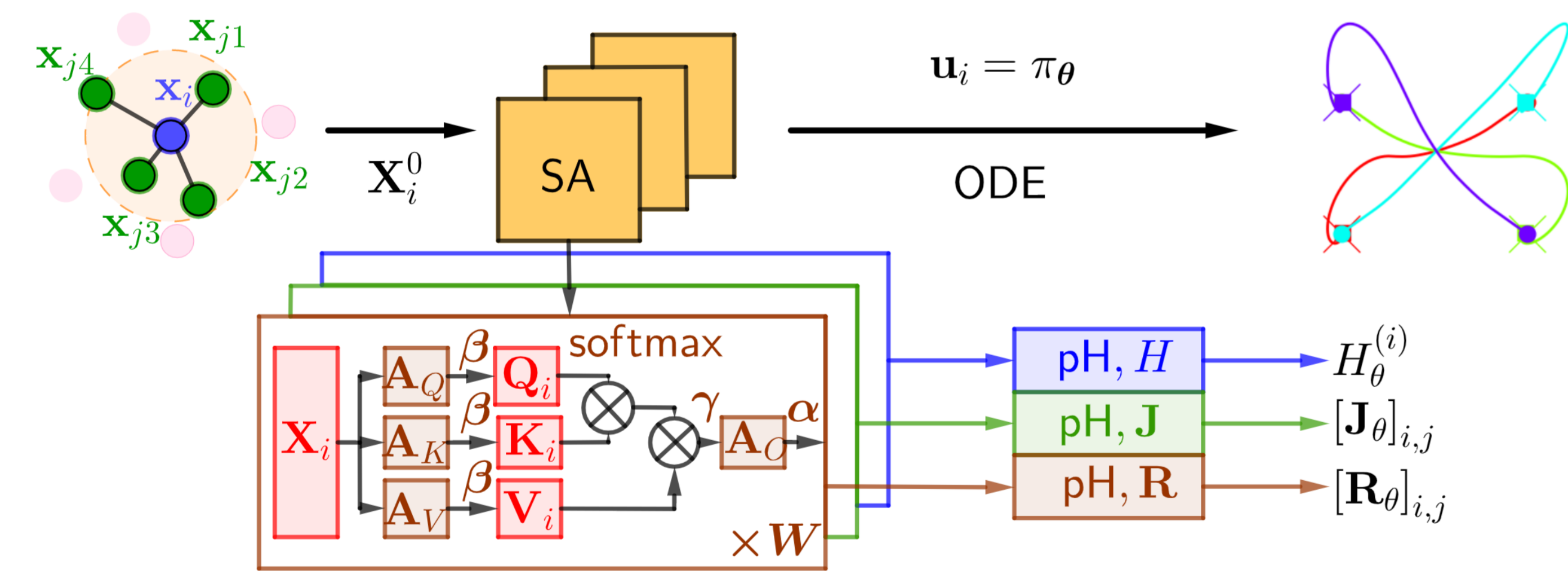
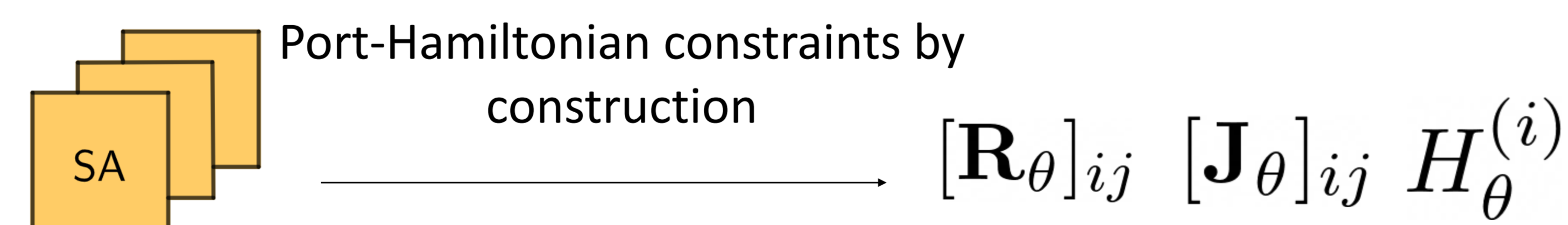
learned and **demonstrated** trajectories

II. LEMURS

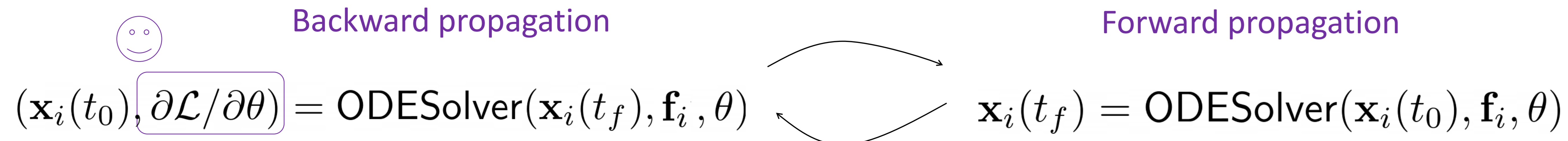
1. **Port-Hamiltonian modelling:** stable and distributed

$$\dot{\mathbf{x}} = (\mathbf{J}_s(\mathbf{x}) - \mathbf{R}_s(\mathbf{x})) \frac{\partial H_s(\mathbf{x})}{\partial \mathbf{x}} + \mathbf{F}_s(\mathbf{x})\mathbf{u}$$

2. **Self-attention:** time-varying team sizes

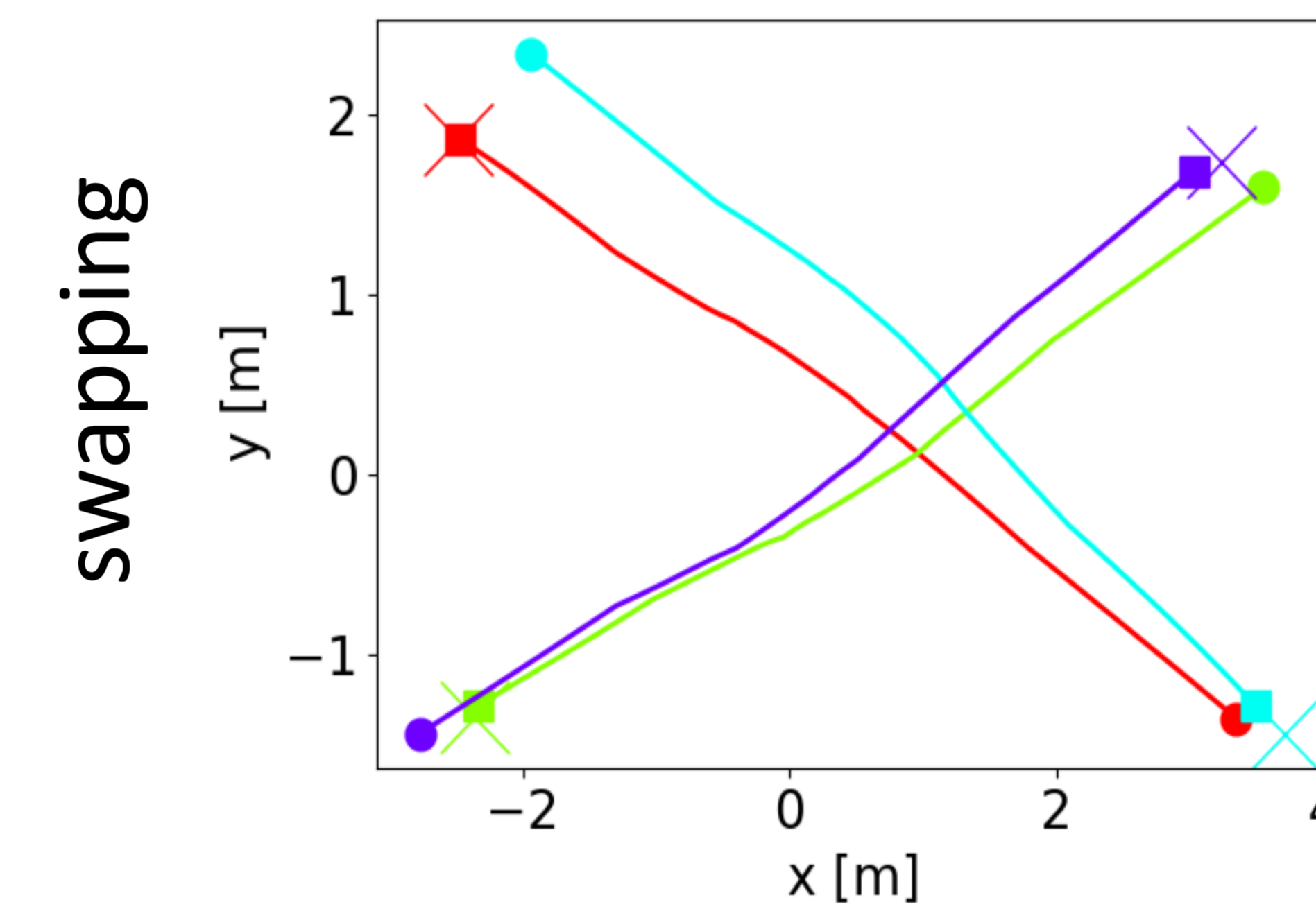


3. **Neural Ordinary Differential Equations:** continuous-time dynamics. Backpropagation using the same ODE solver

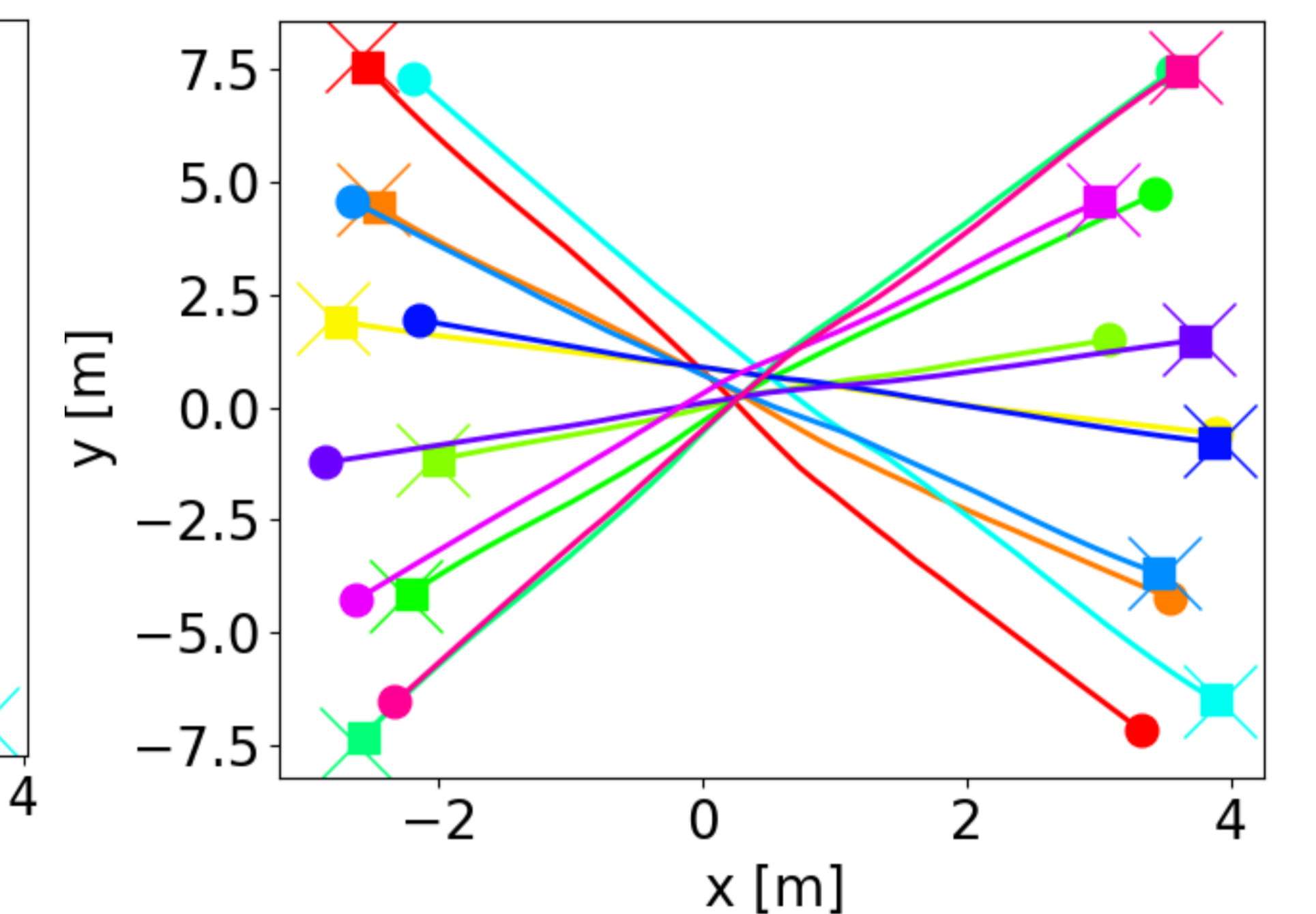


III. Results

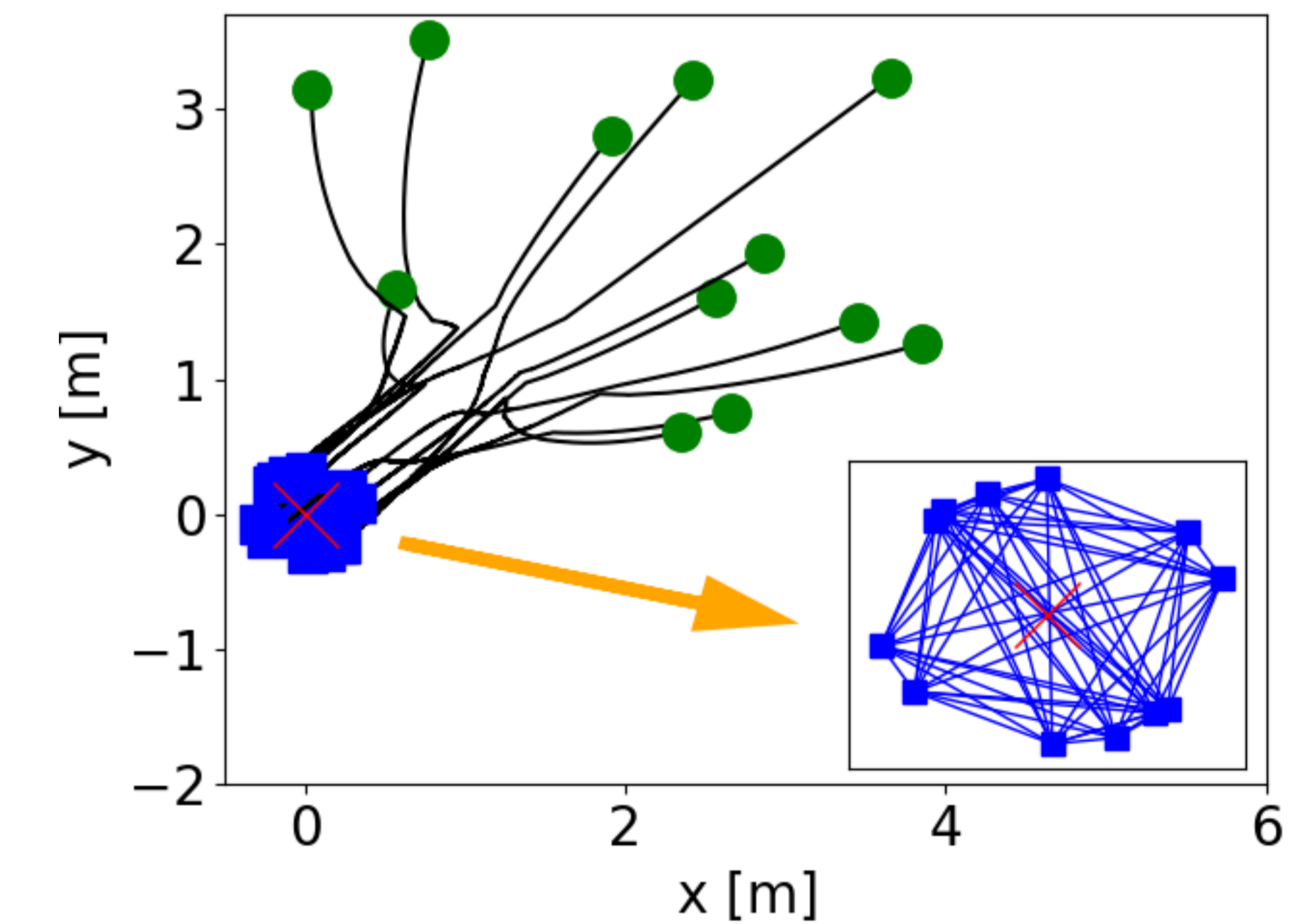
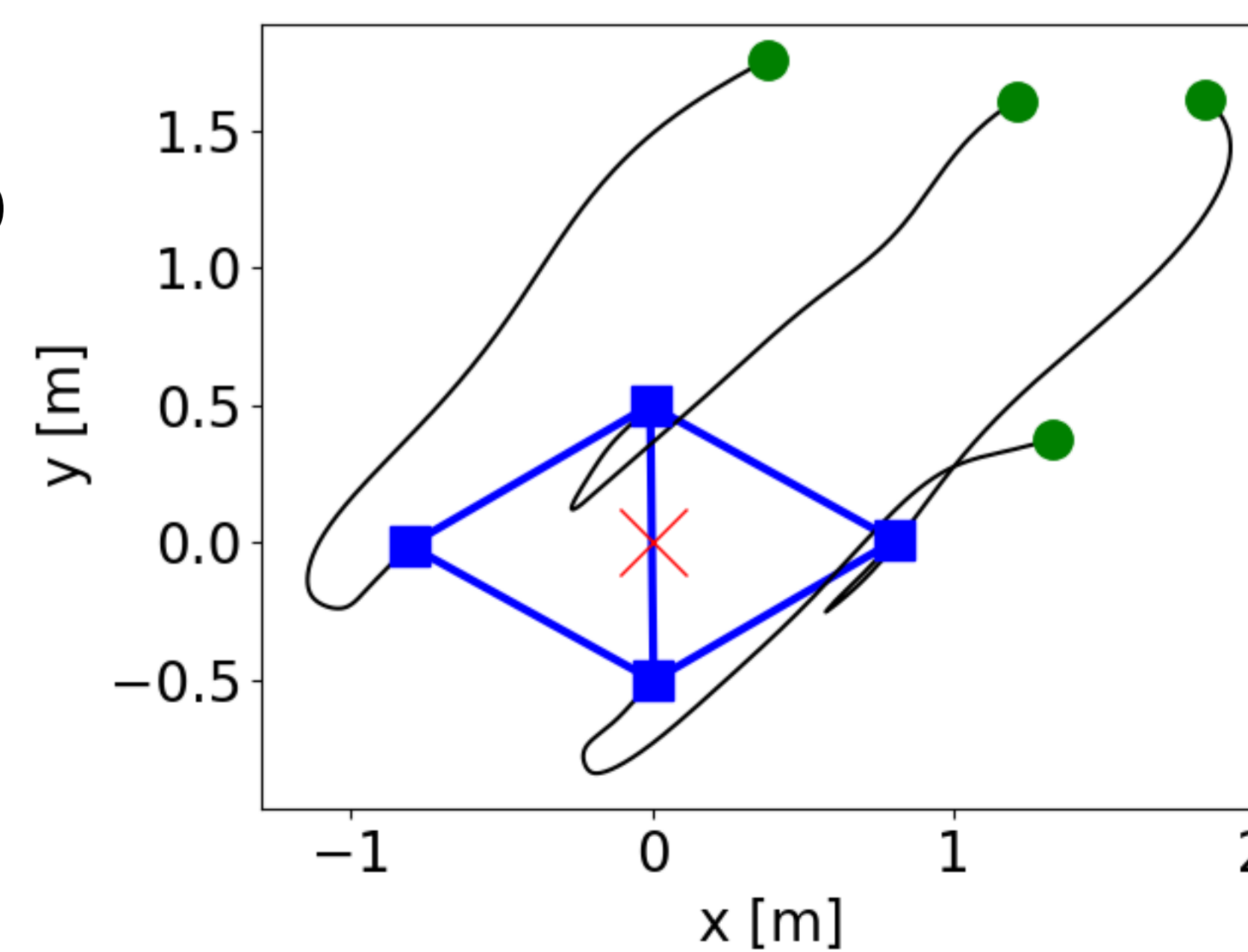
demonstration
only 4 agents



learned policy
scales to many robots!



flocking



IV. Takeaways

LEMURS learns robot interactions from demonstrations using self-attention and Hamiltonian-based neural ODE networks.

LEMURS advances the state of the art by learning control policies that generalize to increasing numbers of robots and time-varying communications, from just state-only trajectories of few robots.