

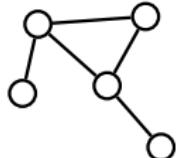
Bryan Van Scoy

✉ bvanscoy@miamioh.edu

🌐 vanscoy.github.io

Miami University

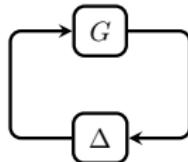
Multi-agent systems



Optimization

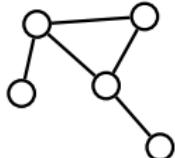


Robust control



All of my teaching and research involves dynamical systems.

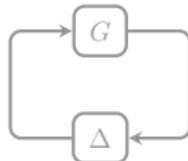
Multi-agent systems



Optimization



Robust control



A system composed of multiple interacting agents.



drone swarms



vehicle platoons



smart grid



wind farm



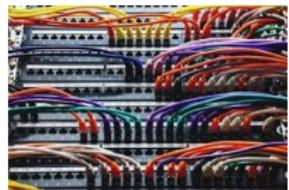
smart factory



server farm



sports



routing

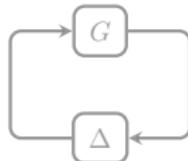
Multi-agent systems



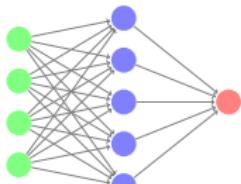
Optimization



Robust control



Making the best or most effective use of a situation or resource.



machine learning



navigation



resource allocation



portfolio optimization

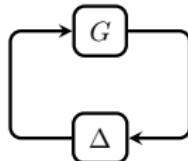
Multi-agent systems



Optimization



Robust control



An approach to controller design that explicitly deals with uncertainty.



drones



robotics

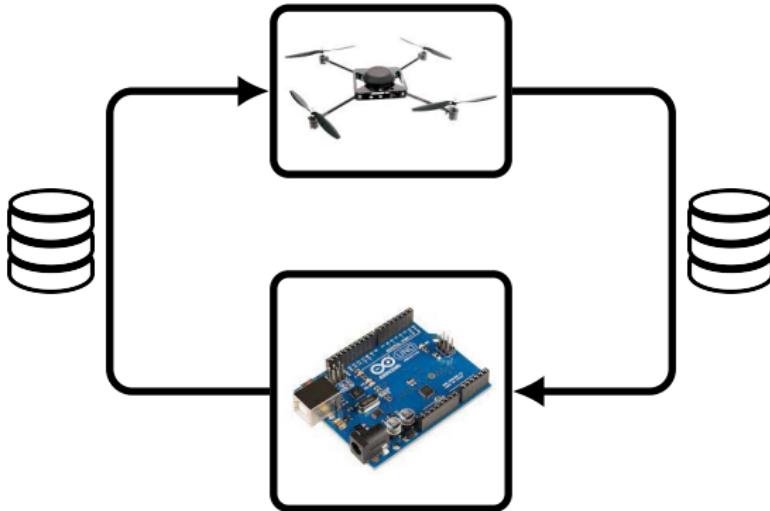


aircraft



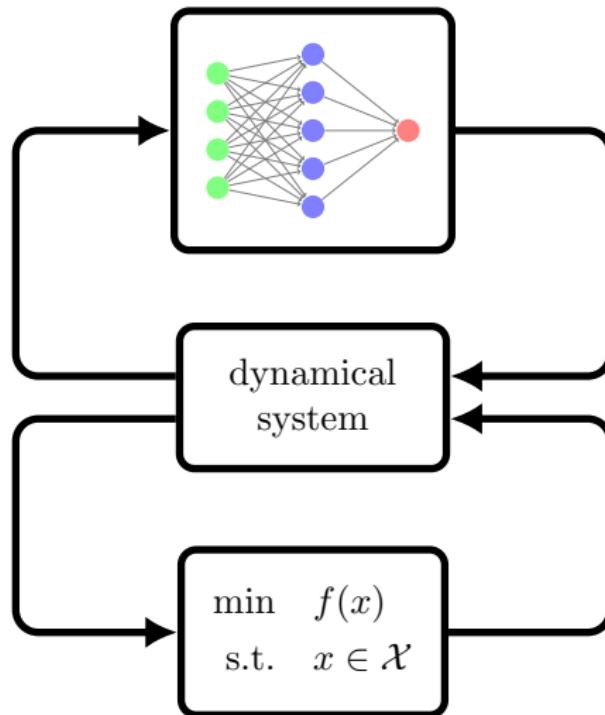
factories

Data-driven control



- many systems are too complex for a simple model
- use data to directly control the system
- connections with *optimization* and *reinforcement learning*

Complex interconnected systems



Multi-agent robotics lab

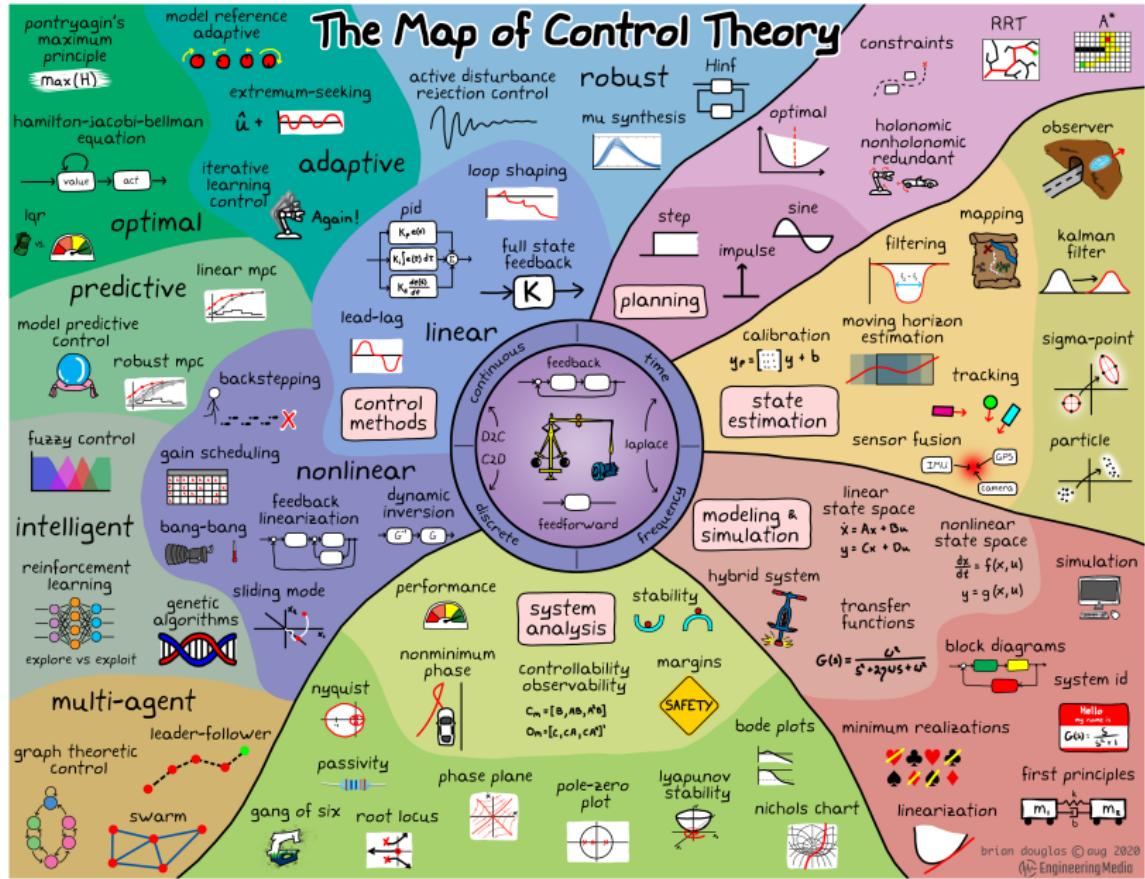


Kilobot by Mike Rubenstein

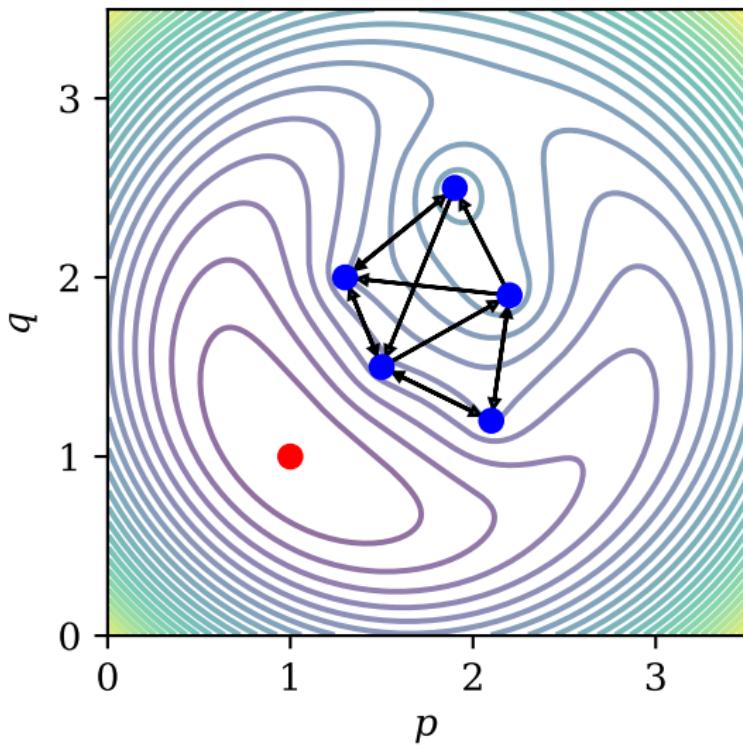


Olympic Drone Show by Intel

- provide hands-on experience with various robotic systems
- utilize for capstone design projects
- testbed for algorithm development



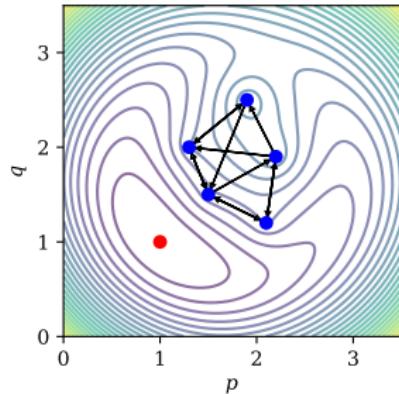
Example: Target localization



Example: Target localization

- target located at position $(p_*, q_*) \in \mathbb{R}^2$
- agent i knows its position $(p_i, q_i) \in \mathbb{R}^2$ and distance to the target

$$r_i = \sqrt{(p_i - p_*)^2 + (q_i - q_*)^2}$$



- the objective function associated to agent i is

$$f_i(p, q) = \frac{1}{2} \left(\sqrt{(p_i - p)^2 + (q_i - q)^2} - r_i \right)^2$$

- the target position is the solution to the problem

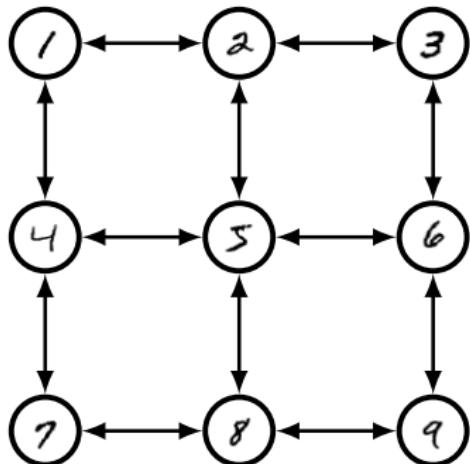
$$\underset{p, q}{\text{minimize}} \quad \sum_{i=1}^n f_i(p, q)$$

Example: Distributed machine learning

Each agent i has

- data (x_i, y_i)
- features $\Phi(x_i)$
- prediction model $\theta^T \Phi(\cdot)$
- error $\|y_i - \theta^T \Phi(x_i)\|$

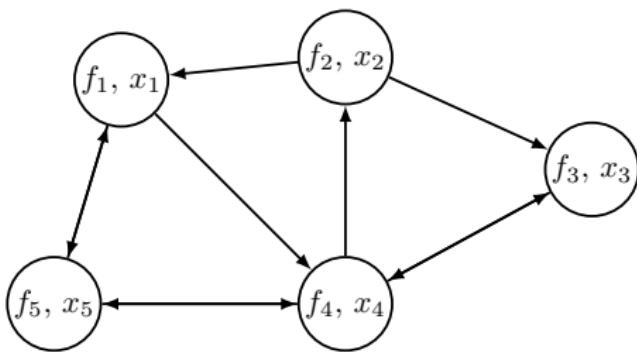
$$\underset{\theta}{\text{minimize}} \quad \sum_{i=1}^n \|y_i - \theta^T \Phi(x_i)\|$$



Agents want a *global* model while keeping their local data *private*.

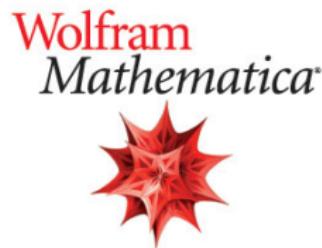
Problem abstraction

$$\begin{aligned} & \underset{x_1, \dots, x_n}{\text{minimize}} && \sum_{i=1}^n f_i(x_i) \\ & \text{subject to} && x_1 = x_2 = \dots = x_n \end{aligned}$$



Want each agent to compute the global optimizer by communicating with local neighbors and performing local computations.

Tools



L^AT_EX

Contact info

- <https://vanscoy.github.io>
- Garland Hall, Room 260 Q