

Department of Mechanical and Aerospace
Engineering
New York University

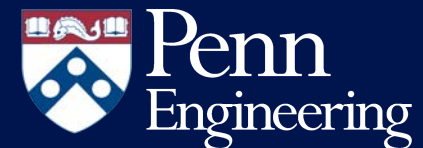
Dec 2, 2019

Swarms of Small Flying Robots

Vijay Kumar

Professor and Nemirovsky Family Dean

University of Pennsylvania



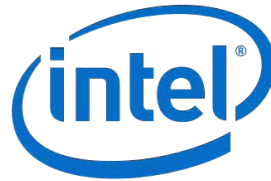
Acknowledgements



Qualcomm



LOCKHEED MARTIN



NVIDIA

Spin off companies

Non profit



Trefo.ai



acquired by
QUALCOMM



*applications to
agriculture*

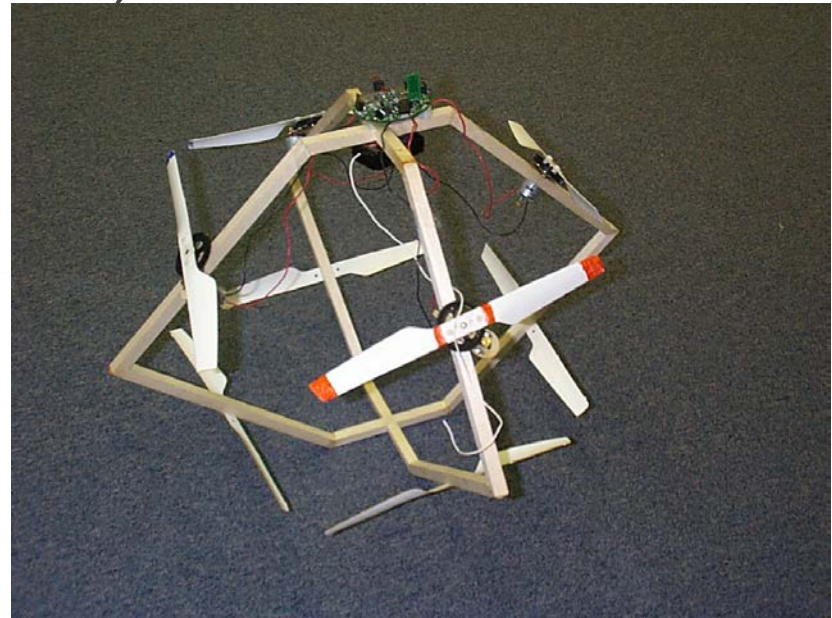
aerial mapping

*mining,
asset mapping*

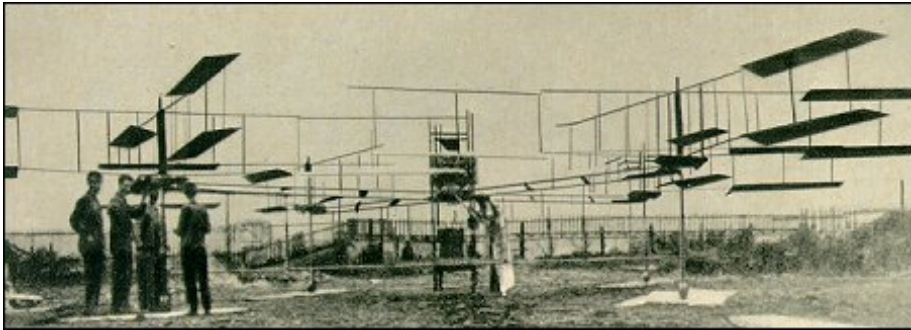
*health,
disaster response*

A Brief History of Aerial Robotics

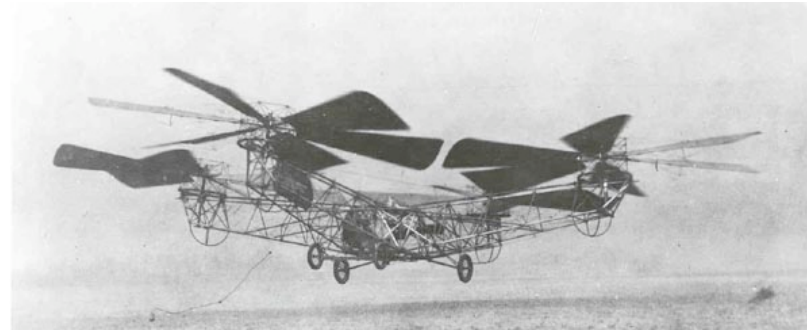
- International Aerial Robotics Competition (1991)
- Early work at GRASP (< 2000)



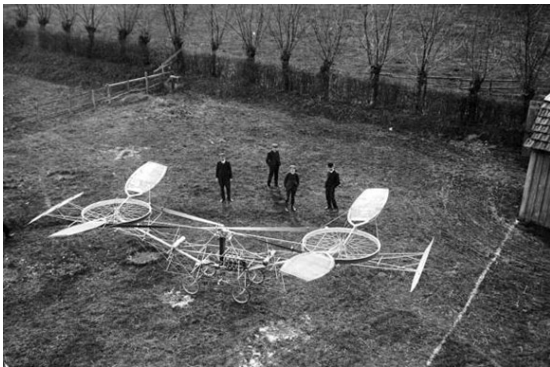
There are no new ideas, only good ideas!



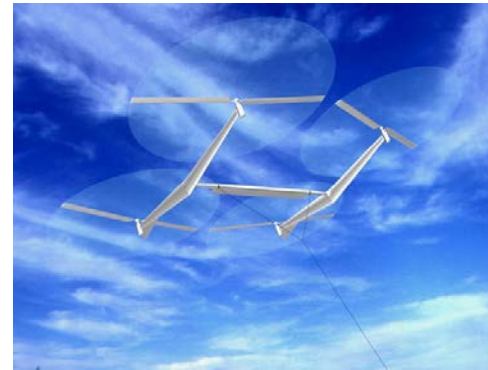
Breguet-Richet Gyroplane No. I, 1907



George de Bothezat's Quadrotor, 1922



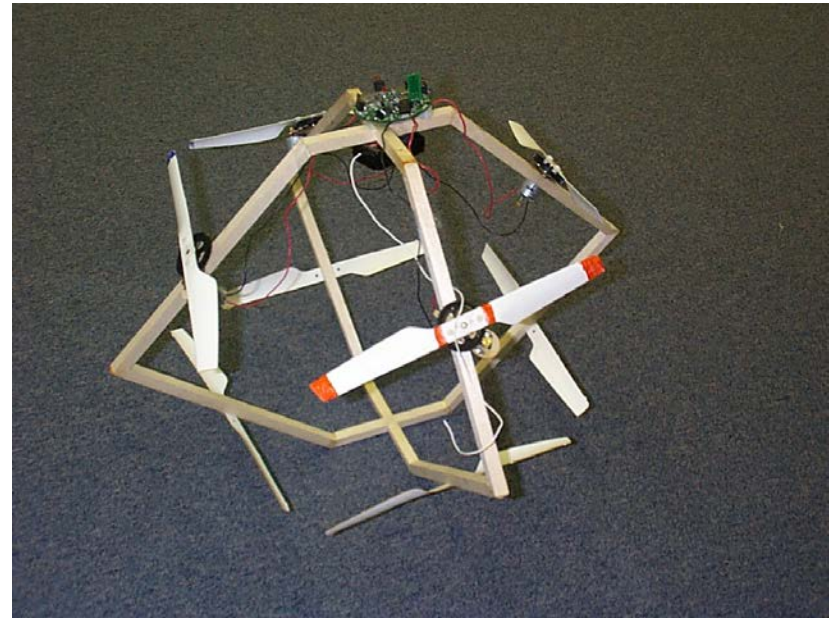
1907, Paul
Cornu: First
to Hover?



Flying
windmill,
2007

A Brief History of Aerial Robotics

- International Aerial Robotics Competition (1991)
- Early work at GRASP (< 2000)



Inertial Measurement Units

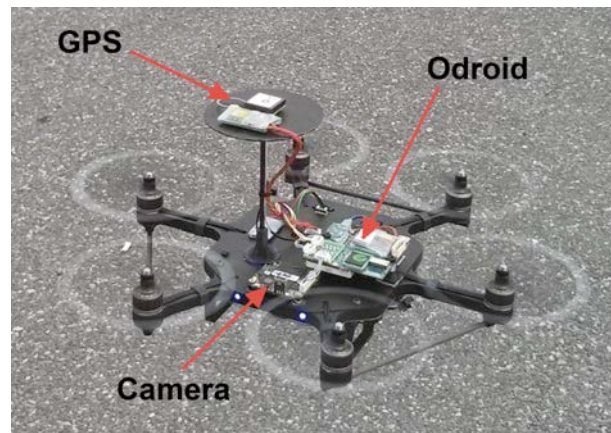
- Accelerometer, airbag sensors (Analog Devices), 1993
- MEMS gyros for electronic stability control (Bosch), 1997
- 3-axis accelerometers for Nintendo Wii, 2006
- 3-axis accelerometer, iPhone (Apple), 2007
- 3-axis accelerometer, 3-axis gyro, 3-axis magnetometer, iPhone 4 (Apple), 2010 *inflexion point*

A Brief History of Aerial Robotics

- International Aerial Robotics Competition (1991)
- Early work at GRASP
- 2008-9 – small multi rotor aircrafts become practical



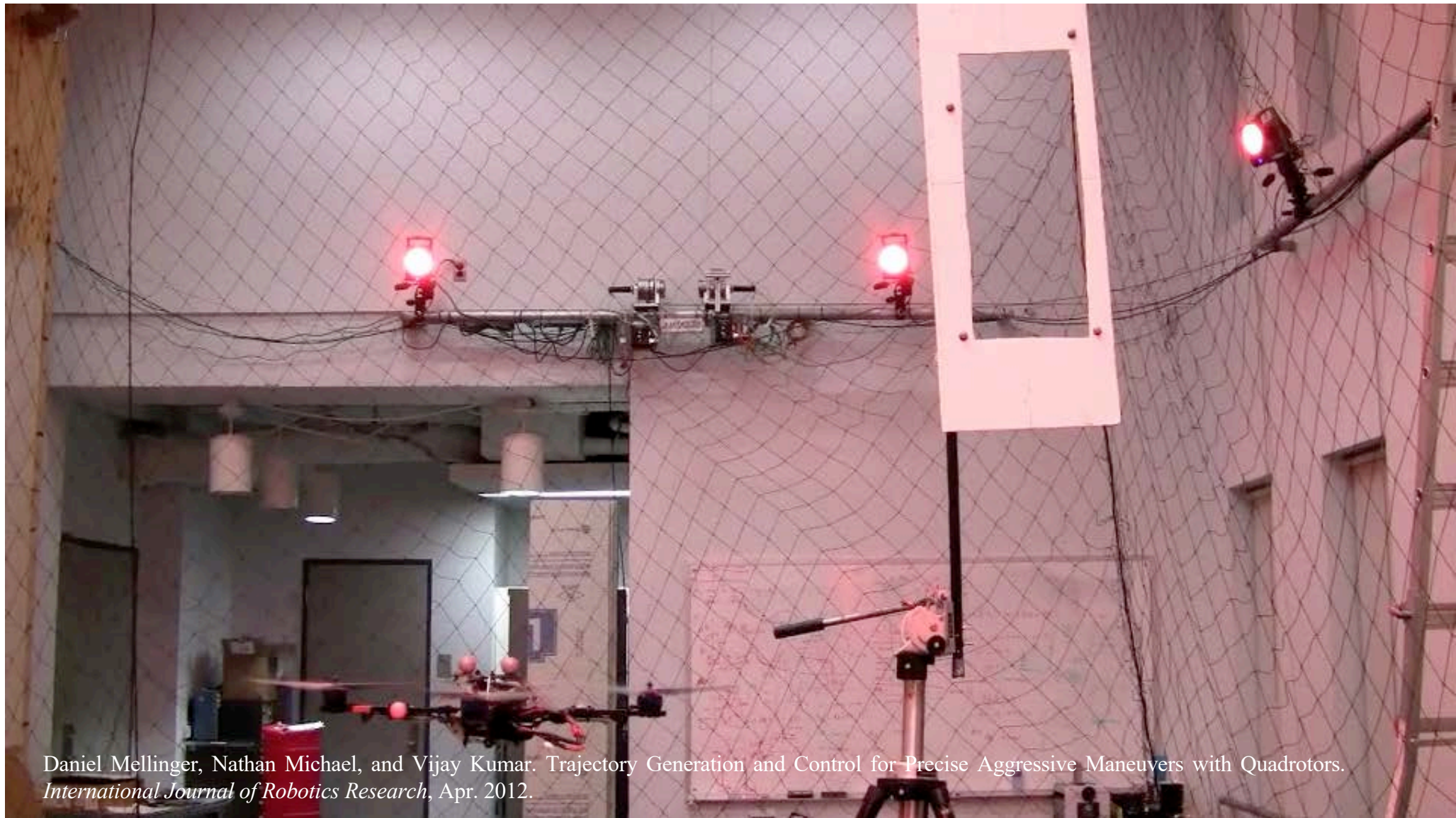
Ascending Technologies



KMel Robotics



Daniel Mellinger and Alex Kushleyev



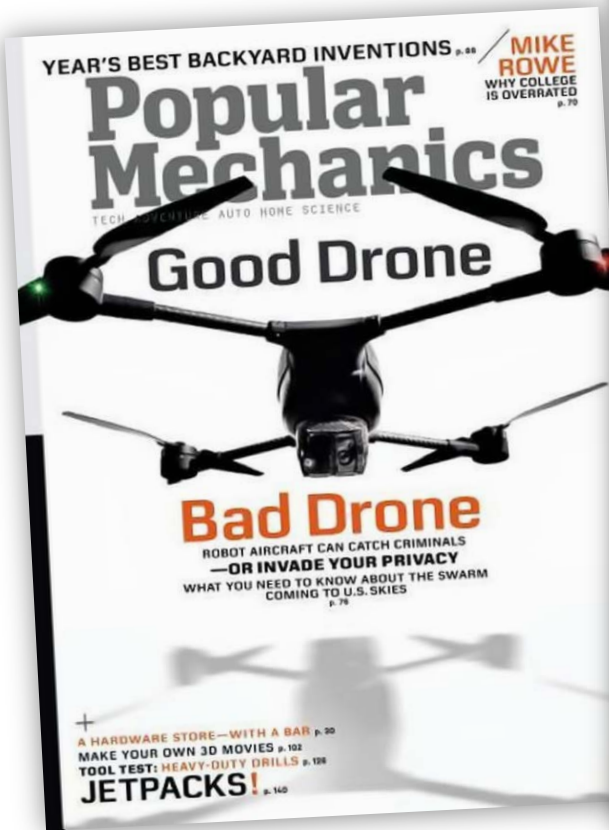
Daniel Mellinger, Nathan Michael, and Vijay Kumar. Trajectory Generation and Control for Precise Aggressive Maneuvers with Quadrotors. *International Journal of Robotics Research*, Apr. 2012.



Daniel Mellinger, and Vijay Kumar. "Minimum Snap Trajectory Generation and Control for Quadrotors," Proc. IEEE International Conference on Robotics and Automation. Shanghai, China, May, 2011.



2015 – Drones everywhere!



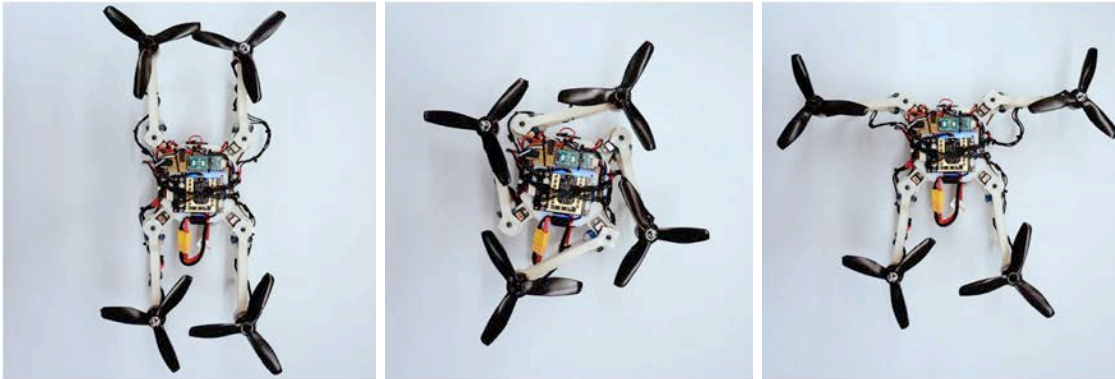
Drone Racing



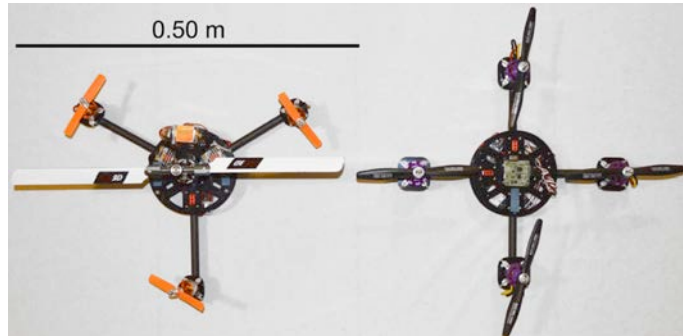
Yash Mulgaonkar



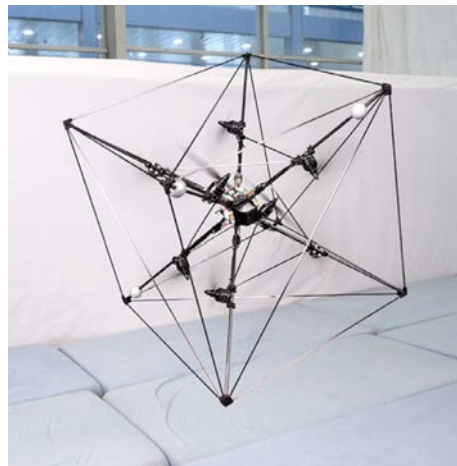
Beyond Quadrotors



Davide Falanga , Kevin Kleber, Stefano Mintchev , Dario Floreano , and Davide Scaramuzza, "The Foldable Drone: A Morphing Quadrotor That Can Squeeze and Fly," 2019.



S. Driessens and P. Pounds, "The triangular quadrotor: a more efficient quadrotor configuration," 2015.

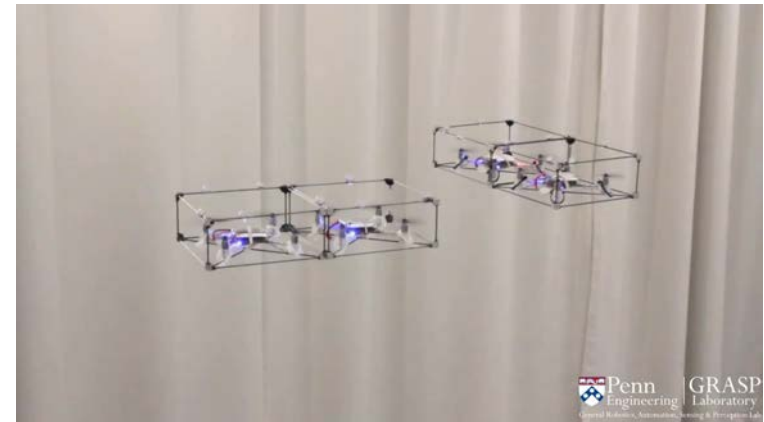


David Saldana, Bruno Gabrich, Guanrui Li, Mark Yim, and Vijay Kumar, "ModQuad: The Flying Modular Structure that Self-Assembles in Midair," 2018.

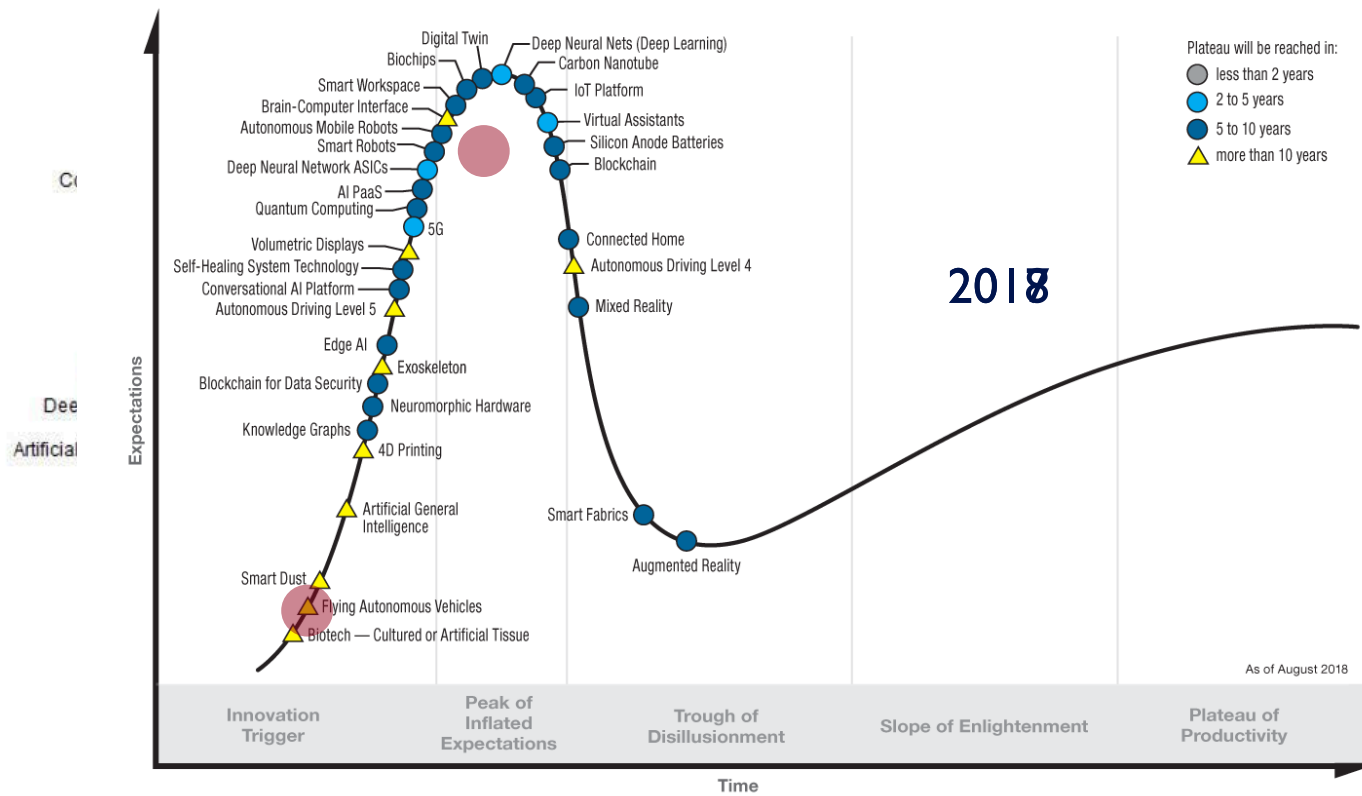


David Saldana

Dario Brescianini and Raffaello D'Andrea, "An omni-directional multirotor vehicle," 2018.

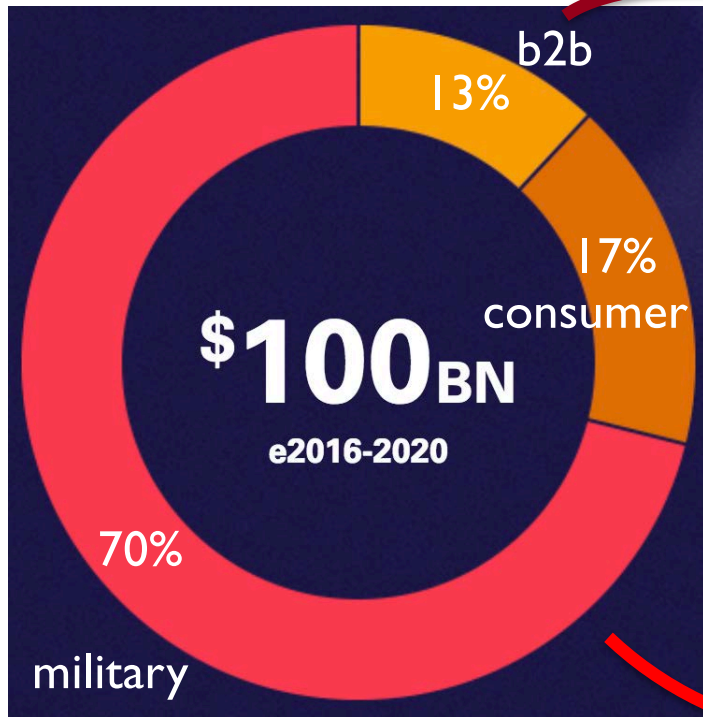


Gartner Hype Cycle



2019 (10 years later)

Market



Goldman Sachs Research

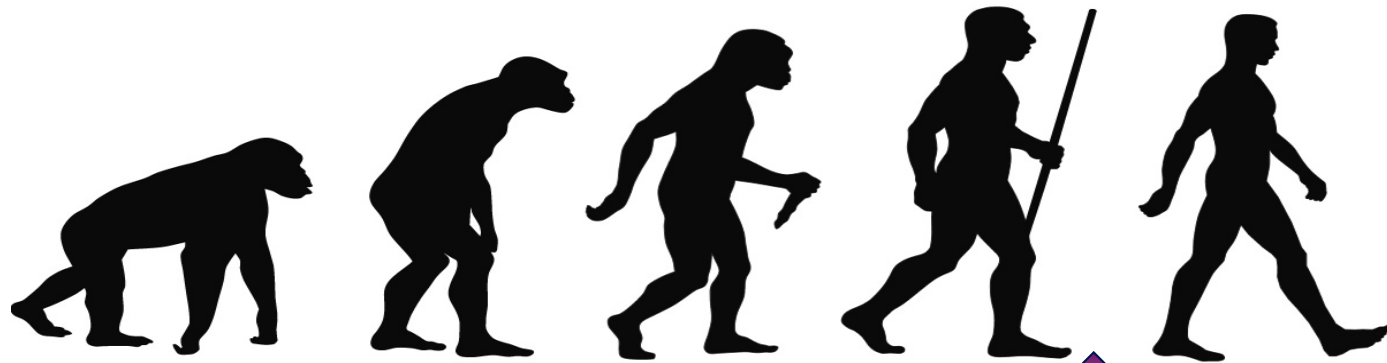
 Penn Engineering



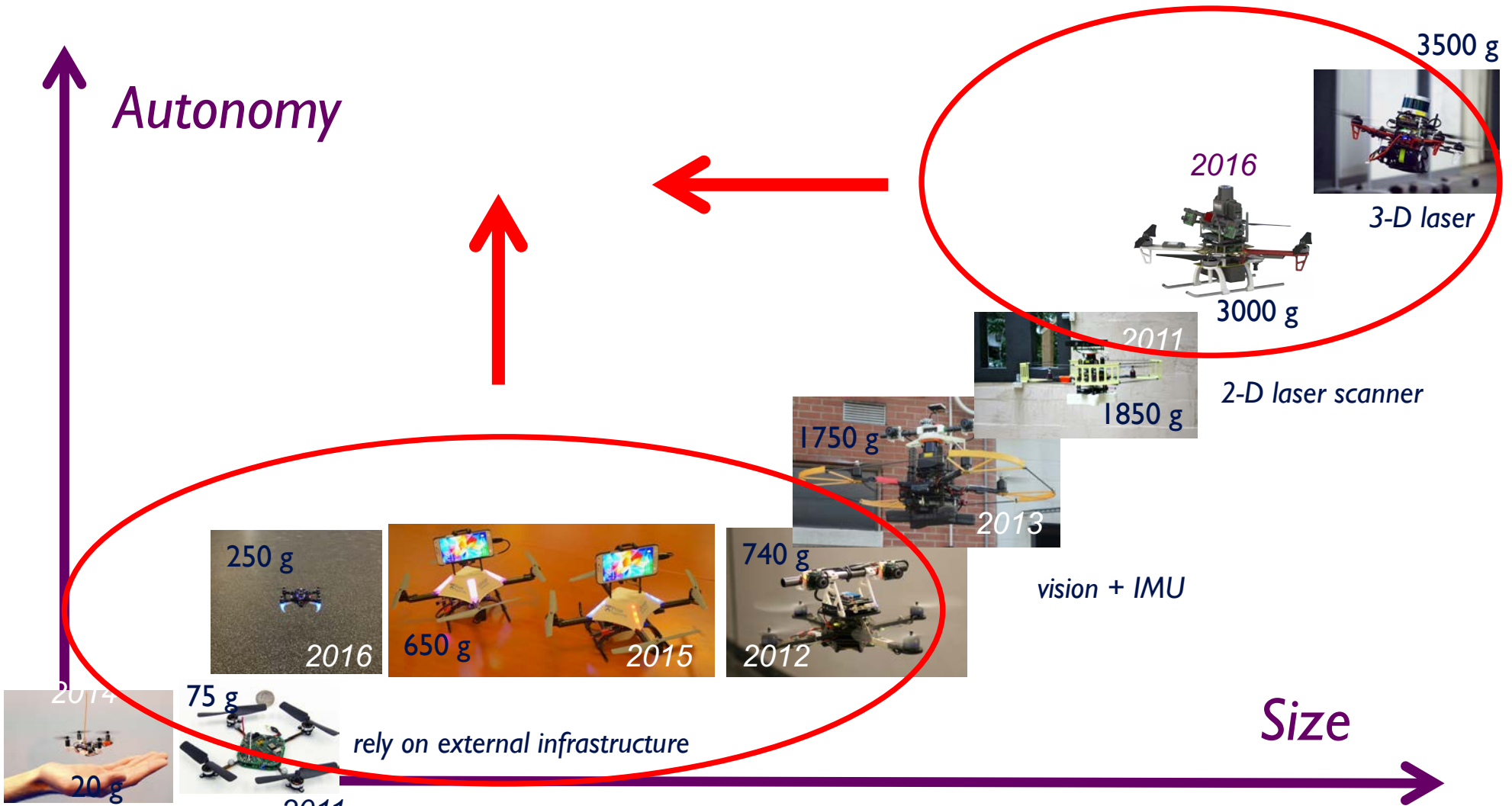
Agriculture, Mining, Health



Aerial Robotics Research (and Commercialization)



Small
We are here! *Safe* *Smart* *Speed* *Swarms*



... and at high speeds



The Falcon

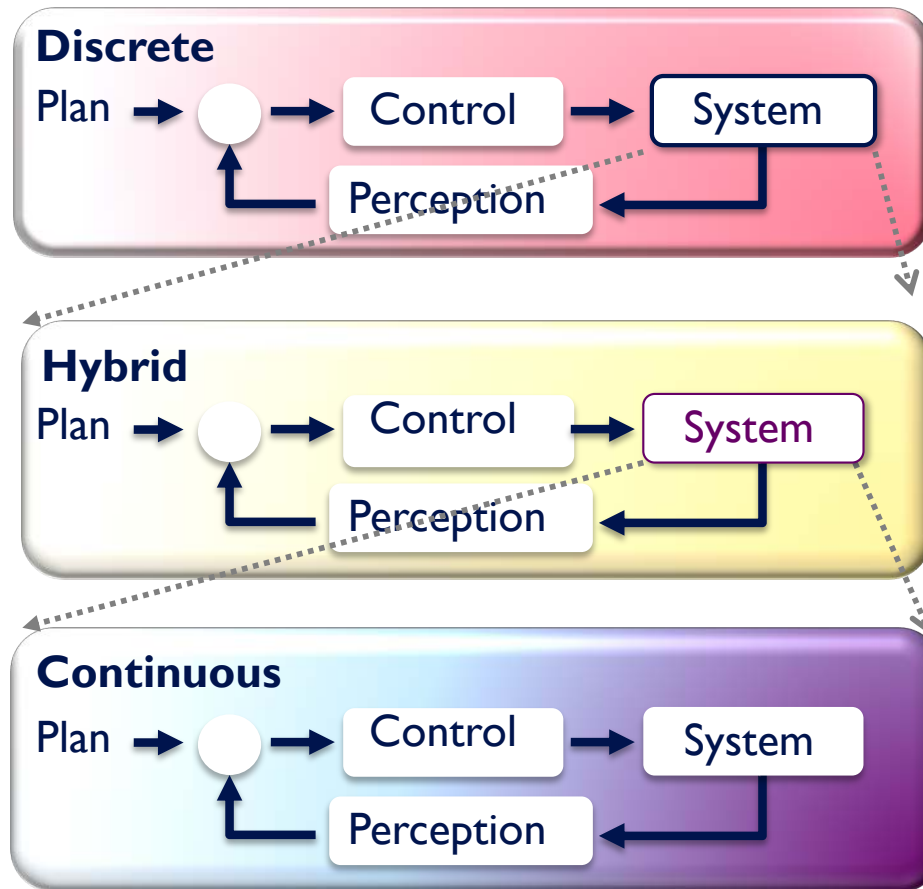


Search and Rescue

Five Challenges

- Perception Action Loops for Autonomy
- State Estimation
- Navigation in Cluttered Environments
- Scaling Down in Size, Weight
- Perception Action Communication Loops for Swarms

I. Nested Perception/Action Loops



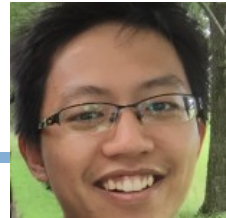
Nonlinear controllers on $SO(3)/SE(3)$

JTY Wen and K Kreutz-Delgado, The attitude control problem, IEEE Transactions on Automatic control, 1991.

D Mellinger and V Kumar, "Minimum Snap Trajectory Generation and Control for Quadrotors," *Proc. IEEE International Conference on Robotics and Automation*. Shanghai, China, May, 2011.

T Lee, M Leok, NH McClamroch, "Nonlinear Robust Tracking Control of a Quadrotor UAV on $SE(3)$," *Asian Journal of Control* 2012.

2. State Estimation (Stereo + IMU)



Ke Sun



Kartik Mohta

Model

$$\mathbf{x}_I = \left(\begin{matrix} {}^I_G \mathbf{q}^\top & \mathbf{b}_g^\top & {}^G \mathbf{v}_I^\top & \mathbf{b}_a^\top & {}^G \mathbf{p}_I^\top & {}^I_C \mathbf{q}^\top & {}^I \mathbf{p}_C^\top \end{matrix} \right)^\top$$

Orientation ${}^I_G \dot{\mathbf{q}} = \frac{1}{2} \Omega(\omega) {}^I_G \mathbf{q}$, $\dot{\mathbf{b}}_g = \mathbf{n}_{wg}$, ${}^G \dot{\mathbf{v}} = C \left({}^I_G \mathbf{q} \right)^\top \mathbf{a} + {}^G \mathbf{g}$, **Cam-IMU extrinsics**
 $\dot{\mathbf{b}}_a = \mathbf{n}_{wa}$, ${}^G \dot{\mathbf{p}}_I = {}^G \mathbf{v}$, ${}^I_C \dot{\mathbf{q}} = \mathbf{0}_{3 \times 1}$, ${}^I \dot{\mathbf{p}}_C = \mathbf{0}_{3 \times 1}$
 Gyro bias Velocity Acc bias Position

Augmented State

$$\mathbf{x}_{C_i} = \left(\begin{matrix} {}^{C_i} \mathbf{q}^\top & {}^G \mathbf{p}_{C_i}^\top \end{matrix} \right)^\top \quad \mathbf{x} = \left(\mathbf{x}_I^\top \quad \mathbf{x}_{C_0}^\top \quad \cdots \quad \mathbf{x}_{C_{N-1}}^\top \right)^\top$$

Stereo Camera Measurement

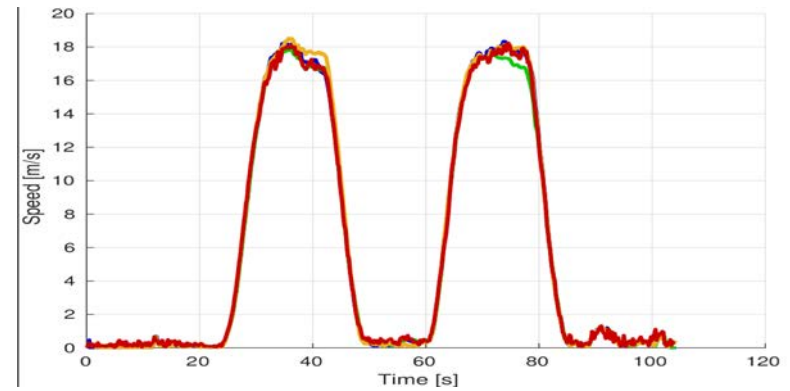
$$\mathbf{z}_i = \begin{pmatrix} u_{i,1} \\ v_{i,1} \\ u_{i,2} \\ v_{i,2} \end{pmatrix} = \begin{pmatrix} \frac{1}{c_{i,1Z}} & \mathbf{0}_{2 \times 2} \\ \mathbf{0}_{2 \times 2} & \frac{1}{c_{i,2Z}} \end{pmatrix} \begin{pmatrix} {}^{C_i,1} X \\ {}^{C_i,1} Y \\ {}^{C_i,2} X \\ {}^{C_i,2} Y \end{pmatrix} + \mathbf{n}_z, \quad {}^{C_i,j} \mathbf{p}_f = \begin{pmatrix} {}^{C_i,j} X \\ {}^{C_i,j} Y \\ {}^{C_i,j} Z \end{pmatrix} \quad j \in \{1, 2\}$$



K. Sun, K. Mohta, B. Pfrommer, M. Watterson, S. Liu, Y. Mulgaonkar, C. J. Taylor, and V. Kumar, "Robust Stereo Visual Inertial Odometry for Fast Autonomous Flight", RAL 2018

Stereo Multistate Constraint Kalman Filter (S-MCKF)

Fast autonomous flight (Top speed at 18m/s)



Autonomous flight in unstructured environment

- Includes various scenes (warehouse, woods, open field, etc).
- Round trip over 700m
- Final drift under 0.5%

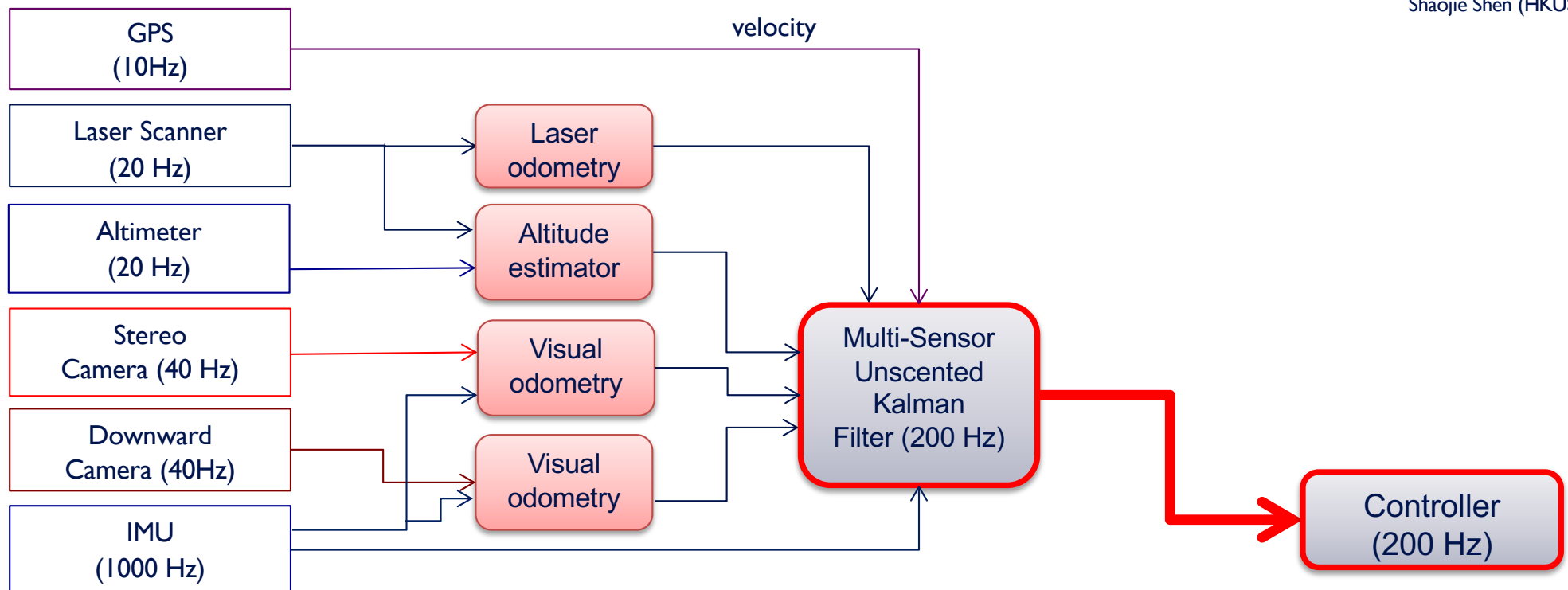


State Estimation

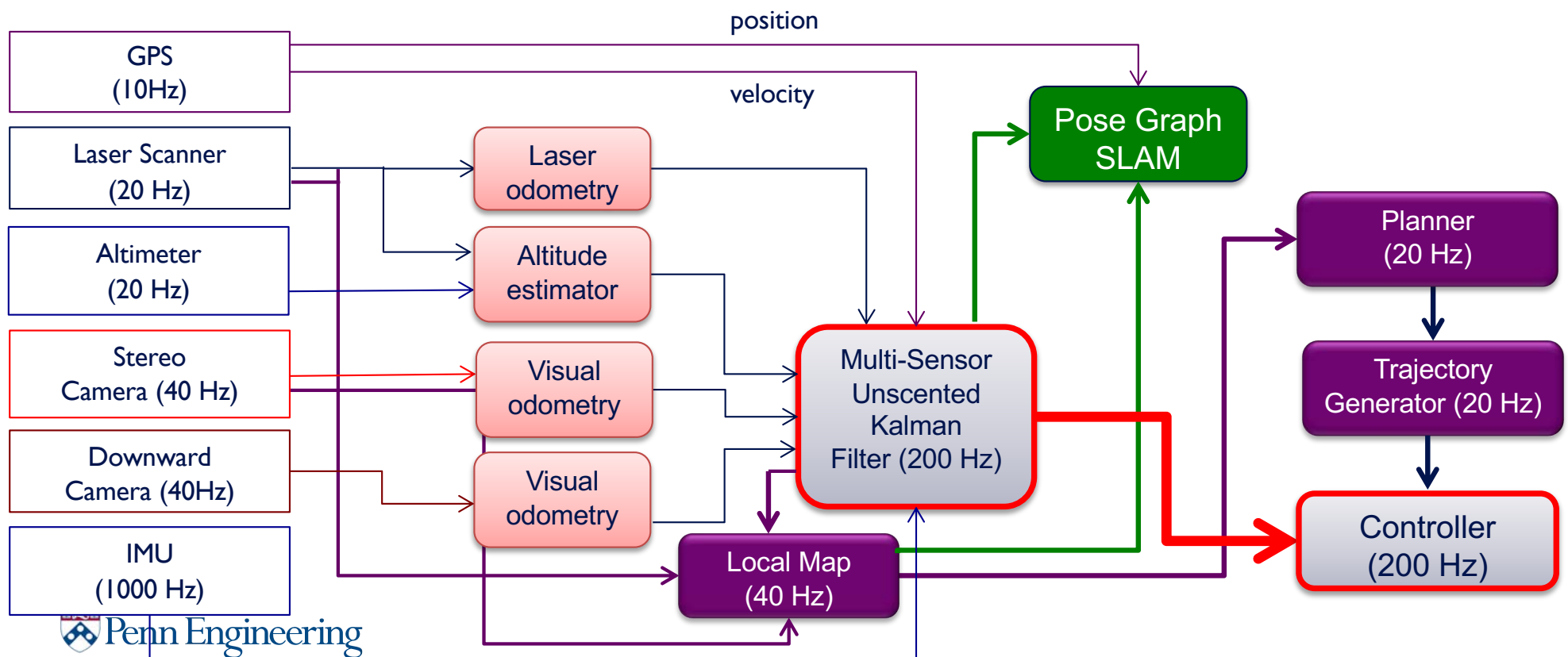
Shaojie Shen, Yash Mulgaonkar, Nathan Michael and Vijay Kumar, "Multi-Sensor Fusion for Robust Autonomous Flight in Indoor and Outdoor Environments with a Rotorcraft MAV," *Proceedings of IEEE International Conference on Robotics and Automation (ICRA)*, 2014.



Shaojie Shen (HKUST)



3. Autonomy: Perception and Action for Navigation



DARPA Fast Lightweight Autonomy (FLA)



Yash Mulgaonkar



Lidar

Cameras + IMU

3 Planning in Cluttered Environments

I. Optimal Control

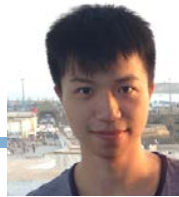
$$\min_{u(t), T} J(x(t), u(t)) + \rho T$$

$$\dot{x} = Ax(t) + Bu(t), u(t) \in \mathcal{U}, \forall t \in [0, T]$$

$$x(0) = x_0, x(T) \in \mathcal{X}^{\text{goal}}, x(t) \in \mathcal{X}^{\text{free}}$$

$$\mathcal{X}^{\text{goal}} \subset \mathcal{X}^{\text{free}}$$

- Relative degree 4 (input and state constraints)
- Non convex
- Safe corridors in different homology classes
- Partially known environment (limited field of view sensors)



Sikang Liu



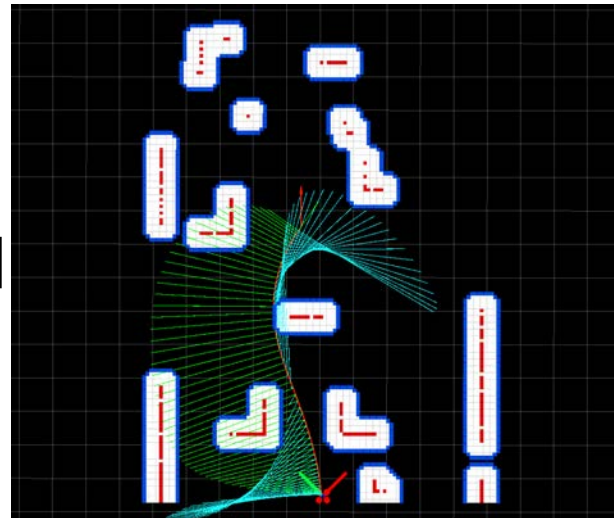
Mike Watterson



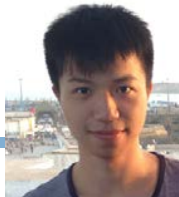
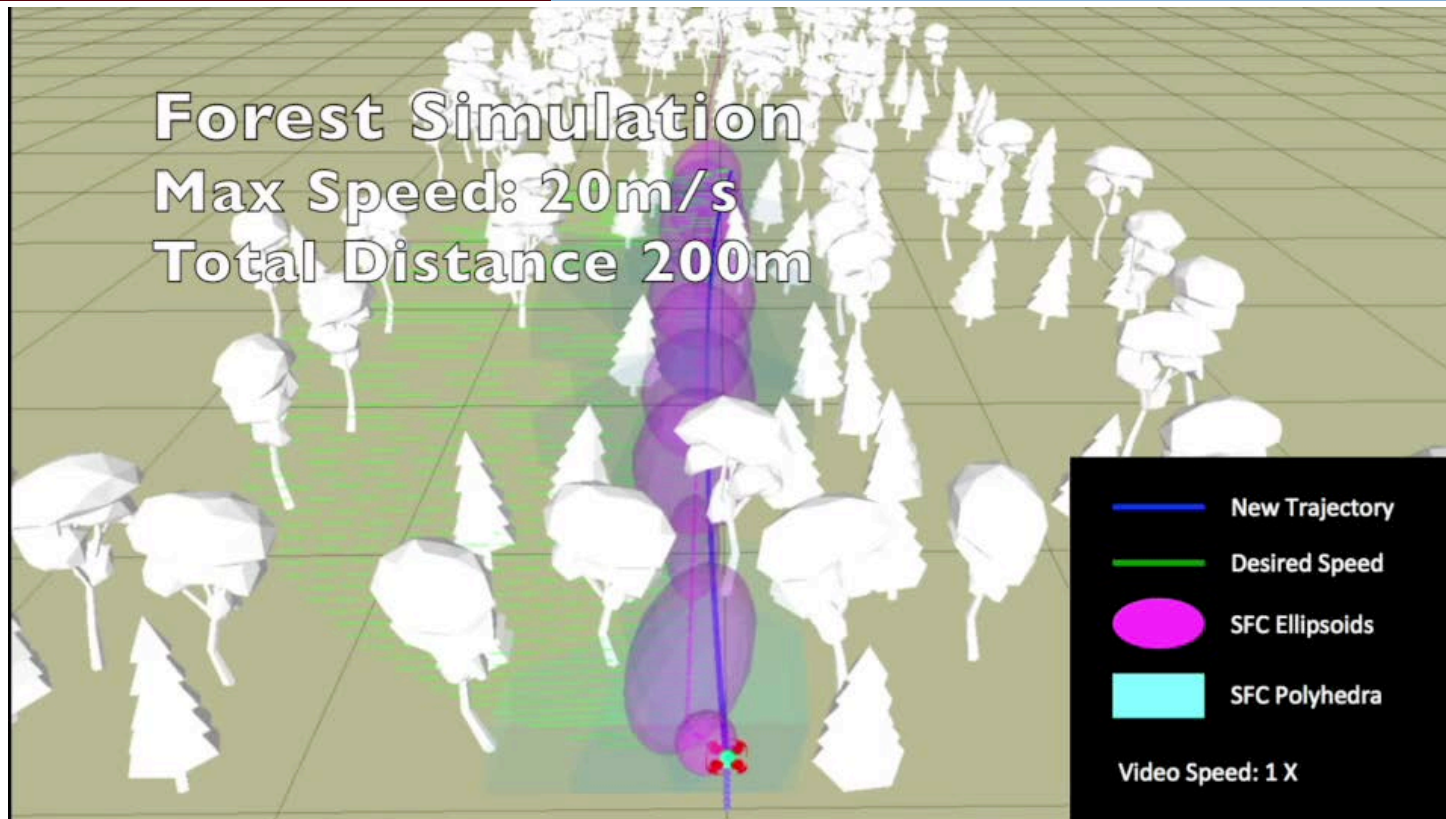
Sarah Tang



Subhrajit
Bhattacharya



FLA (5-20 m/s)



Sikang Liu



Mike Watterson



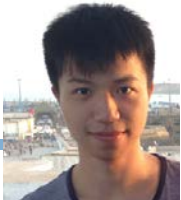
Sarah Tang



Subhrajit
Bhattacharya

- S. Bhattacharya, M. Likhachev and V. Kumar, "Topological Constraints in Search-based Robot Path Planning." *Autonomous Robots*, 33(3):273-290, 2012.
- S. Liu, M. Watterson, S. Tang, and V. Kumar, "High speed navigation for quadrotors with limited onboard sensing," *Robotics and Automation Letters*, 2016.
- S. Liu, N. Atanasov, K. Mohta, and V. Kumar, "Search-based Motion Planning for Quadrotors using Linear Quadratic Minimum Time Control," *IROS* 2017.

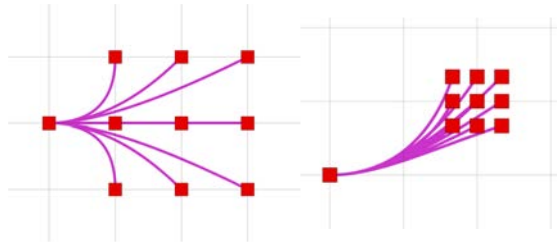
Planning in Cluttered Environments



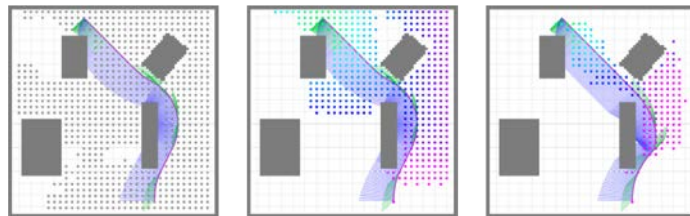
Sikang Liu

2. Search-Based Planning with Motion Primitives

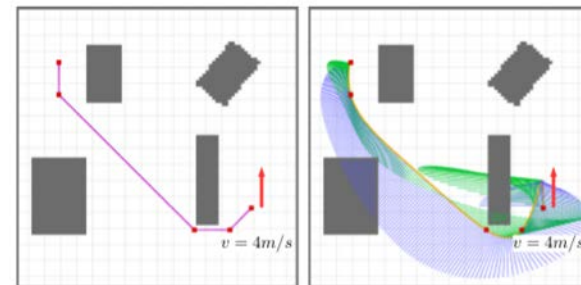
Minimum snap primitives



Search over induced discretization on state space

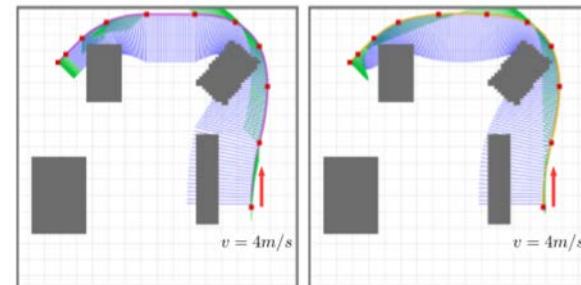


Results for different functionals



(a) $T = 8.5$.

(b) $T = 8.5, J = 296.6$.



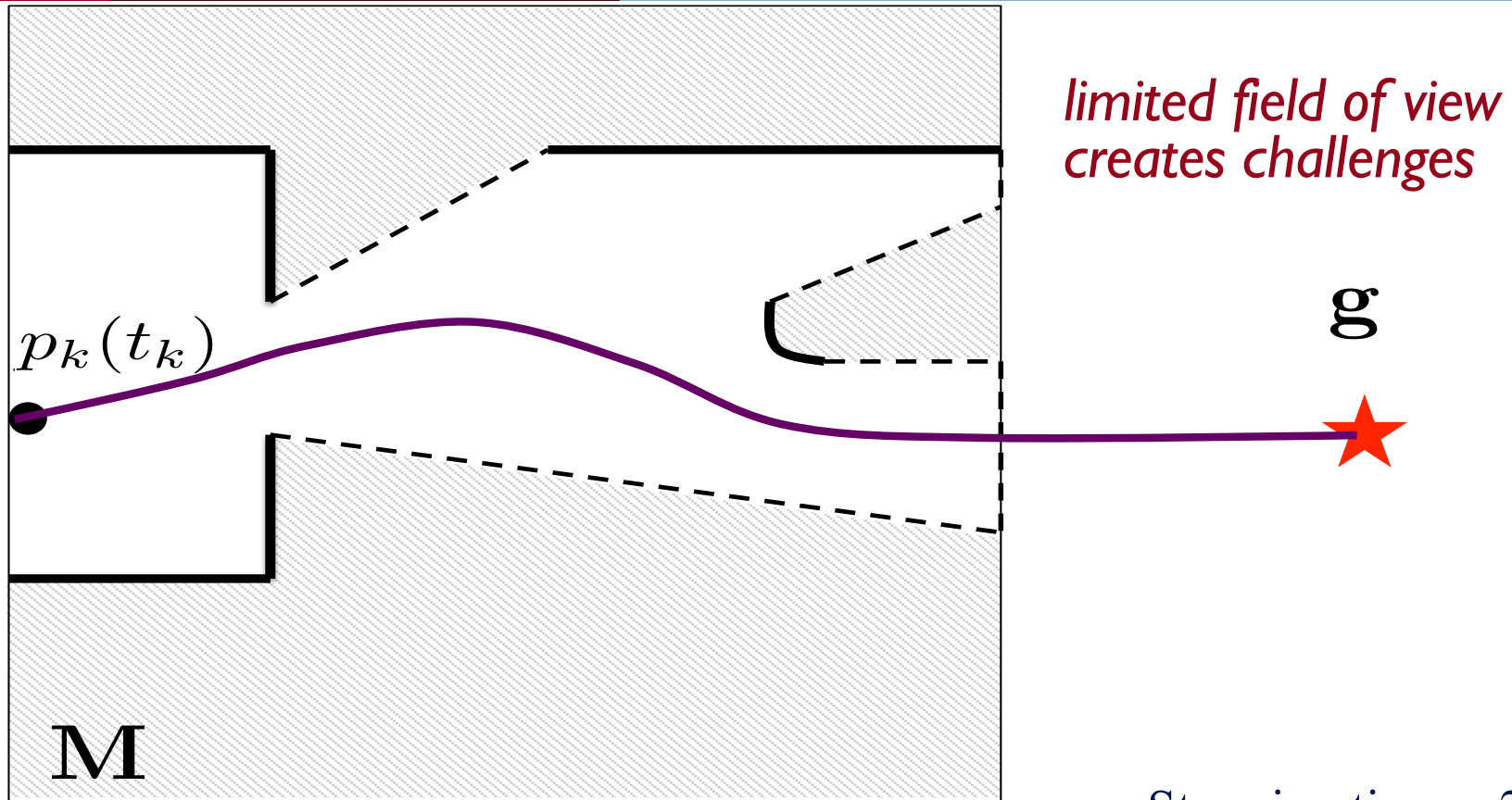
(c) $T = 10, J = 14.0$.

(d) $T = 10, J = 21.3$.



Nikolay Atanasov
(UCSD)

Resolution complete but ...



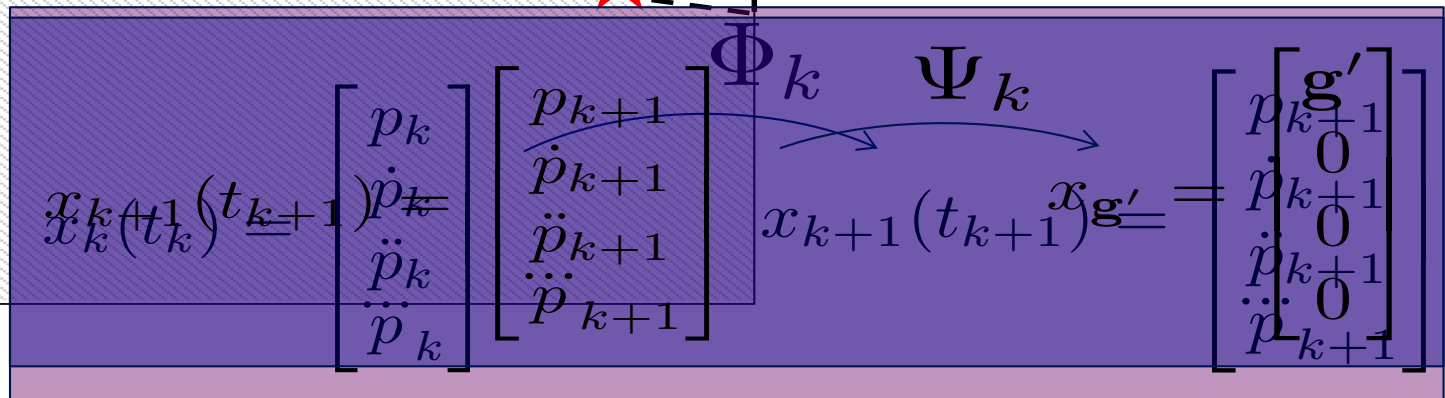
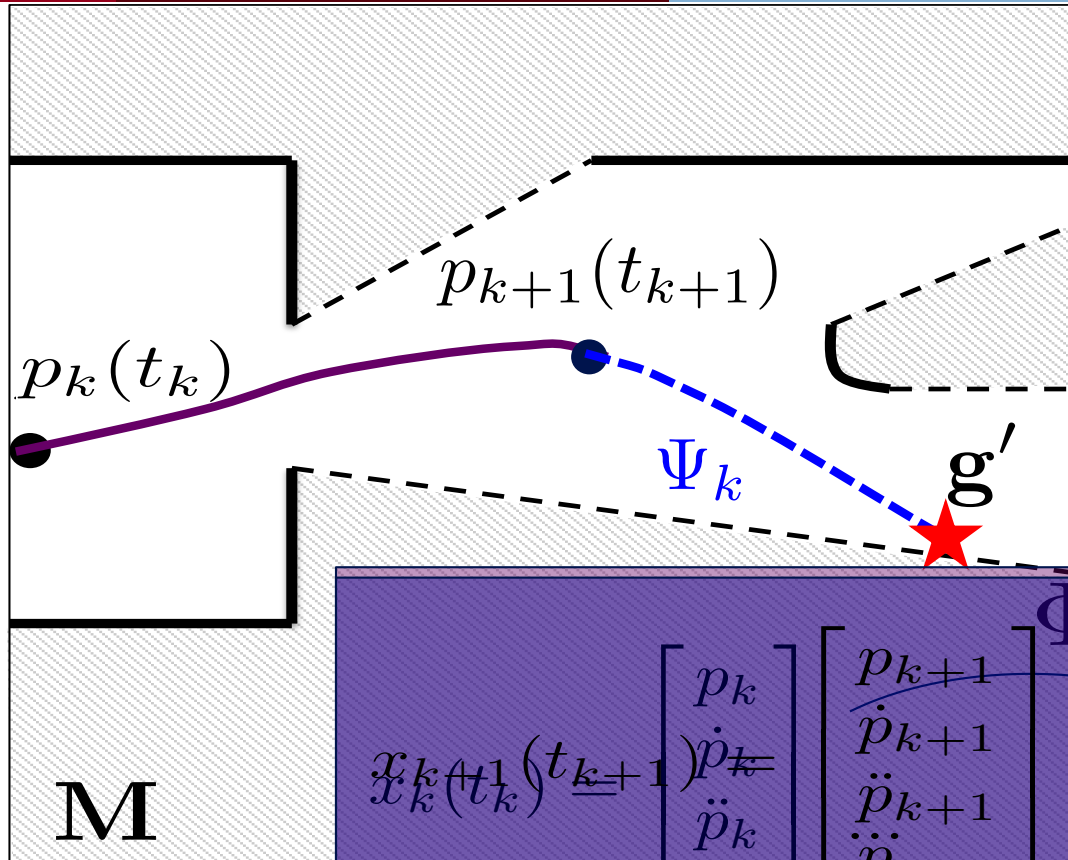
*limited field of view
creates challenges*

 Penn Engineering $v = 20$ m/s, max acceleration $1 g$ \Rightarrow Stopping time ~ 2 s
Stopping distance ~ 20 m

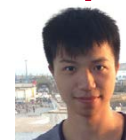
Safety Certificate

- Safety
- Completeness
- Suboptimality

g
★



Autonomous Flight in Unknown GPS-Denied Environment (5 m/s)



Sikang Liu



Mike Watterson



Kartik Mohta



Ke Sun

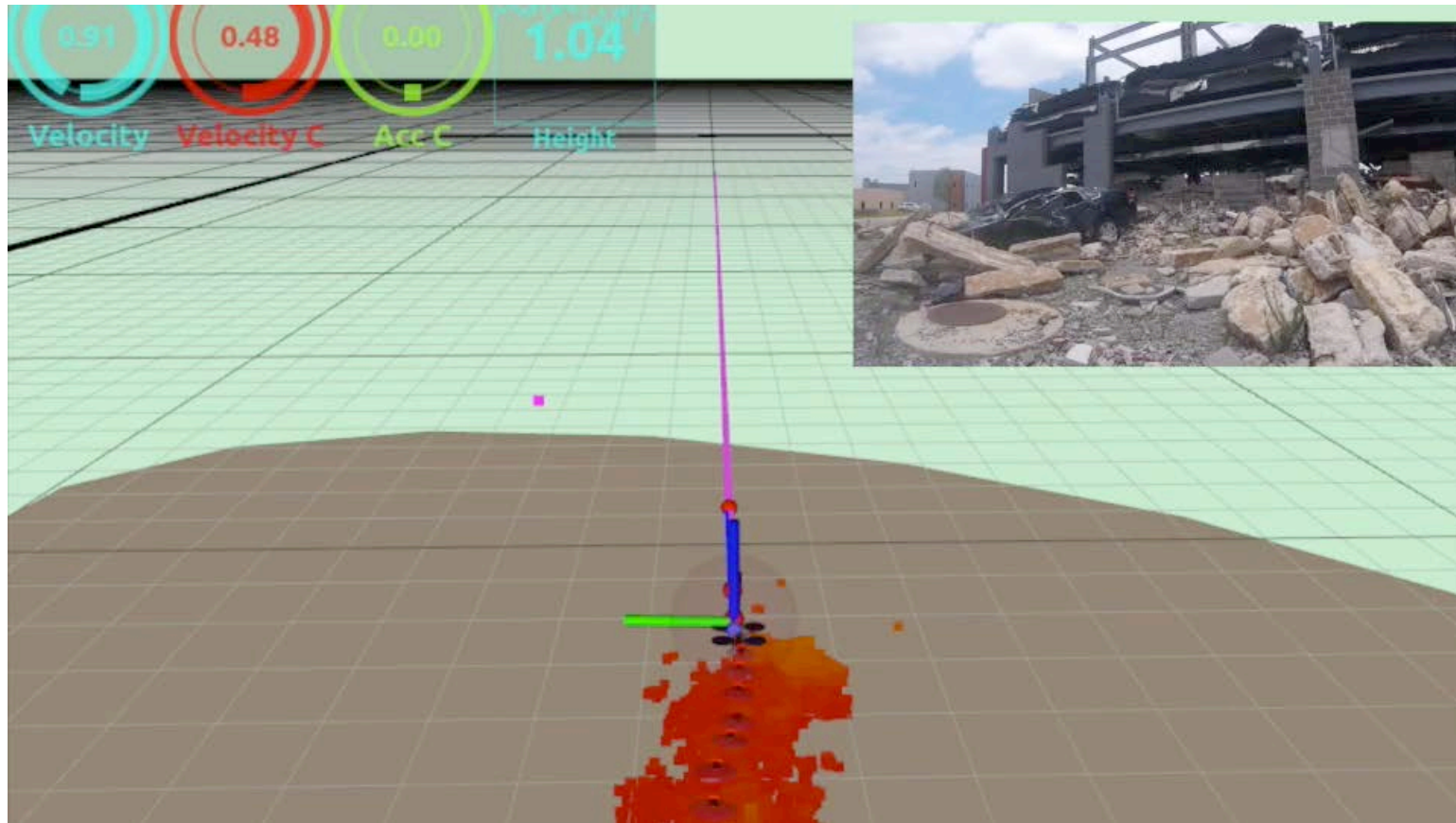


Yash Mulgaonkar

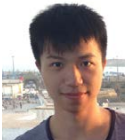


CJ Taylor

Search of Collapsed Buildings



Shreyas Shivakumar



Sikang Liu



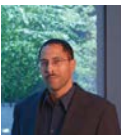
Mike Wattersor



Kartik Mohta



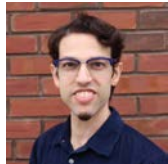
Ke Sun



CJ Taylor



Raghu Balasa



Alex Burka



Amilcar Cipriano



Monica DeGuzman



Jason Derenick



Xuchu (Dennis) Ding



Brandon Duick



Nader Elm



Pete Furlong



Gray Greene



Vinutha Kallem



Nick Lynch



Billy Sisson



James Sui



Justin Thomas



Denise Wong



exyn
technologies



www.exyn.ai

Fully Autonomous Aerial Robot for Mine Inspection and Mapping

Stope Flight: Beyond Visual Line-of-Sight and Communications Range



4. Light Weight Autonomy



2.5 kg quadrotor (2017)

Stereo camera synced with Vector NAV IMU, LiDar, Intel i7



1 kg quadrotor (2018)

Stereo camera synced with Vector NAV IMU, NVIDIA Jetson TX2 + FPGA (low-level pixel-wise operations) – OSRF TOF 3-D camera, 6m range, 100x65 deg, 60 Hz – PMD technologies



250 gram quadrotor (2018)

Qualcomm® Snapdragon Flight™ development board running Snapdragon Navigator™ flight controller and Machine Vision (MV) SDK

S. S. Shivakumar, K. Mohta, B. Pfrommer, V. Kumar and C. J. Taylor, Guided Semi Global Optimization for Real Time Dense Depth Estimation, ICRA 2019.

A. Weinstein, A. Cho, and G. Loianno, and V. Kumar, "VIO-Swarm: A Swarm of 250 gram autonomous", ICRA 2018



Giuseppe Loianno



Shreyas Shivakumar



Kartik Mohta

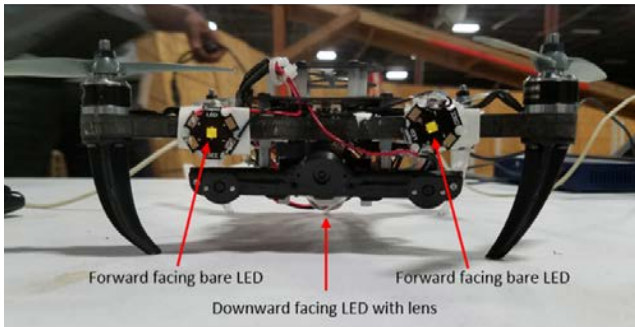


Ke Sun



C.J. Taylor

Robustness to Collisions



250 gram quadrotor

Qualcomm® Snapdragon Flight™ board with
Snapdragon Navigator™ flight controller



Tiercel

133 gram quadrotor capable of sustaining collisions

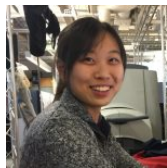
Qualcomm® Snapdragon Flight™ board with forward-facing stereo cameras, a downward facing camera for VIO, onboard WiFi and GPS



Yash Mulgaonkar, Wenxin Liu, Dinesh Thakur, Kostas Daniilidis, Vijay Kumar, The Tiercel: A novel autonomous micro aerial vehicle that can map the environment by flying into obstacles, IEEE Robotics and Automation Letters, submitted (2020)



Yash Mulgaonkar



Wenxin Liu

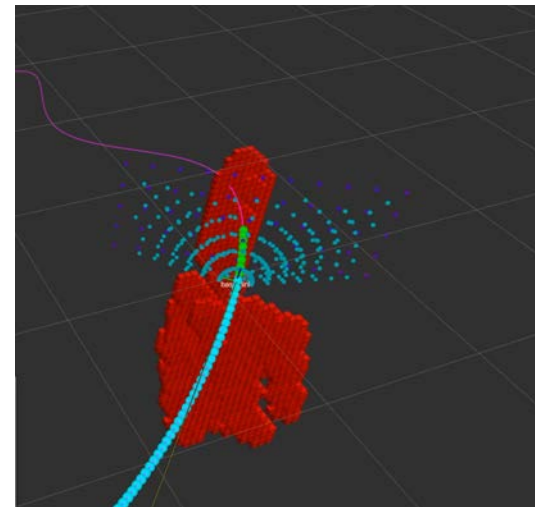
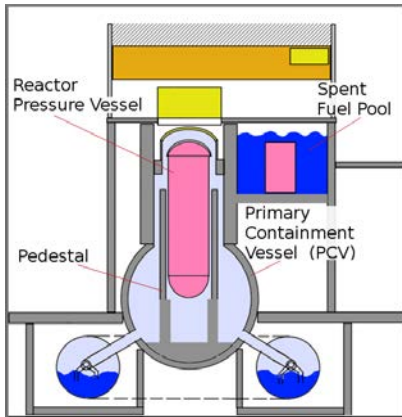


Giuseppe Loianno



Dinesh Thakur

Autonomous Flight in Fukushima Daiichi Reactor Unit 1



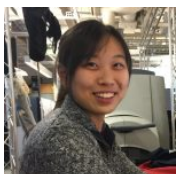
Dinesh Thakur



Giuseppe Loianno



Laura Jarin-Lipschitz

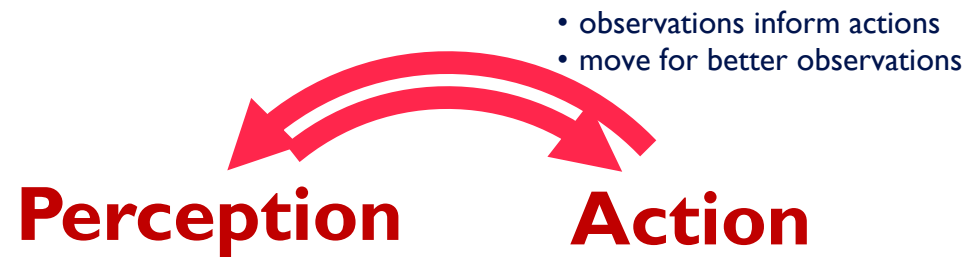


Wenxin Liu



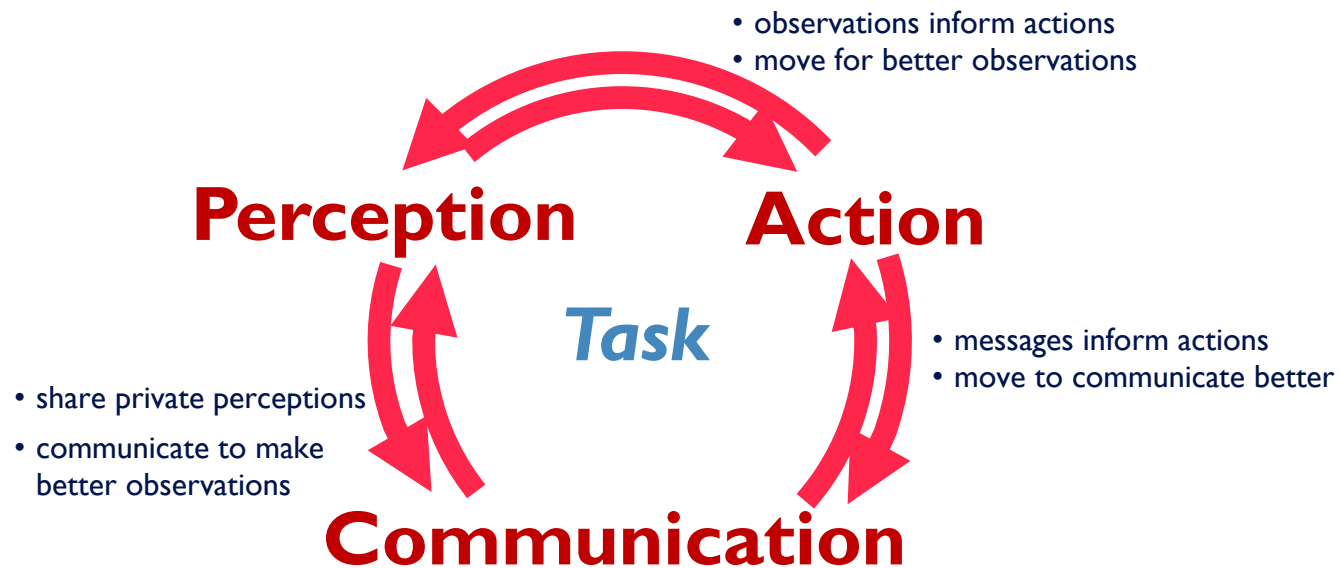
Elijah Lee

5. Aerial Robot Swarms



Perception—Action Loops

5. Aerial Robot Swarms

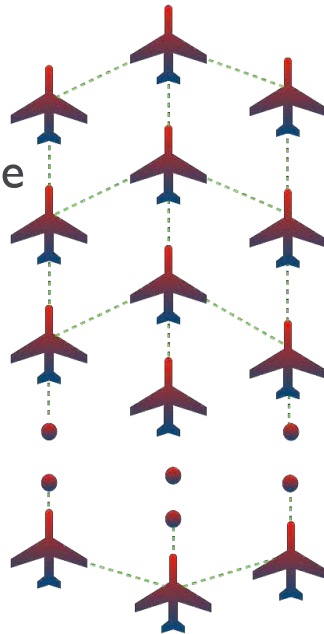


Perception—Action—Communication Loops

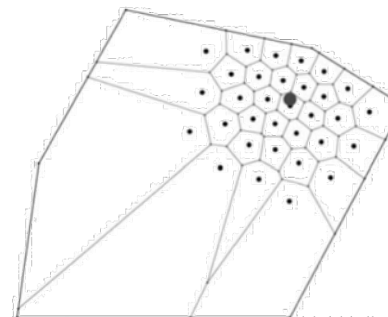
Decentralized Multi-Robot Teams

- Centralized methods are not practical for real-world robot deployments (Turpin, '14)
 - Partial observability by individual agents
 - Limited communication
- Decentralized, correct-by-construction policies available only for very simple cases
 - simple communication and sensing models
 - edge or cloud computation
 - point robots

Tanner, Pappas and Jadbabaie, 2004



A. Weinstein, A. Cho, and G. Loianno, and V. Kumar, "VIO-Swarm: A Swarm of 250 gram autonomous quadrotors" ICRA 2018.



Sensor Coverage
Cortes, 2004

Belta and Kumar, 2005



Aaron Weinstein



Giuseppe Loianno

Distributed Learning: PAC Loops



Jimmy Paulos



Ekaterina Tolstaya



Arbaaz Khan



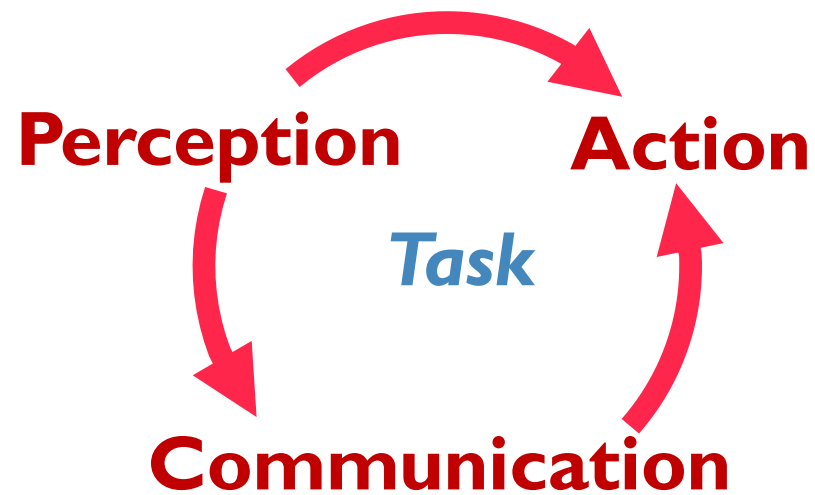
Steven Chen



Dinesh Thakur



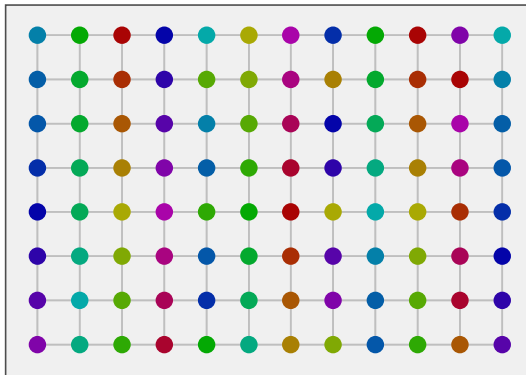
Laura Jarin-Lipschitz



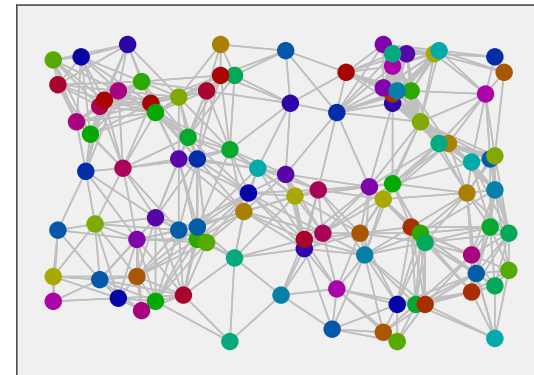
➤ Key Ideas:

- Learn **communication policies**
- Learn **action policies**
- Learn **planning policies**

Graph Neural Networks



CNNs

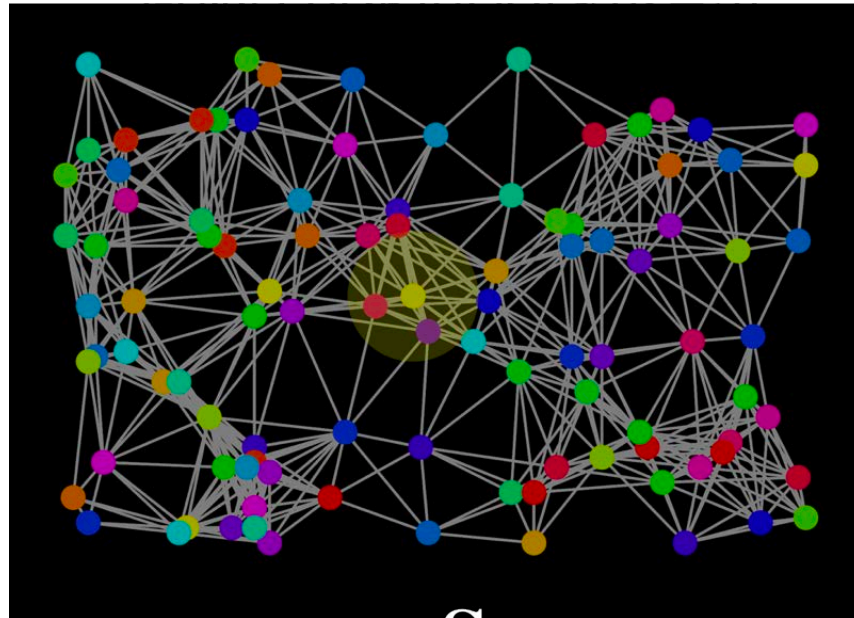


GNNs

Aggregate information at each node from neighboring node using graph adjacency properties

- Robots act on relative position and velocity information
 - Must stay close to each other
 - Must avoid collisions
 - Must “align” themselves

Aggregate over belief states of neighbors



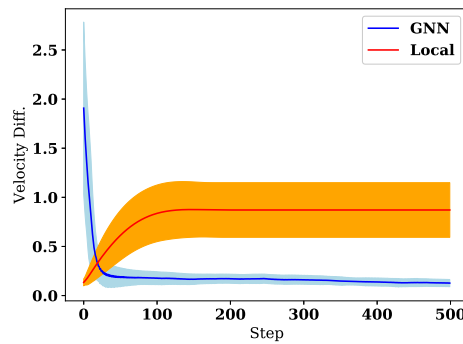
$$\mathbf{z} = \sum_{k=0}^K h_k \mathbf{S}^k \mathbf{x}^{(k)} = \mathbf{H}(\mathbf{S}) \mathbf{x}$$

Flocking

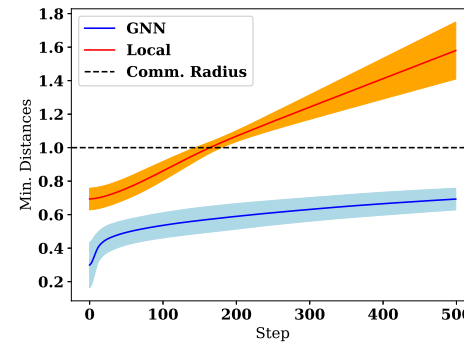
GNN-learned control policy (K=3)

$$\mathbf{u}_i^\dagger = - \sum_{j \in \mathcal{N}_i} (\mathbf{v}_i - \mathbf{v}_j) - \sum_{j \in \mathcal{N}_i} \nabla_{\mathbf{r}_i} U(\mathbf{r}_i, \mathbf{r}_j).$$

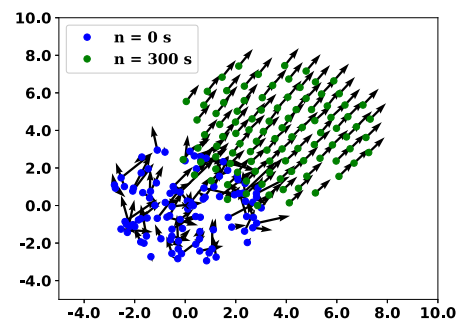
Tanner, Pappas and Jadbabaie, 2004



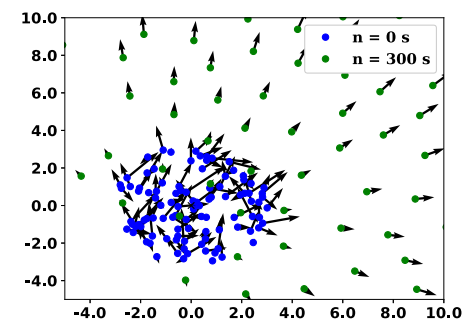
(a) Average difference in velocities



(b) Average minimum distance to a neighbor



(c) Flock positions using the GNN



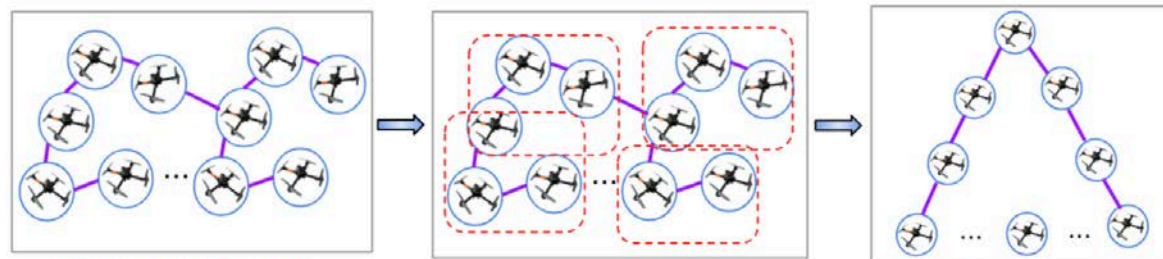
(d) Flock positions using the local controller

Graph Policy Gradients



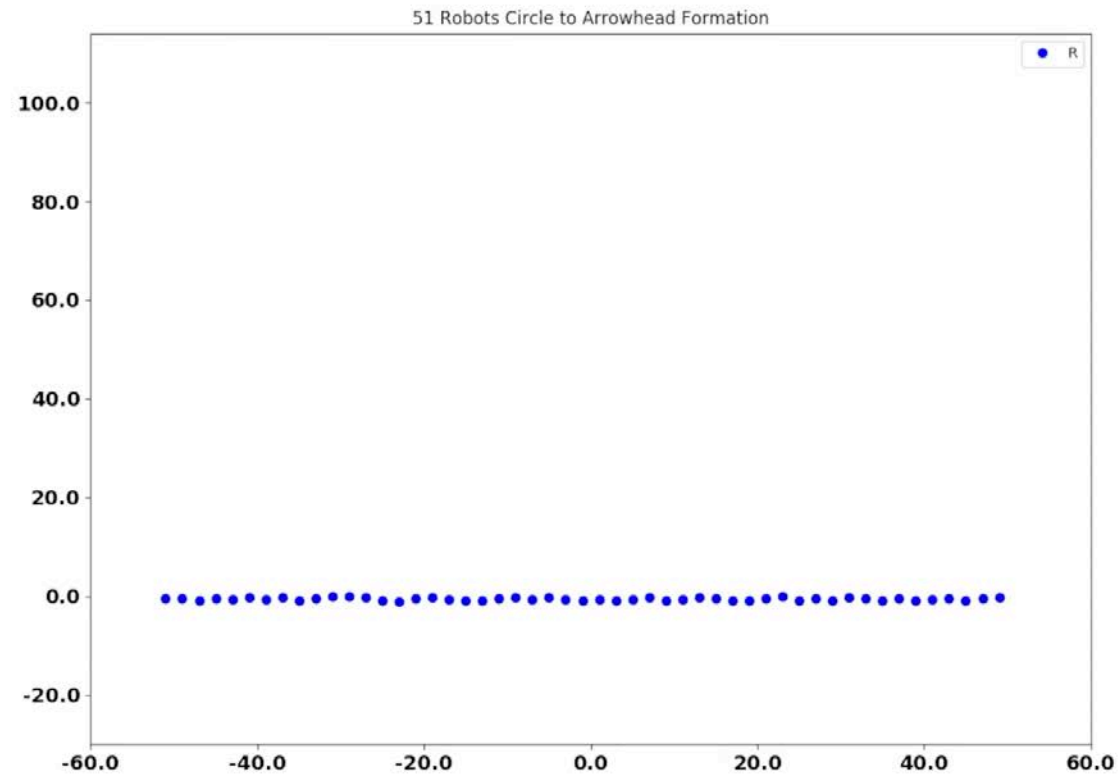
Arbaaz Khan

- Train GNNs on a small number of robots
 - Information from k-hop neighbors is aggregated by each robot
 - Local controllers are learned by each robot
 - Centralized reward used to train the robots
- Extend to swarms with larger numbers
 - Transfer of policies to larger groups with similar “local” graph properties



Arbaaz Khan, Vijay Kumar, and Alejandro Ribeiro, Graph Policy Gradients for Large Scale Unlabeled Motion Planning with Constraints, IEEE International Conference on Robotics and Automation, submitted (2020)

Graph Policy Gradients for Large Scale Formation Control



Conclusion

- Autonomy using smartphone grade processors/sensors
- 10x improvement in performance/price
- Applications to search and rescue and precision agriculture
- Integration of model-based and data-driven methods

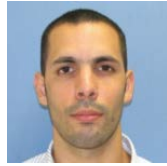
AI 1.0  AI 2.0  AI 3.0



Elizabeth Beattie



Steven Chen



Avi Cohen



Sambeeta Das



Jnaneshwar Das



Luis Guerrero



Laura Jarin-Lipschitz



Jimmy Paulos



Jim Keller



Arbaaz Khan



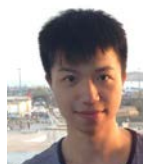
Monroe Kennedy



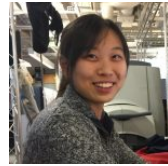
Elijah Lee



Rebecca Li



Sikang Liu



Wenxin Liu



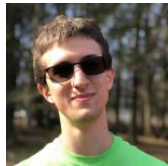
Giuseppe Loianno



Kartik Mohta



Dan Mox



Ian Miller



Yash Mulgaonkar



Ty Nguyen



Tolga Özaslan



Chao Qu



Rattanachai Ramaititima



Kelsey Saulnier



Ke Sun



James Svacha



David Saldana



Daigo Shishika



Ed Steager



Dinesh Thakur



Sarah Tang



Ekaterina Tolstaya



Mike Watterson



Mickey Whitzer



Aaron Weinstein