

UAV Traffic Management (UTM)

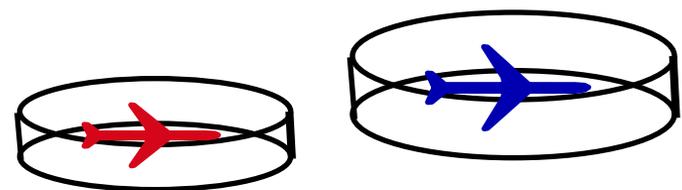
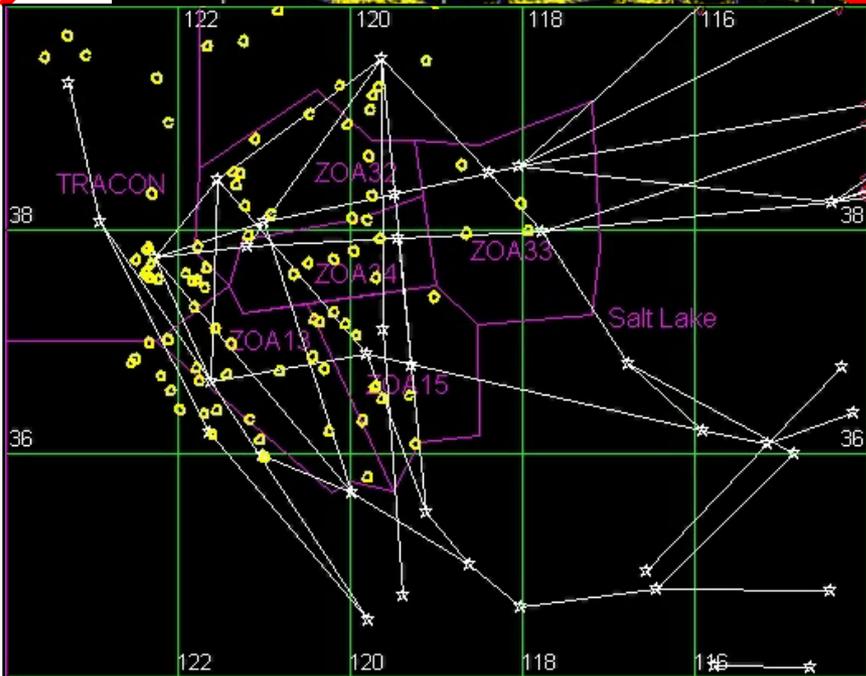
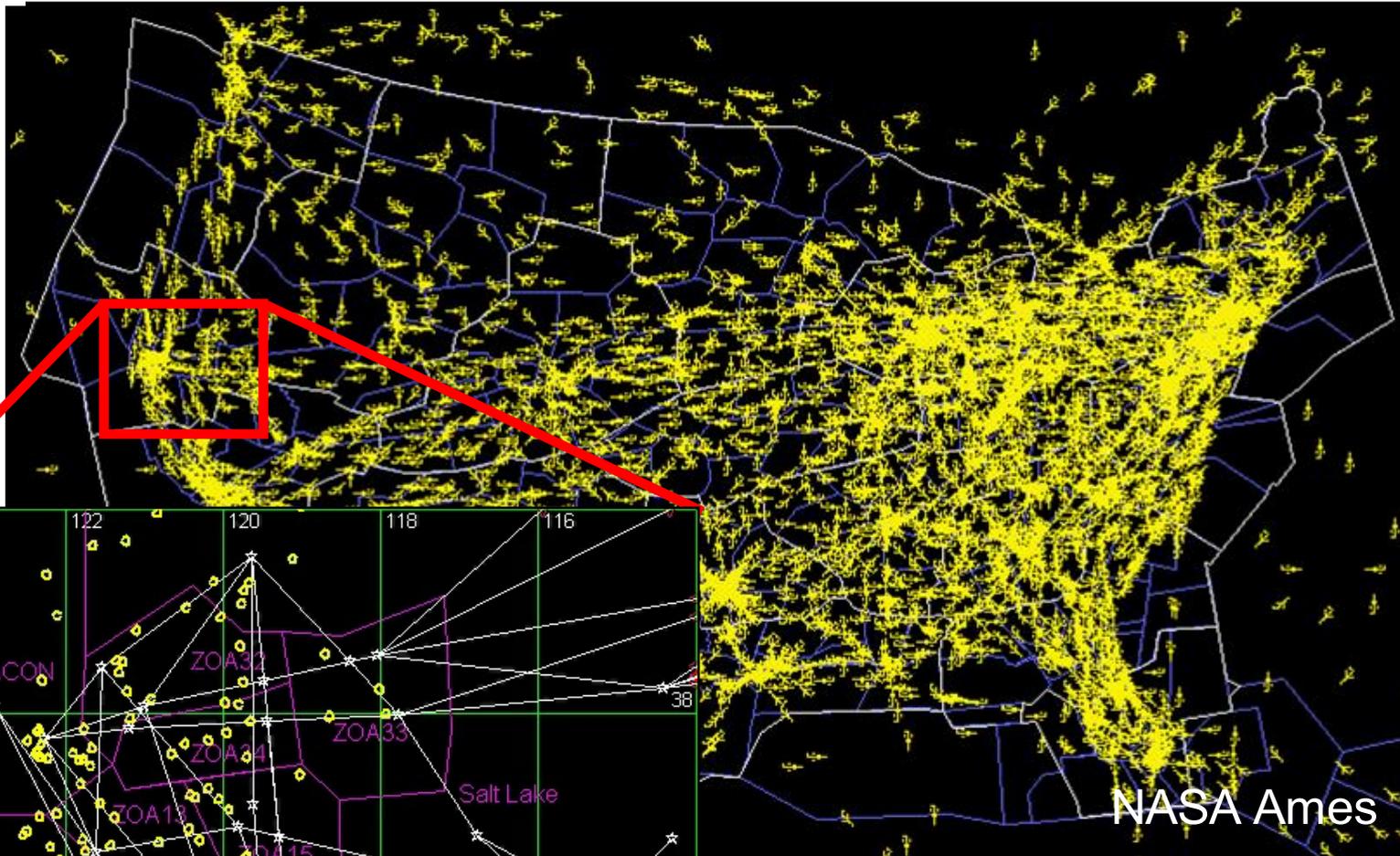
Claire Tomlin



Department of Electrical Engineering and Computer Sciences
University of California at Berkeley

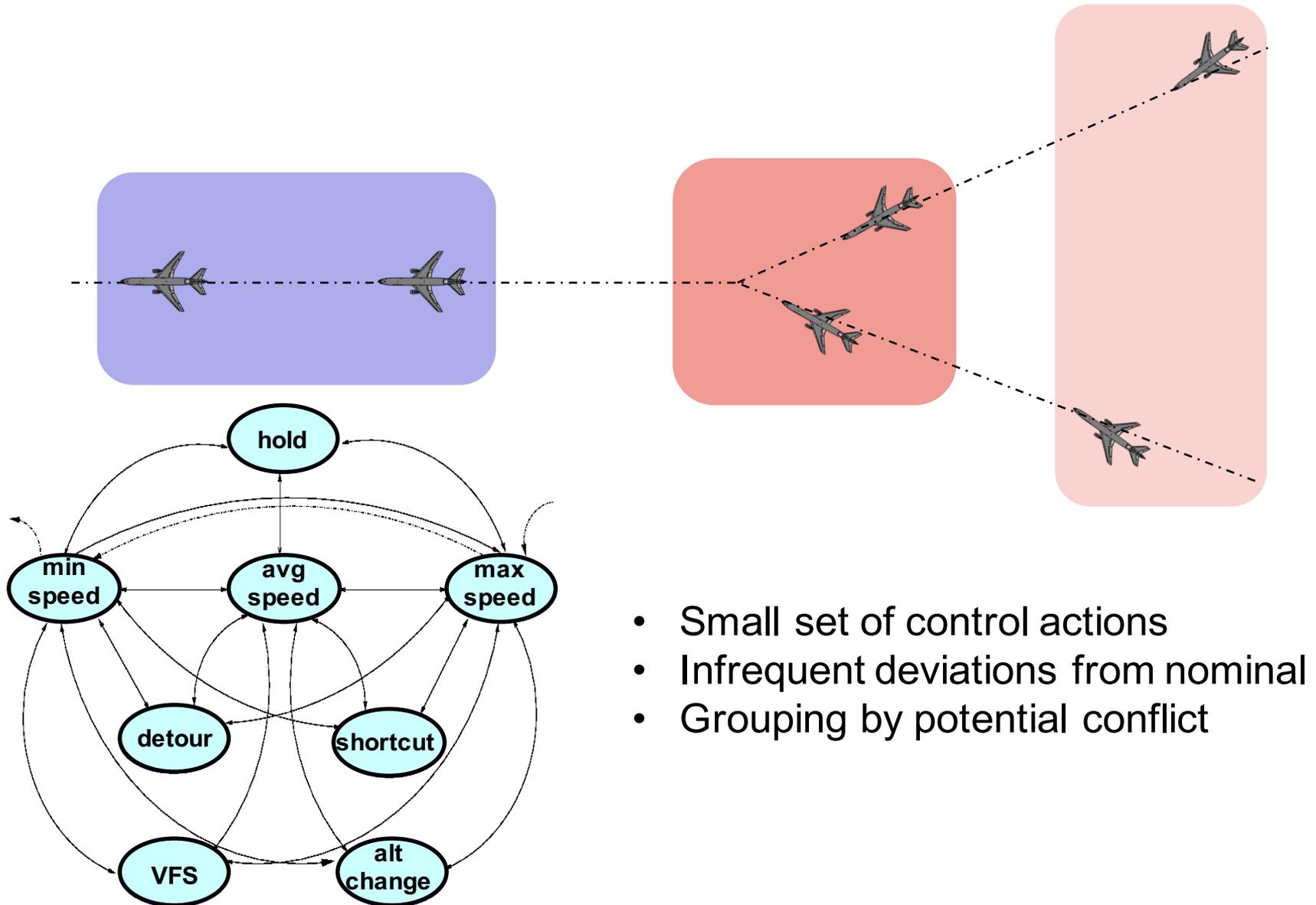
February 9 2018

Air Traffic Control



[FACET, ETMS, NASA Ames]

Controller must keep aircraft separated



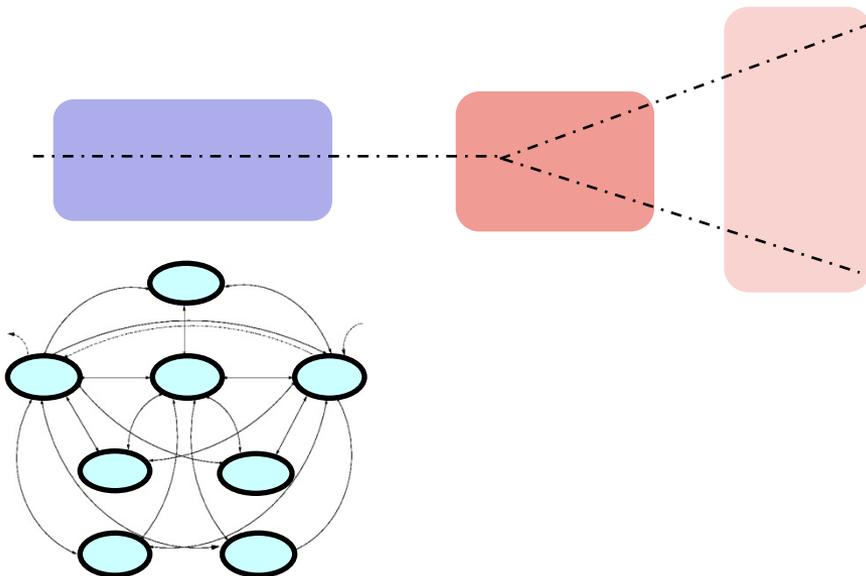
Growing numbers of UAV applications



[Amazon]



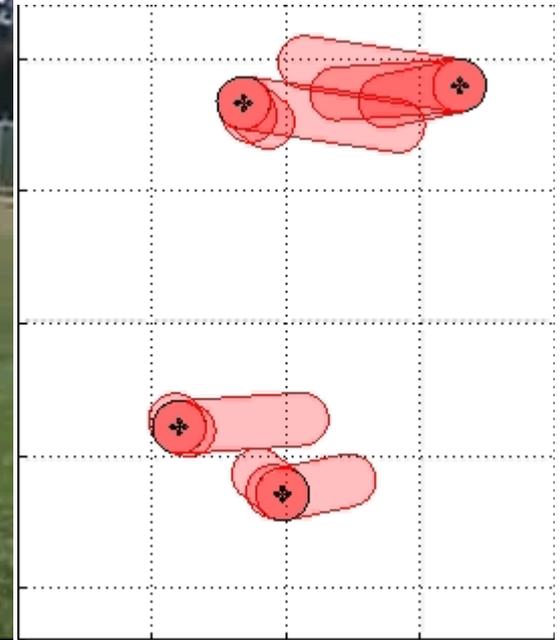
[Google]



1. Safety
2. Simplicity
[computation]
3. Ability to adapt to new information
[data-driven]

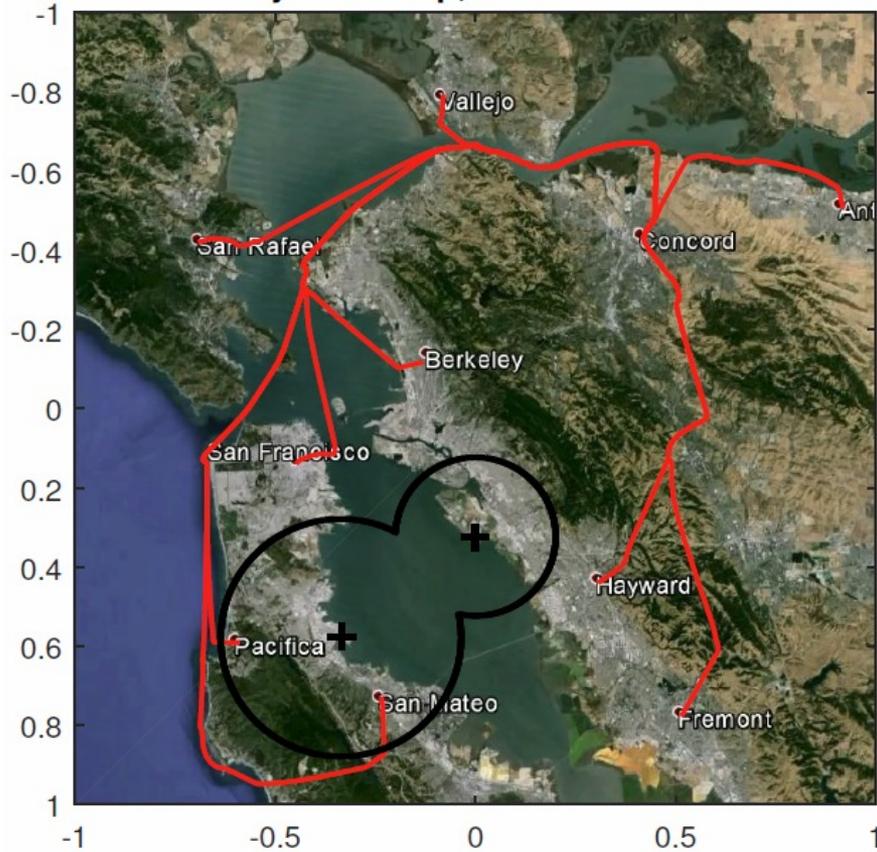
Example 1: Collision Avoidance

Pilots instructed to attempt to collide vehicles

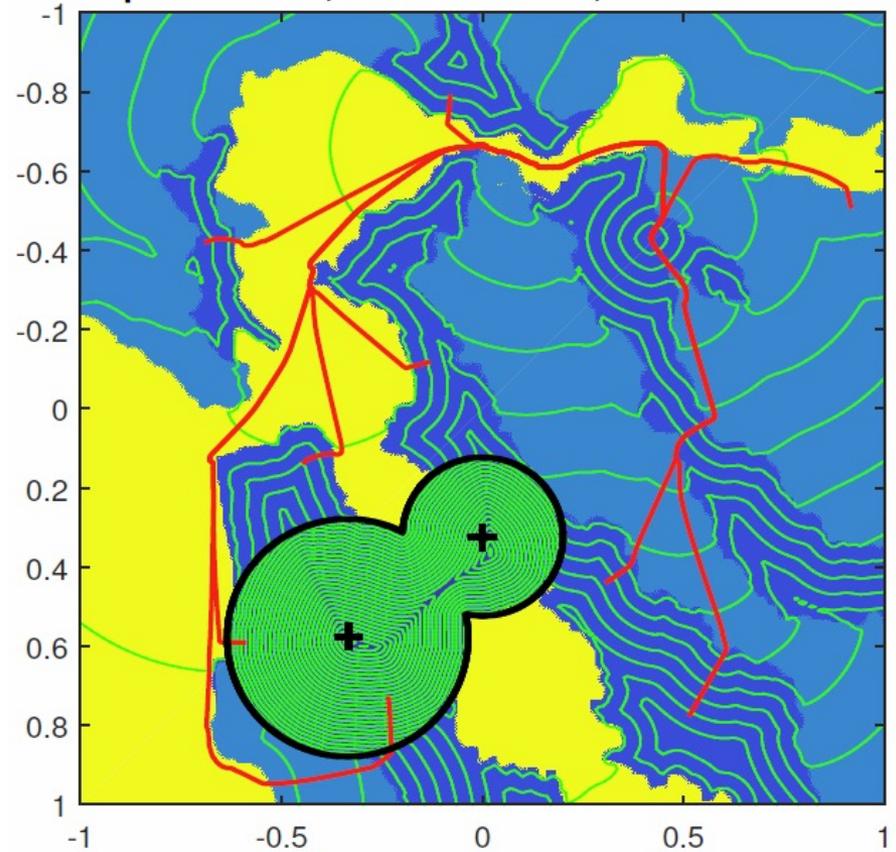


Example 2: Platooning UAVs

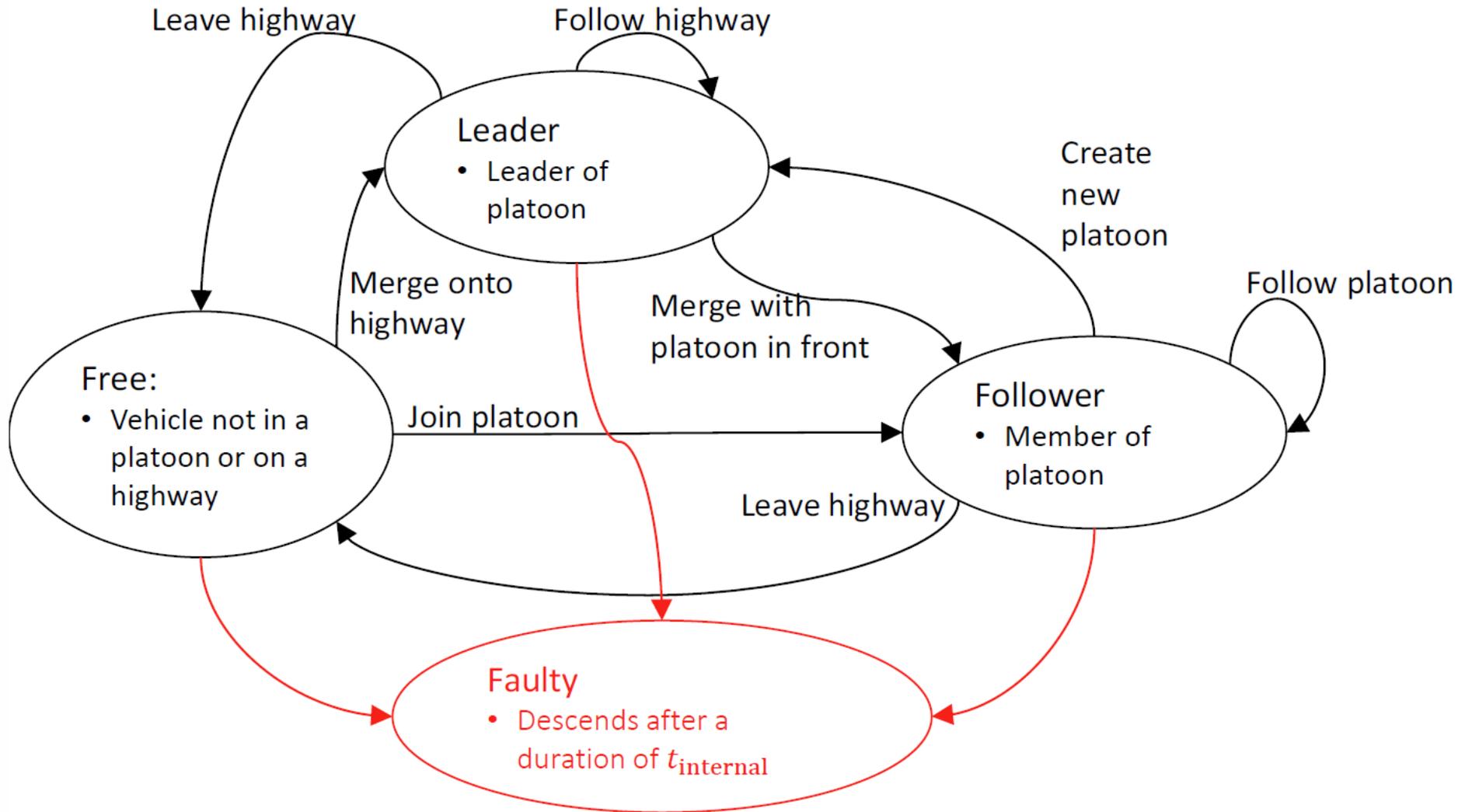
Bay Area Map, Shortest Paths



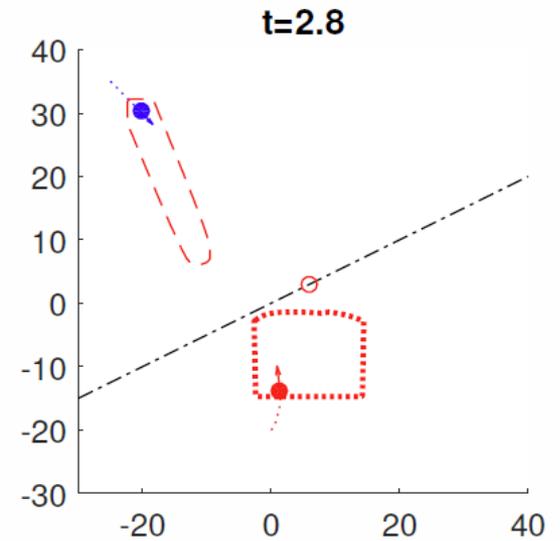
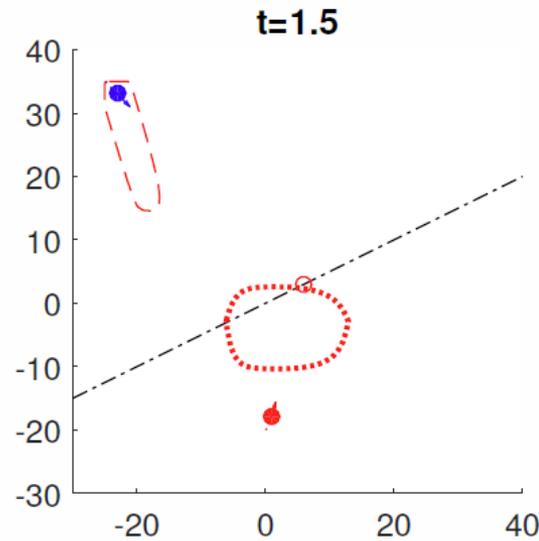
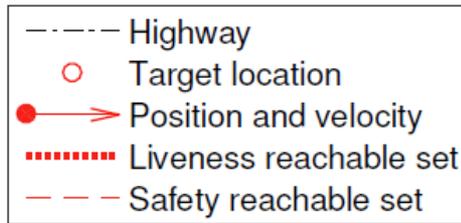
Speed Profile, Shortest Paths, Value Function



Example 2: Platooning UAVs

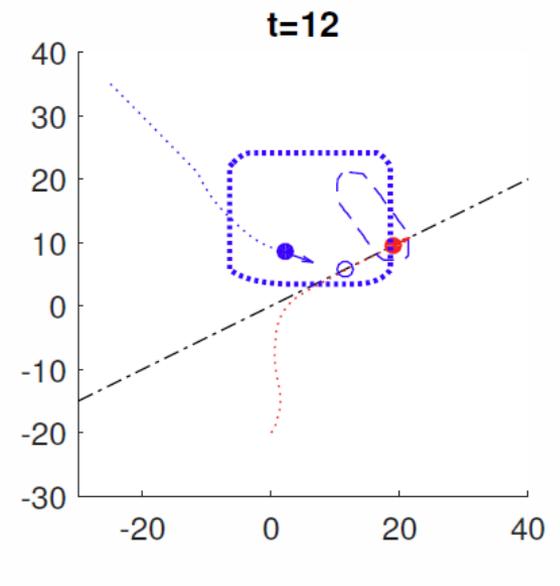
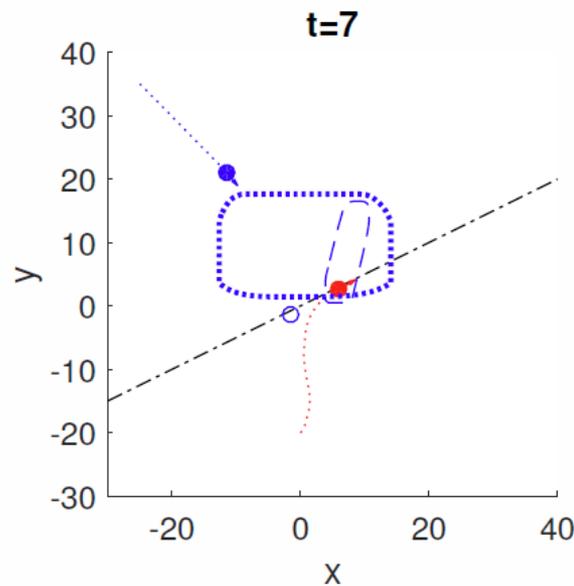


Merge and join

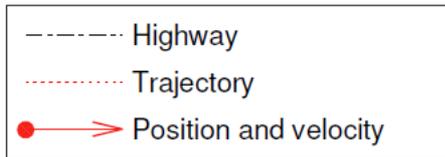


Red vehicle merges onto highway

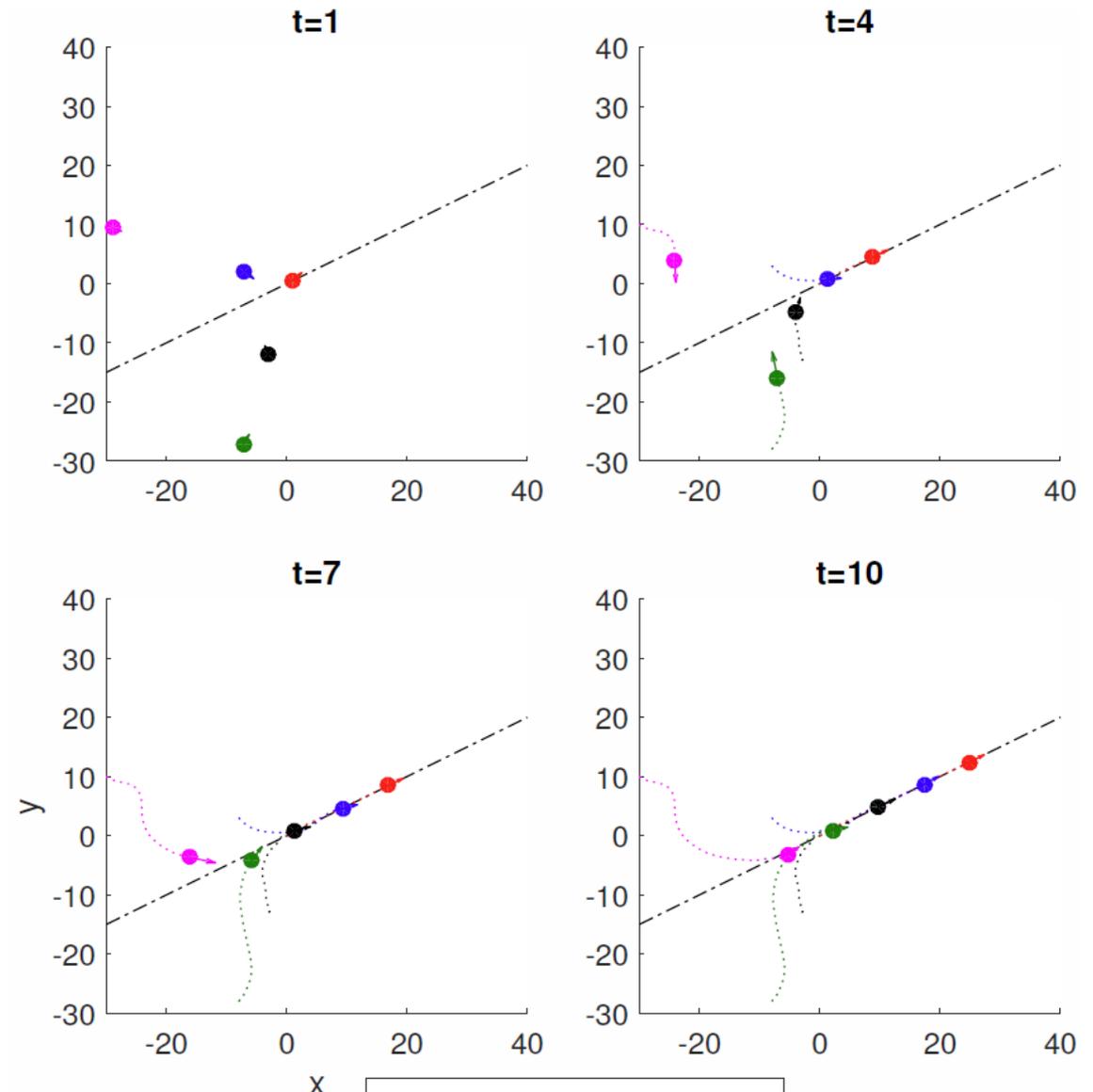
Blue vehicle joins red vehicle's platoon



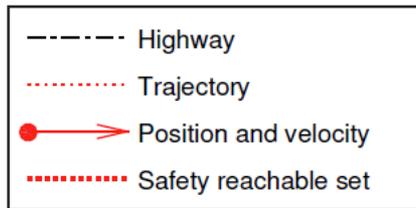
Merge and join



4 vehicles join platoon following red vehicle



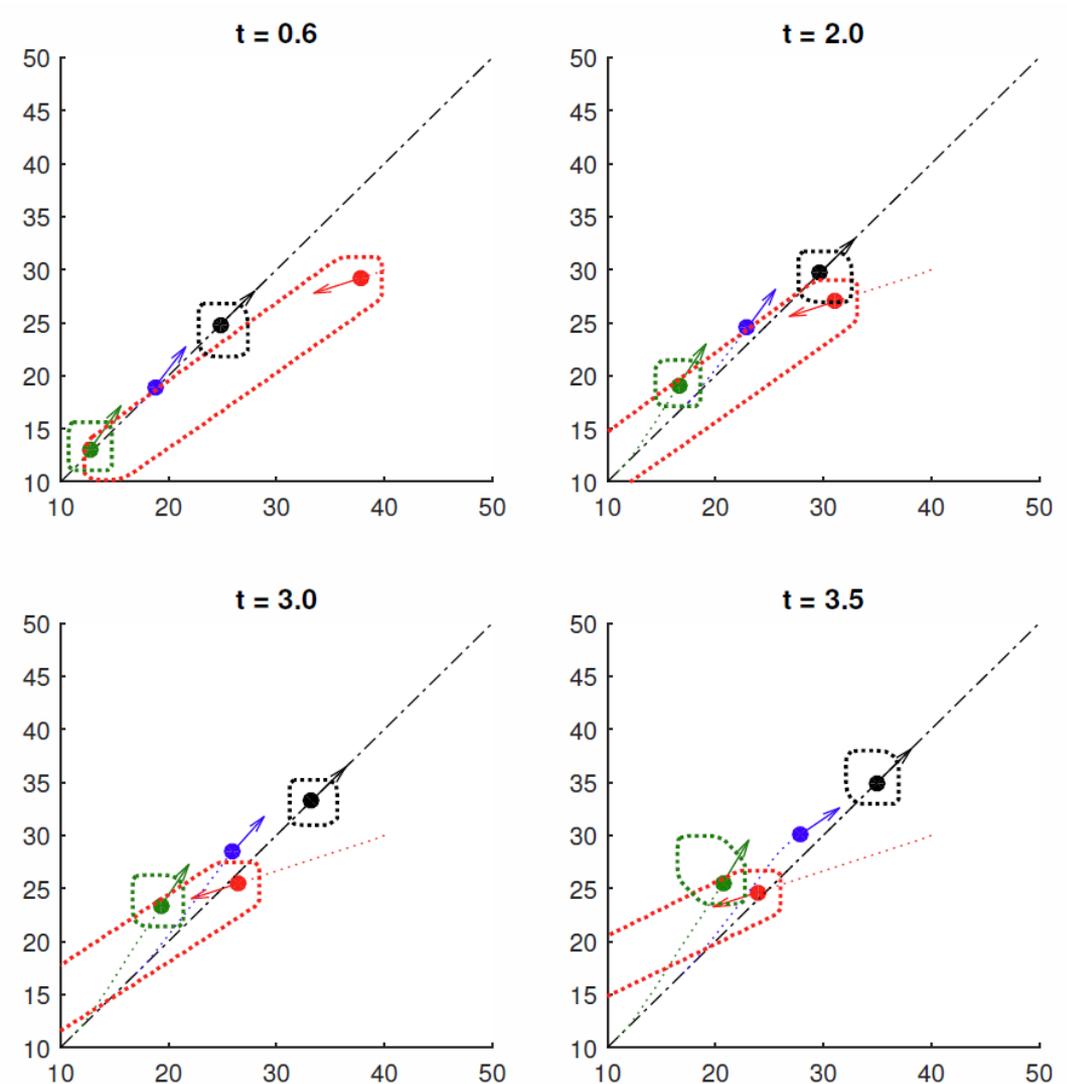
Intruder



Platoon responding to intruder (red vehicle)

Reachable sets for blue vehicle are shown

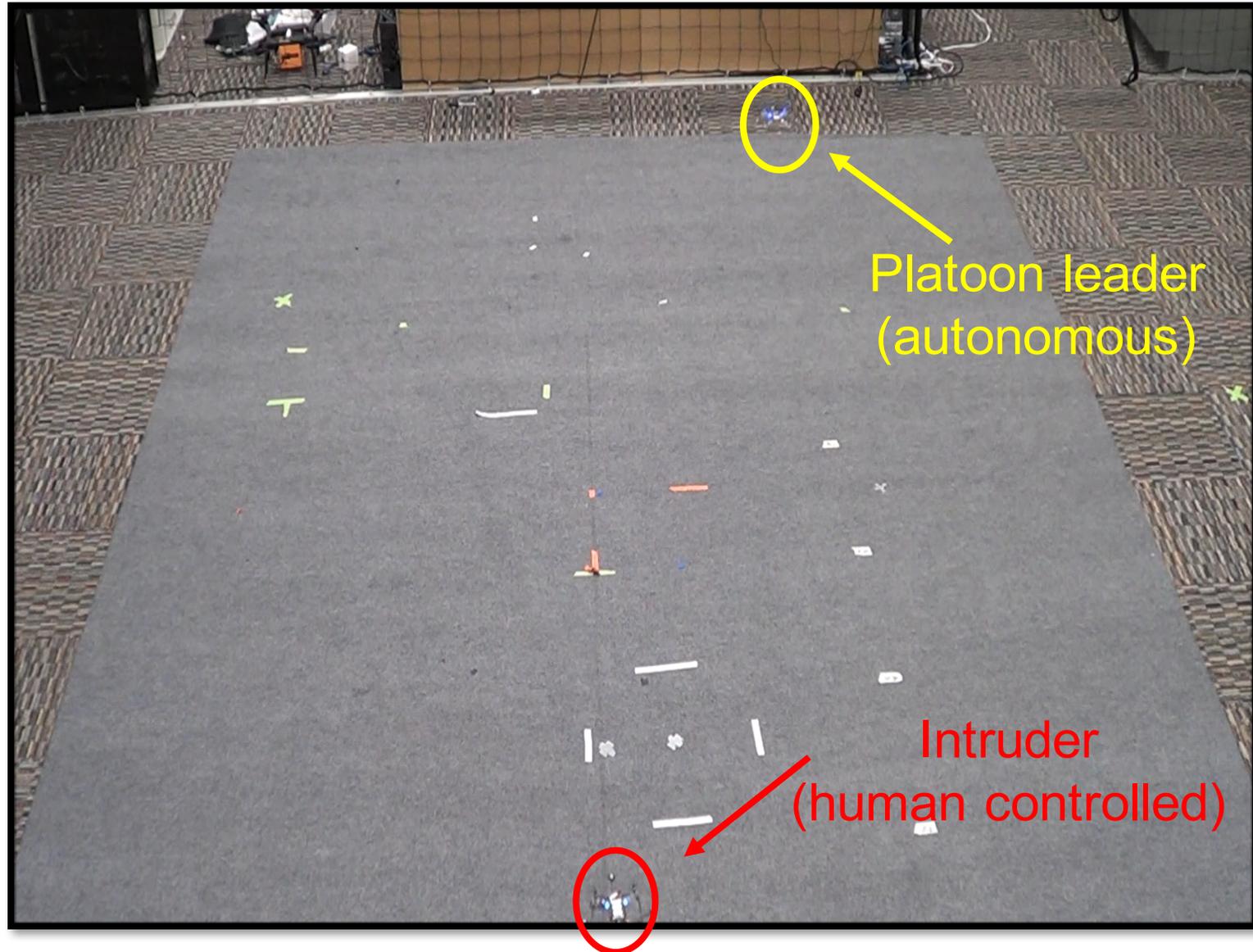
Blue vehicle must stay outside of all dotted boundaries



Experiments: Form Platoon



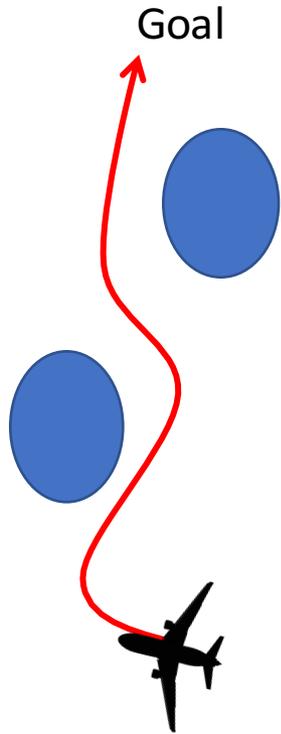
Experiments: Intruder Avoidance



Experiments: Change Highways

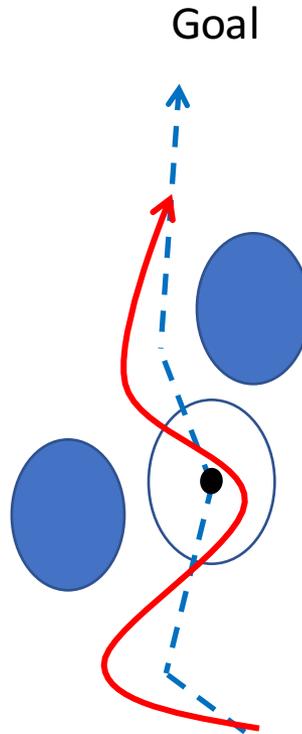


Slow and Accurate Planning



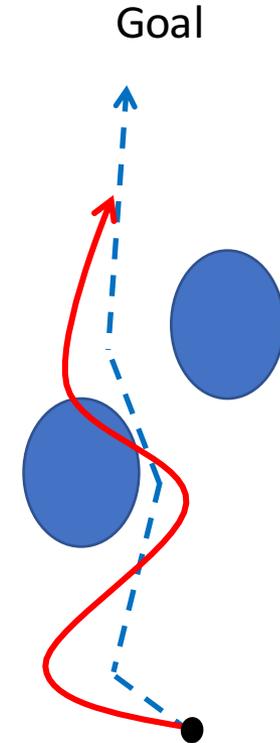
- Optimal control
- Guarantees on safety and goals
- Handles external disturbances (e.g. wind)
- Slow to compute

The Proposal



- Precompute a tracking error bound around the simple planning model
- Usable with lots of path/trajectory planners

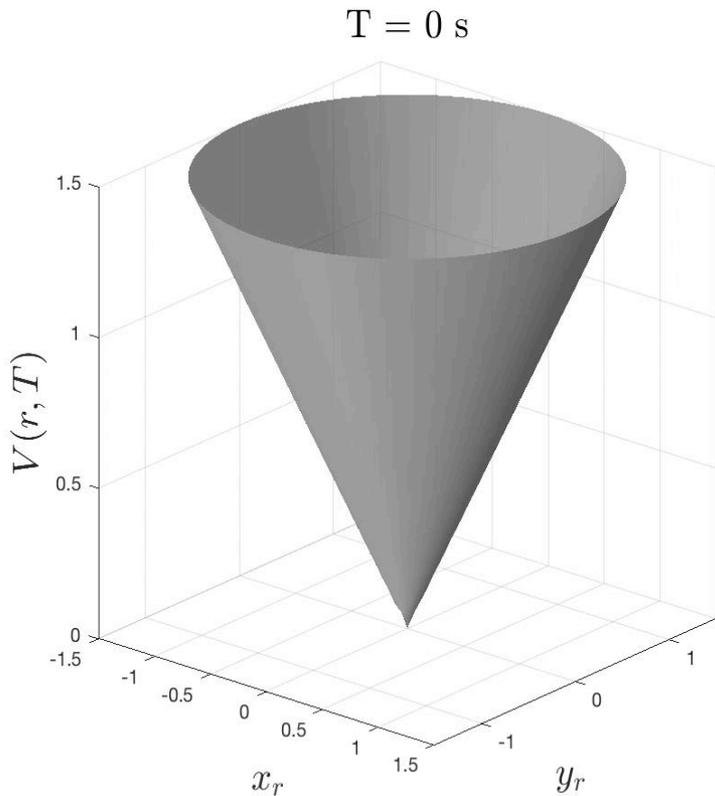
Fast (but less accurate) Planning



- Very fast with simple dynamics
- May not capture all system behavior
- Not necessarily robust to disturbances

Precomputed Tracking Bound

Goal: Map initial relative state to worst possible tracking error over time

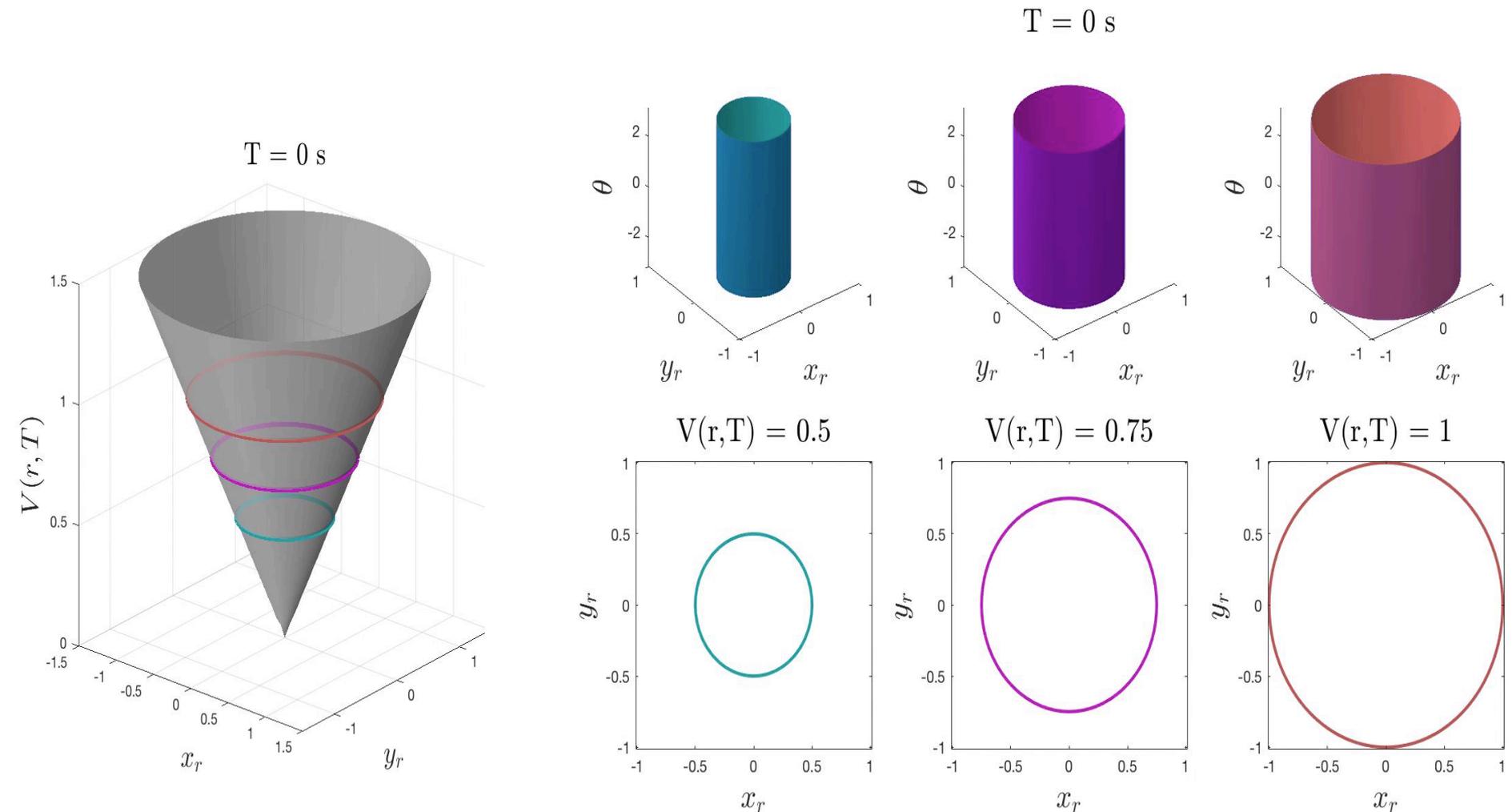


Planning system tries to **maximize** cost

Tracking system tries to **minimize** cost

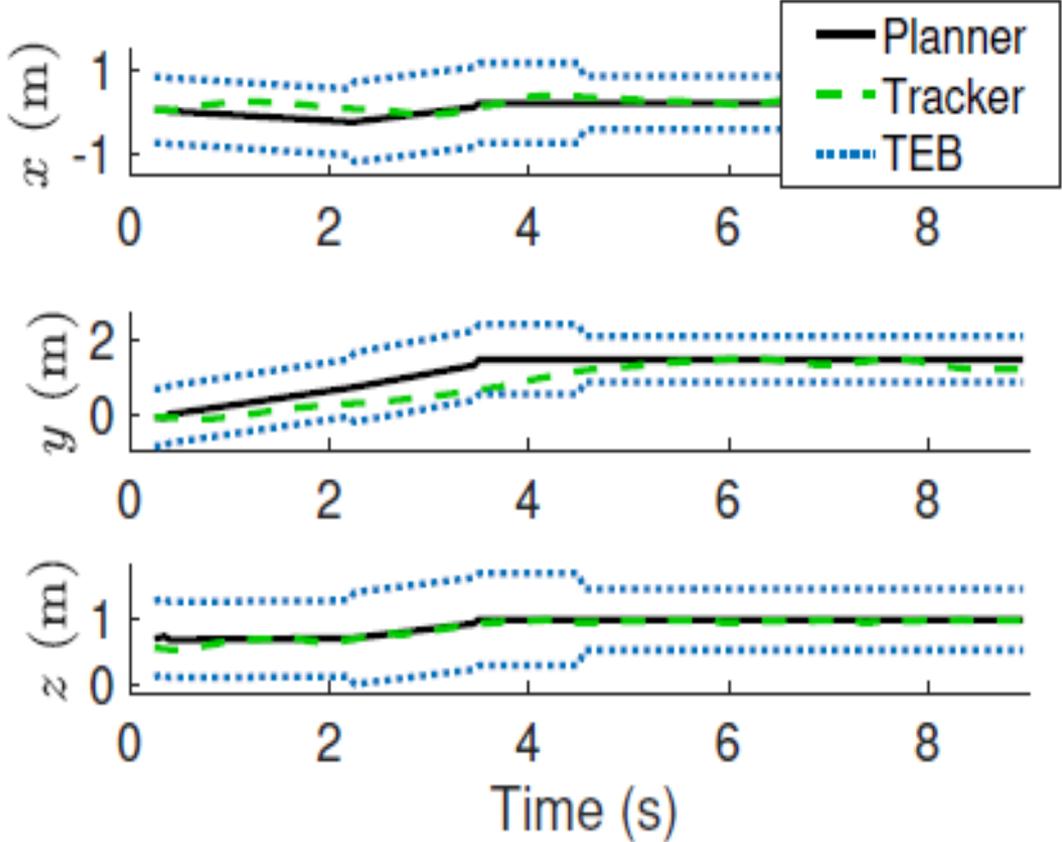
Capture **maximum** cost over time

Precomputed Tracking Bound



Smallest Invariant Level Set = Tracking Error Bound

Fast and Slow Planning



D. Fridovich-Keil*, S. Herbert*, J. Fisac, S. Deglurkar, and C. J. Tomlin, "Goal-Driven Dynamics Learning via Bayesian Optimization," 56th IEEE Conference on Decision and Control, Dec 2017.

Learn models from data...

... but stay safe while learning

- **Safety:**

- A nominal model with error bounds
- Reachable sets computed to ensure safety in worst case

- **Performance:**

- Use online learning to update model
- Cost function used to generate control action within the safe set

Safe Learning

with online model validation

Berkeley
University of California

New Vistas: Personal Air Mobility



[AeroVelo, Aurora, Vahana, Terrafugia...]

Thanks

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