

# Combining Formal and Probabilistic Modeling in Resilient Systems Design

Azad M. Madni

University of Southern California



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Azad M. Madni  
Michael Sievers  
University of Southern California

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# Outline

- 21<sup>st</sup> Century DoD Systems
- Engineered Resilience
- System Modeling
- Formal Probabilistic Approach
- Exemplar Problem
- Experimentation Testbed
- Initial Findings
- Summary

# 21<sup>st</sup> Century DoD Systems



- High complexity (hyper-connectivity, dependencies)
- Long-lived (> 20 years)
- Likely to be extended / adapted for over lifetime
- Stringent physical and cyber security
- Need dependability + adaptability + proactive error prevention
- Need to operate safely in dynamic, uncertain environments subject to disruptions
- To address these challenges, we need new models, methods and tools

# Engineered Resilience is a Difficult and Messy Problem...Why?



- **Requirements:** can be imprecise (especially initially)
- **Actions:** can be unclear (especially initially)
- **Environment:** can be unknown or partially known (partial observability, unknown hostile and/or deceptive actors)
- **System states:** can be ambiguous (uncertainty)

**These characteristics are incompatible  
with traditional, invariant modeling methods**

# Systems Modeling

- Primary means for engineering systems including resilient systems
- A fragmented area for engineering resilient systems
- Most serious problems result from the gap between requirements and models that need to reflect requirements
  - contribute to poor flow down of system requirements to software requirements
- Different aspects of system behavior represented by different models
  - need to harmonize different models

# System Modeling Requirements



- Verifiability (correctness)
- Flexibility (to adapt to changing conditions)
- Bidirectional reasoning support (resilience-related decisions)
- Scalability and extensibility (no. of agents, interconnections)
- Exploit partial information to generate value (not “data hungry”)
- Learn from new observations (system and environment states)

# Formal Probabilistic Approach



- Combines **formal** and **probabilistic** modeling with **heuristics**
  - enables flexibility (resilience)-verifiability (safety) tradeoffs
  - heuristics: help contain combinatorial explosion in state space
- Exploits **reinforcement learning**
  - learning of system states and environment with new evidence
  - system states model: Partially Observable Markov Decision Process
  - provides value even with partial information
- Employs a **layered architecture**
  - planning and decision making (top level) and control (bottom level)
  - decisions and information flow from top level to bottom level
  - execution constraints flow from bottom level to top level
  - **global objectives have precedence** over local goals (if conflict)
- Defines new construct: **Resilience Contract**
  - balance system **verifiability** and system **flexibility** requirements



# Exemplar Problem: Multi-UAV Operations



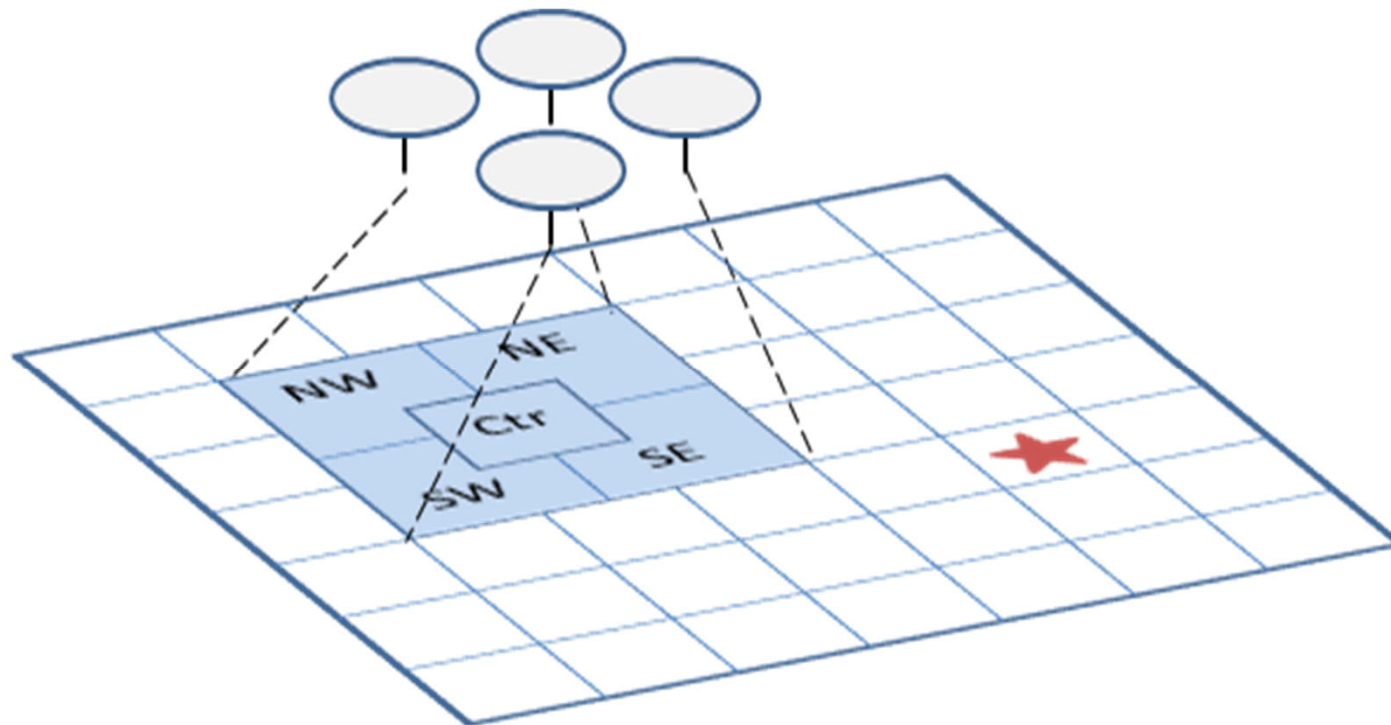
- UAVs used in missions in which environment is largely unknown and potential hostile with deceptive actors
- UAVs have to complete mission safely with original / descoped objectives
- UAVs can experience malfunctions and disruptions
- A mathematical model of environment is seldom available
- UAVs have collection assets to sense the environment
- UAVs can employ reinforcement learning to progressively learn the environment from sensed information
  - e.g., use RL to navigate through changing, partially observable environments

# What Resilience Means for Problem Domain



- **Operate safely** in dynamic, uncertain environments
  - tolerate / survive systemic faults and failures
  - adjust / adapt to environmental disruptions
  - protect / defend against physical and cyber threats
  - reconfigure / restructure to minimize impact of disruptions (e.g., security breaches, loss of sensing node or comm link)
- **Accomplish goals** with incomplete information
  - e.g., navigate safely to destination with partial observability

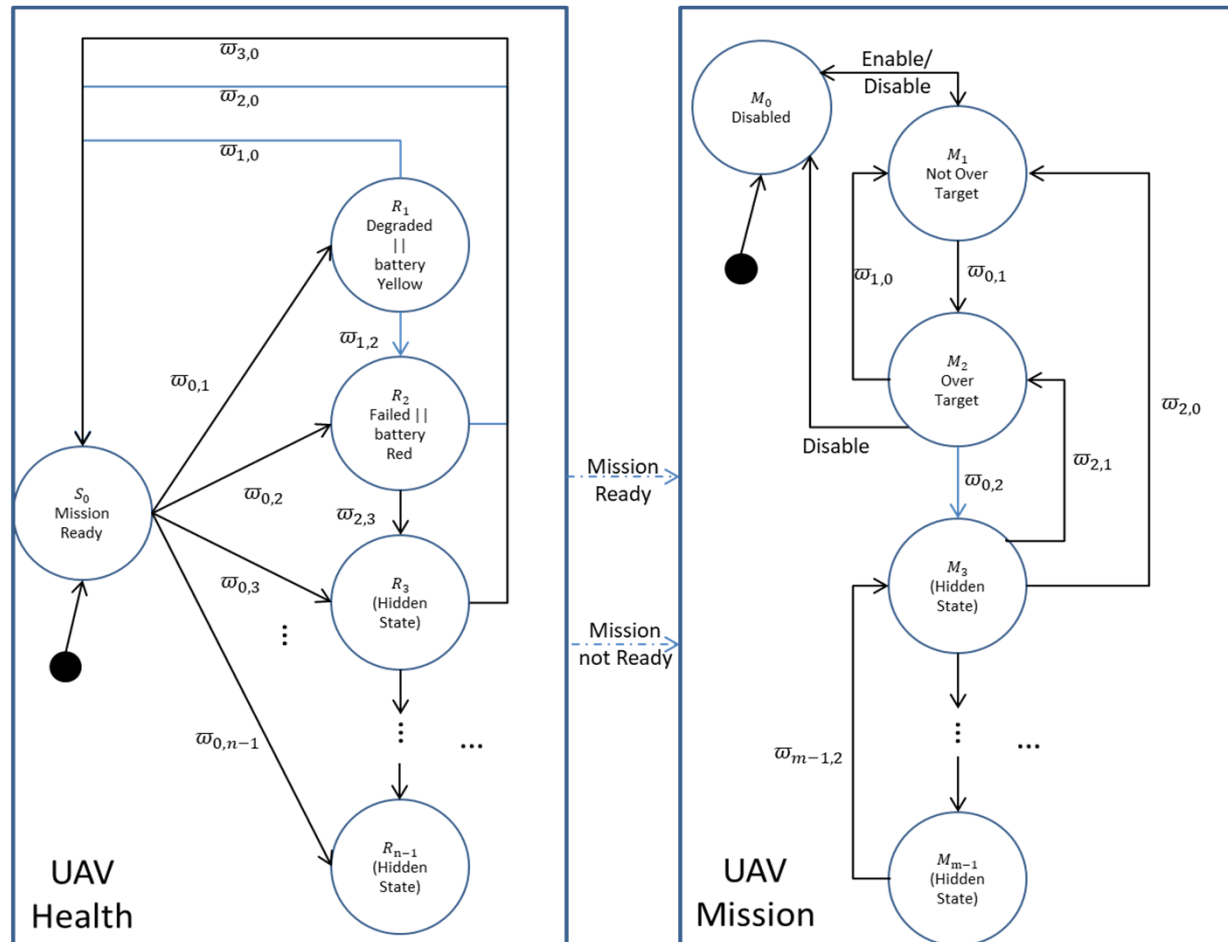
# UAV position relative to a recon target (red star) and FOV (blue)



# Exemplar Contracts

1.  $\neg \text{overTarget} \ \&\& \ \text{healthy} \ \&\& \ \text{batteryGreen} \rightarrow \text{move\_to\_target}$
2.  $\neg \text{batteryRed} \ \&\& \ \text{degraded} \ || \ \text{batteryYellow} \rightarrow \text{move\_to\_base}$
3.  $\text{batteryRed} \ || \ \text{failed} \rightarrow \text{land}$
4.  $\text{unknownHealth} \ || \ \text{unknownBattery} \rightarrow \text{move\_to\_base}$
5.  $\text{overTarget} \ \&\& \ \text{CTR} \ \&\& \ \text{healthy} \rightarrow \text{takeImages} \ \& \ \text{hover}$
6.  $\text{overTarget} \ \&\& \ \text{NW} \ \&\& \ \text{healthy} \rightarrow \text{takeImages} \ \& \ \text{move SE}$
7.  $\text{overTarget} \ \&\& \ \text{NE} \ \&\& \ \text{healthy} \rightarrow \text{takeImages} \ \& \ \text{move SW}$
8.  $\text{overTarget} \ \&\& \ \text{SW} \ \&\& \ \text{healthy} \rightarrow \text{takeImages} \ \& \ \text{move NE}$
9.  $\text{overTarget} \ \&\& \ \text{SE} \ \&\& \ \text{healthy} \rightarrow \text{takeImages} \ \& \ \text{move NW}$

# Simplified POMDP: Health and Mission Models



# Multi-UAV Dashboard Prototype



## ■ Capabilities

- customizable dashboard for monitoring and control of simulated or physical vehicles

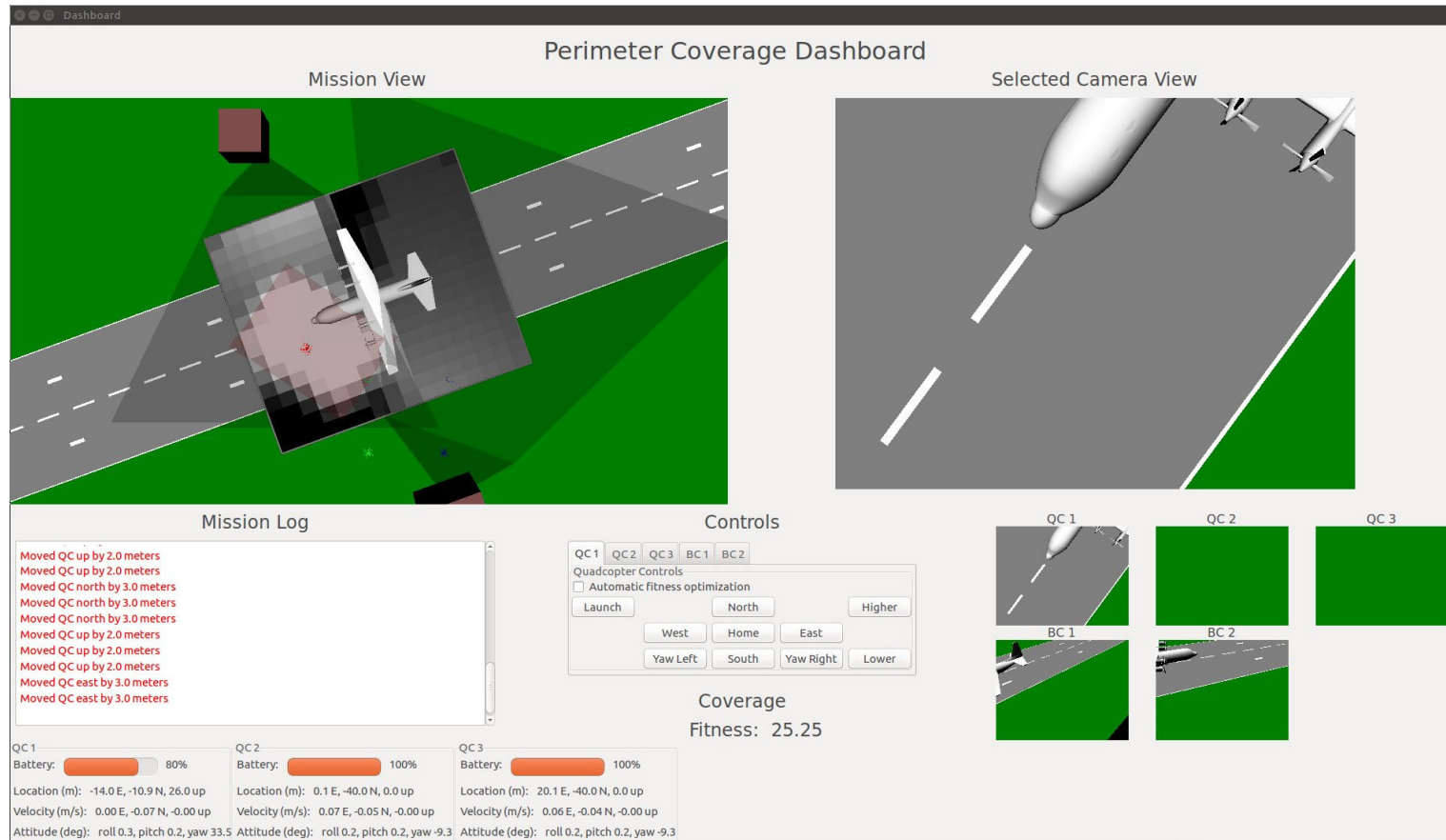
## ■ Underlying technologies

- dronekit platform with visualization facilities
- quadcopters (hardware) and quadcopter simulation models
- quadcopter planning and decision-making model
- quadcopter controller

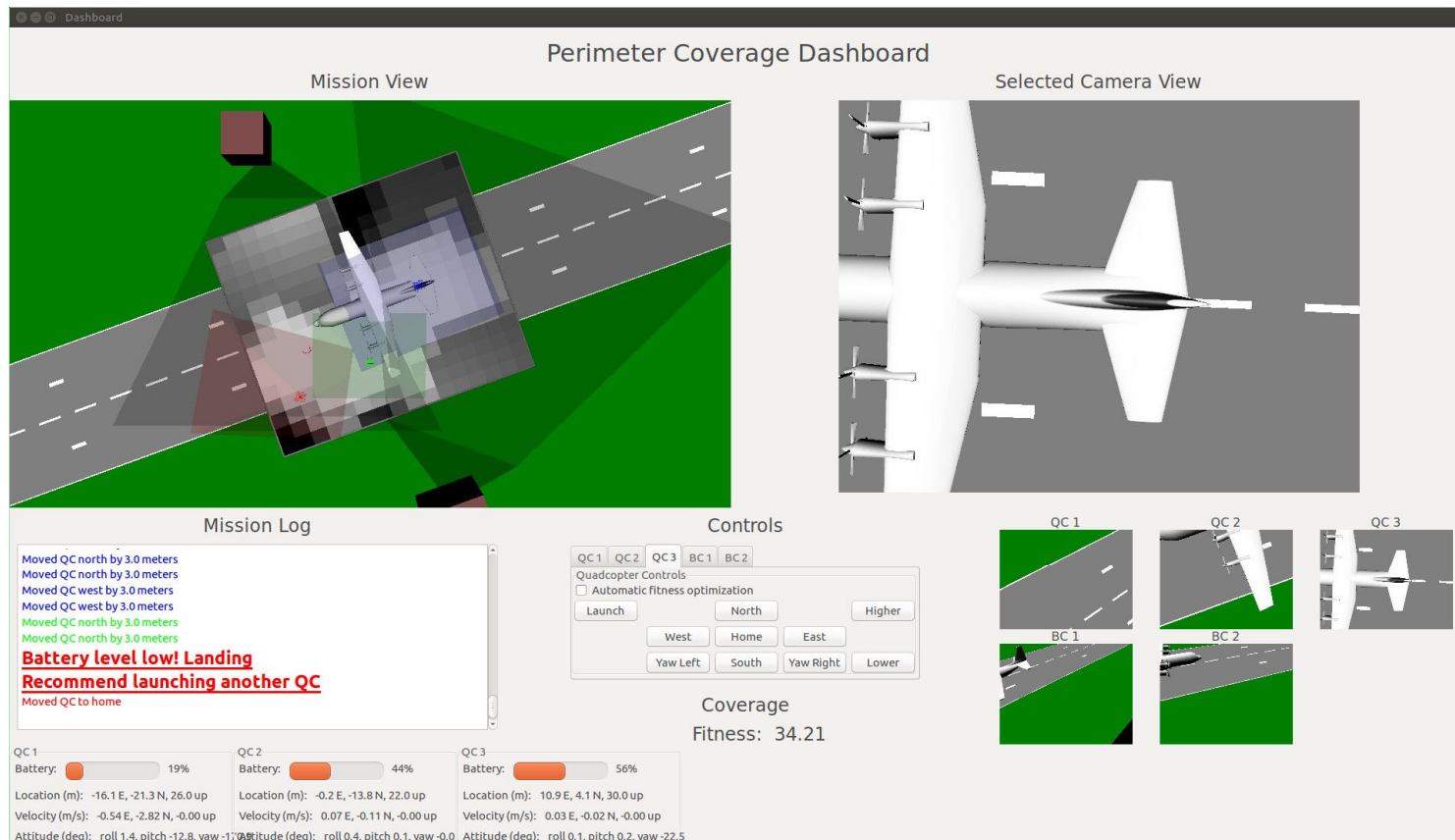
## ■ Key capabilities

- simulated vehicles exhibit behavior of physical vehicle
- same commands used to control vehicle models and physical vehicles (quadcopters)
- can switch from simulated vehicles to physical vehicles

# Dashboard Showing Camera View of Flying Quadcopter

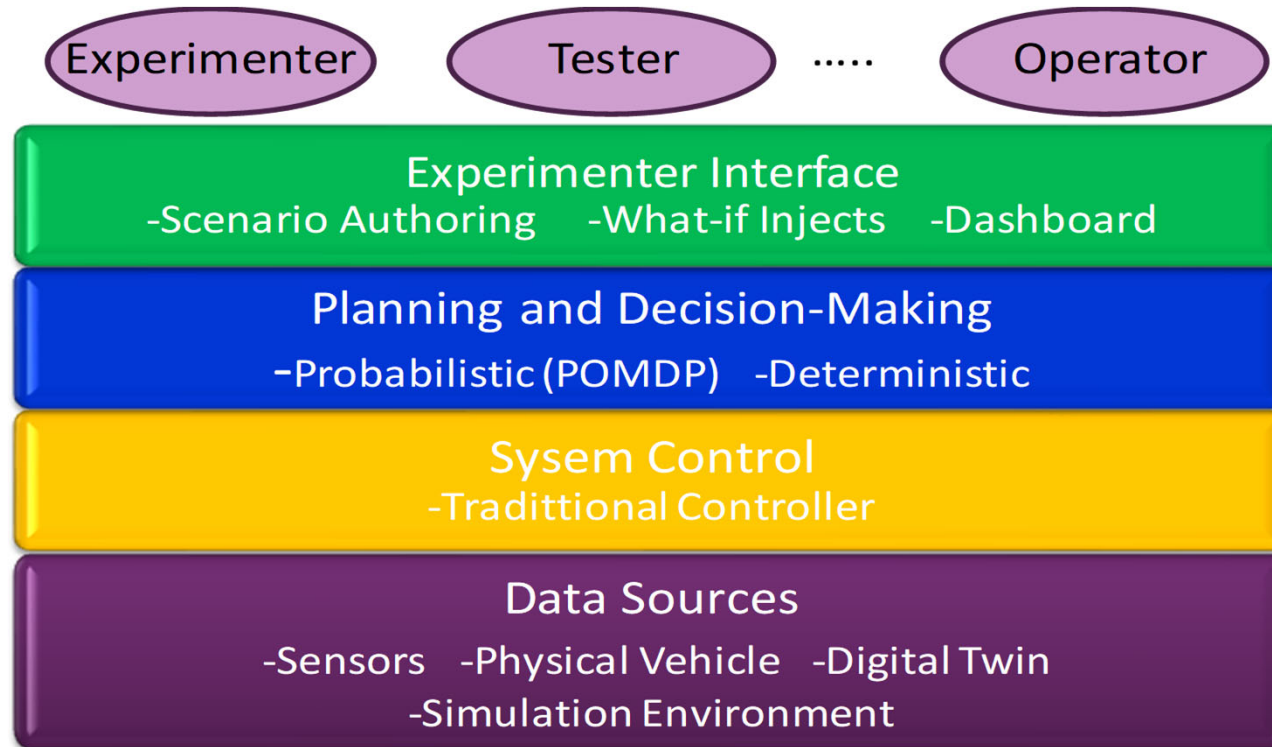


# Dashboard Showing 3 Flying QCs With One Low on Battery and Landing





# Experimentation Testbed Architecture



# Prototype Testbed

## ■ Quadcopters

- driven by Raspberry Pi and Navio Flight Controller
- full IMU: 3-axis accelerometers, rate gyros, magnetometer
- inputs from laptop and/or remote controller
  - control values (throttle, roll-pitch-yaw)
  - perform autonomous flight

## ■ Instrumented Testbed

- layered architecture (UI, planning and DM, control, data sources)
- customized Python scripts for vehicle control
  - dronekit framework and commands
- semi-autonomous flights
  - launch, take-off, hover, and perform limited waypoint navigation
- context-sensitive monitoring and control dashboard
  - monitor vehicle status and control vehicle
  - communicate with simulated vehicles and physical system

# Prototype Testbed Hardware



# Findings To-Date

- Key problem in implementing hybrid models
  - resolving mismatch between planning & decision-making layer and vehicle control layer
- Mismatch resolution
  - ensure that propagated commands from PDM layer to controller do not violate physical and regulatory constraints
  - propagate execution constraints from control layer to PDM layer for PDM layer to take into account when issuing commands
  - incorporate heuristics (e.g., priorities, region of influence) to resolve conflicts and simplify computation

# Findings To-Date (cont'd)



- POMDP and vehicle controller work on different time scales
  - dynamics model runs every 0.01 seconds (accuracy)
  - POMDP runs slower (high level decisions/commands)
    - waypoint navigation problem with goal of minimizing response time to action
    - ideal sampling period for POMDP determined experimentally
- Simultaneous creation of prototype and testbed - good strategy
  - introduced rigor in experimentation
  - compatible with introducing Digital Twin
  - currently: able to switch between simulation model and physical system
  - future: incorporate operational data from physical twin into Digital Twin
- Monitoring and execution dashboard – a key capability
  - facilitated understanding and debugging of vehicle behaviors



# Summary

- 21<sup>st</sup> Century Systems need to be safe, resilient and affordable
- Have to operate in uncertain, hostile and deceptive environment with partial observability
- System model verifiability is needed for system safety
- System model flexibility is needed for resilient behavior
- Such capabilities beyond traditional systems modeling capabilities
- Resilience Contract, a probabilistic approach based on POMDP, when coupled with heuristics and reinforcement learning, can satisfy safety, resilience and improved performance needs
- Our research is demonstrating viability of this approach with different CONOPS and resilience mechanisms

**Thank You**