

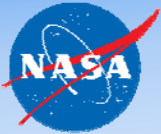
# NATO IST-136



## Trusted Autonomy

### Autonomy and Autonomous Systems at NASA LaRC's Autonomy Incubator

B. Danette Allen, PhD  
Head, Autonomy Incubator  
NASA Langley Research Center (LaRC)  
21 March 2016



# Three Goals & Many Challenges



## 1. Build a Multi-Disciplinary Team

- Mechanics/Electronics/
- Controls
- Computer Science/Programming
- Psychology/Human Factors
- **Machine Learning**
- Signal Processing/Computer Vision

## 2. Enable New Missions

- Space
- Aeronautics
- Science



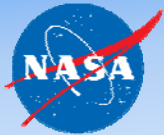
## 3. Create a Testbed for Autonomous Systems

- Open Software Architecture (AEON)
  - Data Distribution Service (DDS)
- Langley Autonomy & Robotics Center
- CERTAIN



## Autonomy Challenges

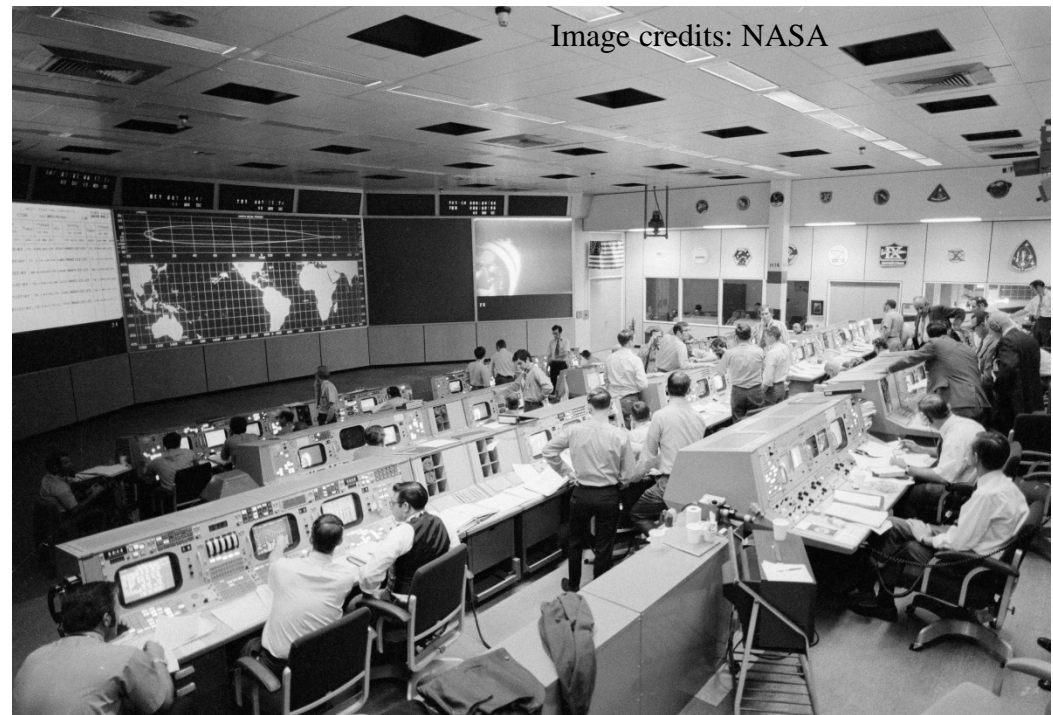
- **Human-Machine Interaction**
- **Data-rich/degraded/deprived environments**
- **Size, Weight And Power (SWAP)**
- **Sensor Fusion**
- **Adaptive Control**
- **Geo-containment**
- **Sense/Detect and Avoid (DAA)**
- **Precision navigation**
- **Localization**
- **Adaptation and Learning**
- **Performance Standards**
- **Verification and Validation (V&V)**
- **Certification/Trust**
- **Test and Evaluation (T&E)**



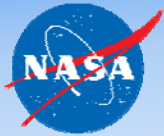
# NASA's Manned Missions



- Historic and current ATM and space exploration paradigms are human-centric. Humans are aided by automation to make intelligent decisions and intervene as needed, especially in off-nominal situations.



Five of the seven Apollo missions that attempted to land on the Moon required real-time communications with controllers to succeed.



# NASA's Unmanned Missions

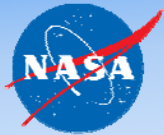


- Historic and current ATM and space exploration paradigms are human-centric. Humans are aided by automation to make intelligent decisions and intervene as needed, especially in off-nominal situations.



Things have changed but...  
humans are still hovering around monitors waiting to intervene.

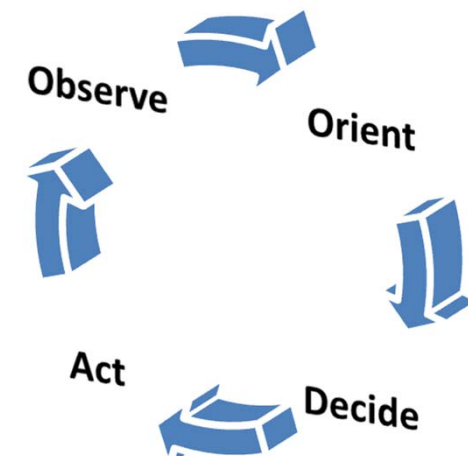


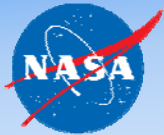


# NASA's Missions and Autonomy



- **Human intelligence** applied to supervision, control, and intervention of operations will **no longer** be **viable** due to system/mission complexity, short reaction/decision time, communication delays, distance, or hostile environments.
- Systems with **machine intelligence**: capable of responding to expected and unexpected situations:
  - trusted and certified-safe systems capable of
  - sensing and perception
  - situation assessment/awareness
  - decision-making
  - taking action
  - teaming with humans
  - and knowledge acquisition (**learning**)

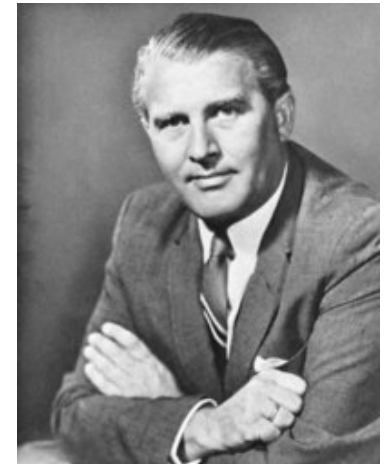




# Human vs. Machine Intelligence

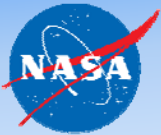


- “The best computer is a man, and it’s the only one that can be mass-produced by unskilled labor.” — Wernher von Braun



- Apollo 13 Control Room
- Gene Kranz “in a box”?

Image credits: NASA



# Three Goals & Many Challenges



## 1. Build a Multi-Disciplinary Team

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## 2. Enable New Missions

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- Aeronautics
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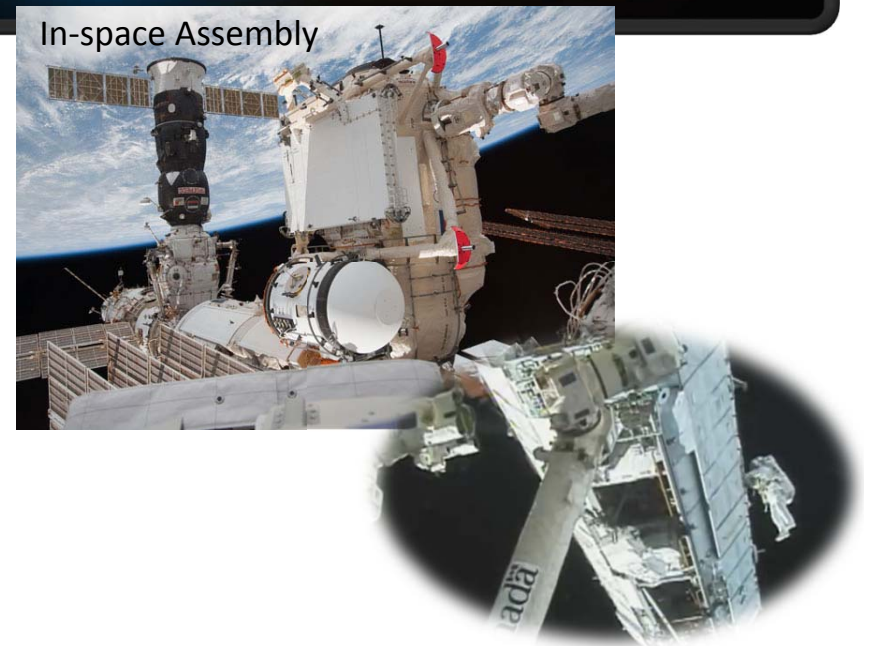
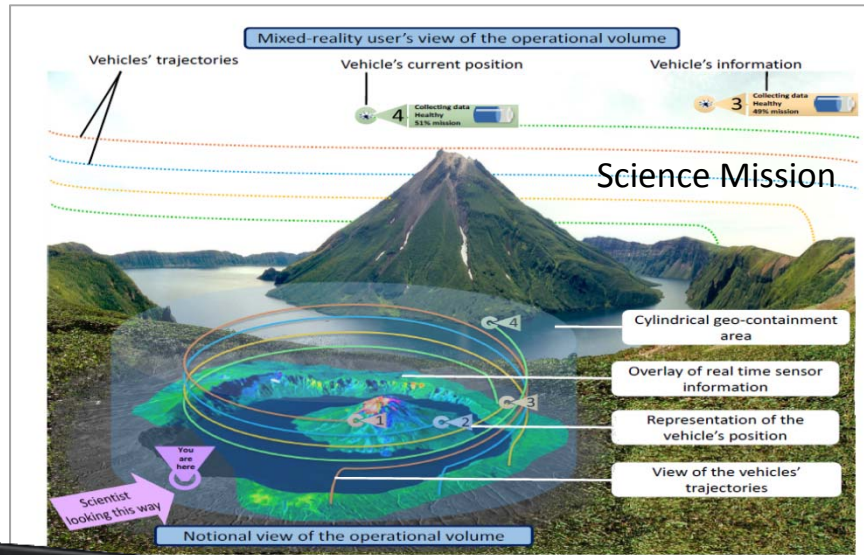
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## Autonomy Challenges

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- Data-rich/degraded/deprived environments
- Size, Weight And Power (SWAP)
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- Geo-containment
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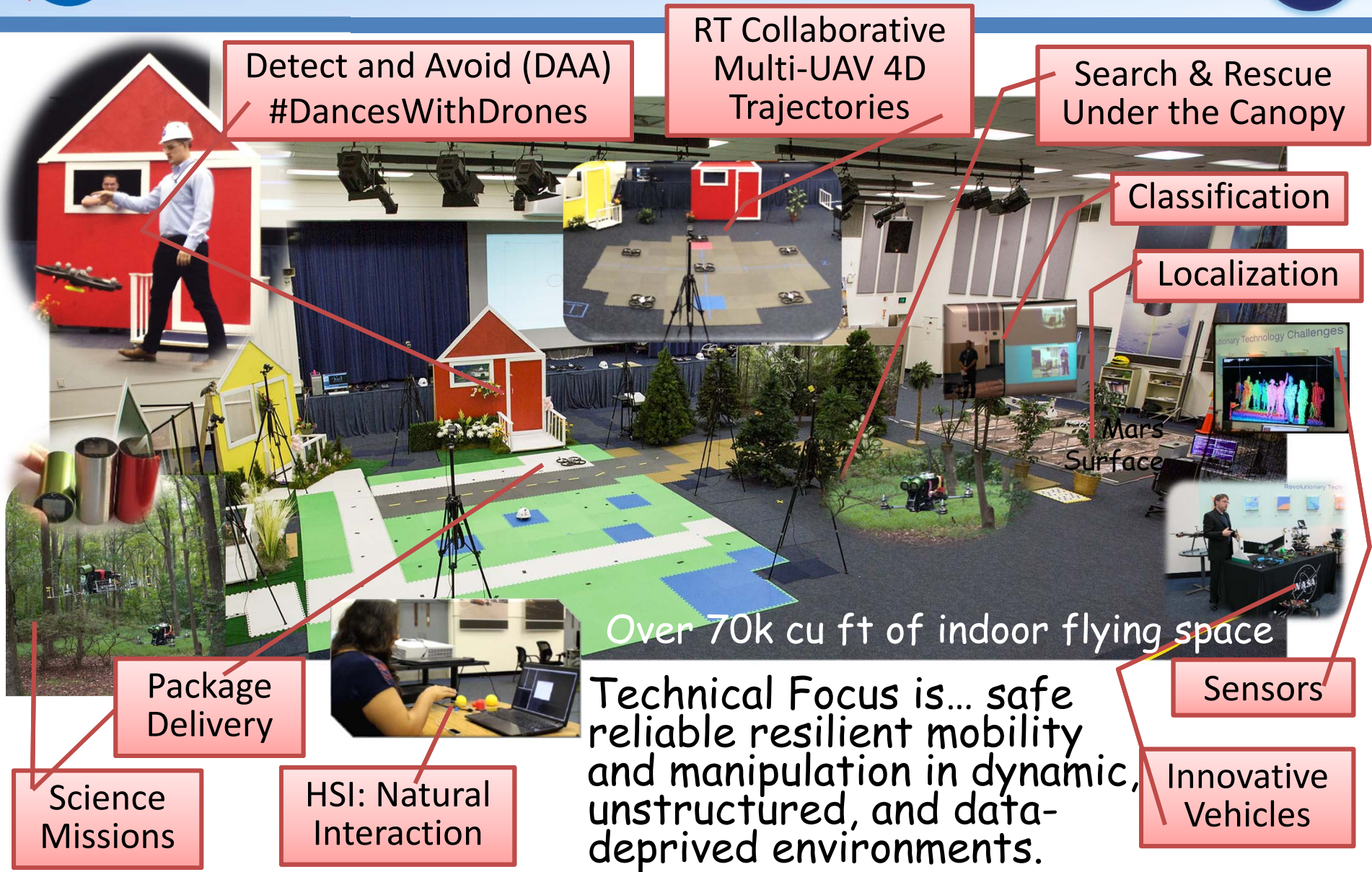


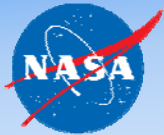




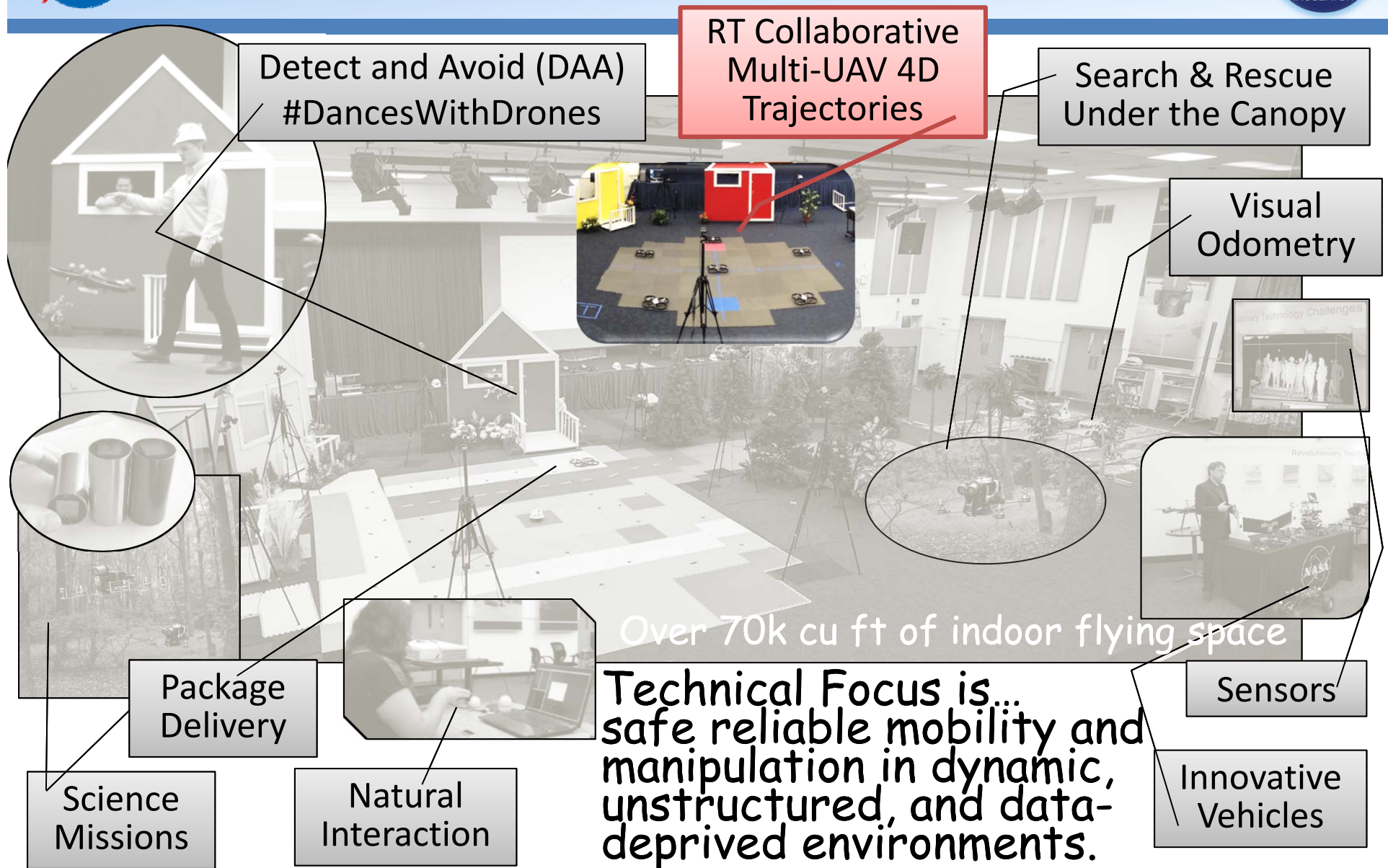


# Autonomy Incubator R&D Overview

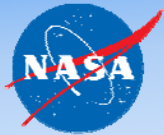




# Autonomy Incubator R&D





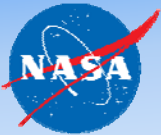


# Multi-Agent Collaborative Trajectories



Working with UIUC





# Trajectory Generation Algorithm



- Trajectory generation algorithm
  - Autonomous generation for  $n$  heterogeneous vehicles with Quintic Bézier Hodographs

- Optimize trajectories

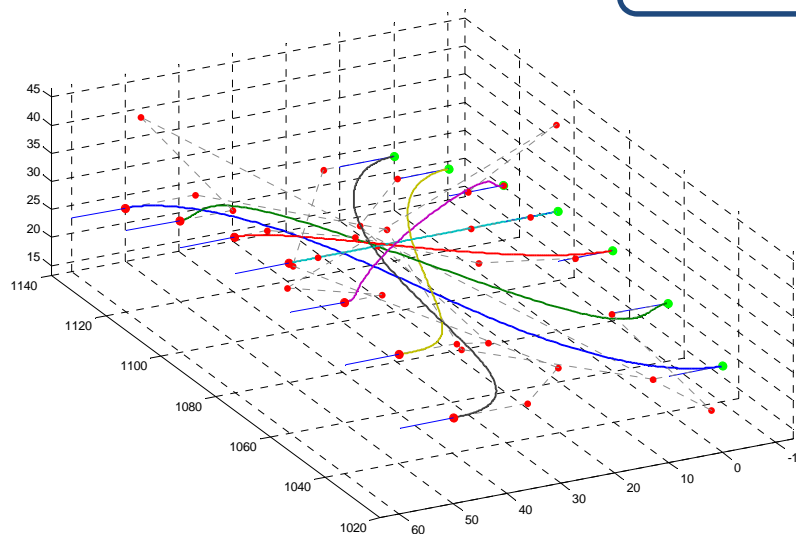
$$\min_{\Theta_i \times \dots \times \Theta_N} J(\cdot)$$

$J(\cdot)$  : cost function

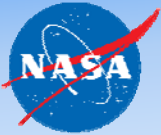
- Enforce constraints

- Safety inter-vehicle distance
- Speed
- Acceleration
- Flight path-angle
- Turning rates

Customizable to vehicle-specific dynamic constraints



- UAVs coordinated along their trajectories
- Vehicles communicate over a time-varying network, vehicles are aware of each other
- Makes the system robust external disturbances



# Trajectory Generation



- Library of trajectories for gesture recognition

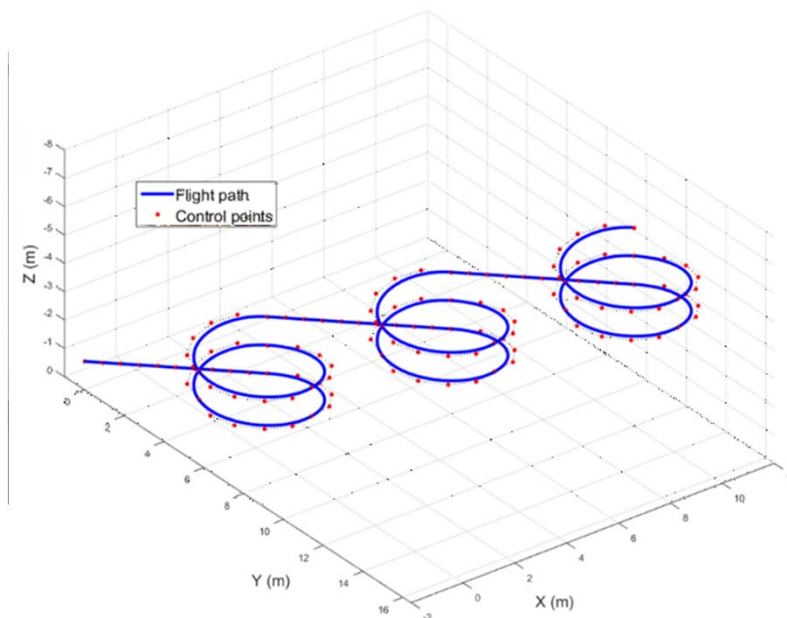
- Integrated with gesture recognition

- Using Bézier curves

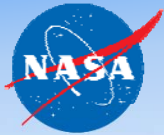
$$p(\zeta) = \sum_{i=0}^n \bar{p}_i \binom{n}{i} (1-\zeta)^{n-i} \zeta^i$$

$\bar{p}_i$  : control points

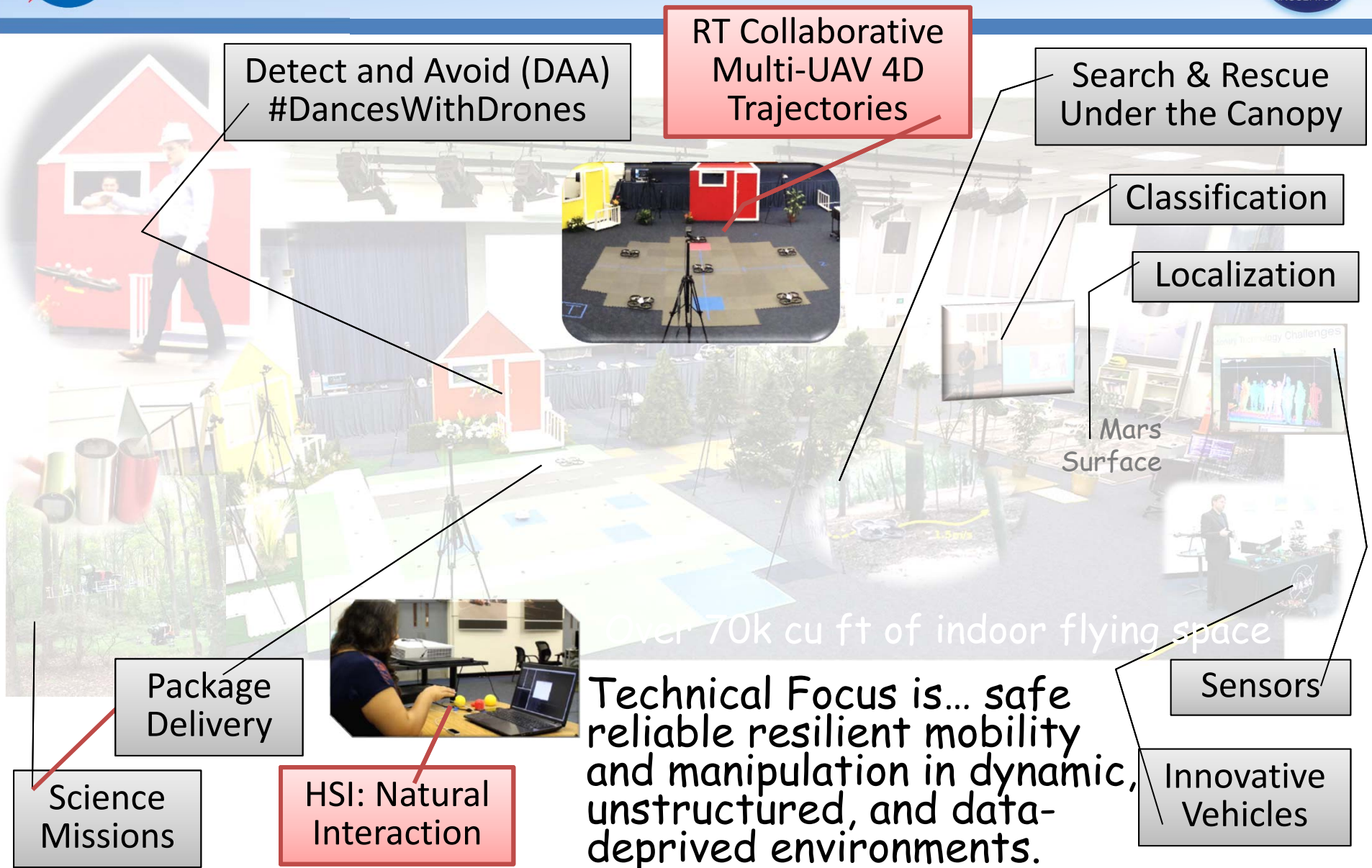
$\zeta \in [0, 1]$  : curve parameter



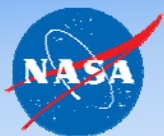
Example application: Design a set of trajectories to perform air-sampling missions



# Autonomy Incubator R&D Overview



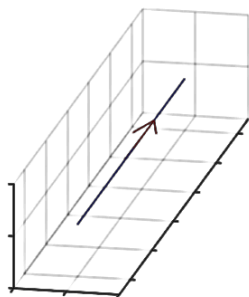




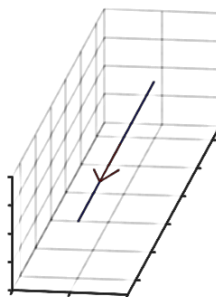
# Gesture Library



**Forward**



**Backward**



**Right**



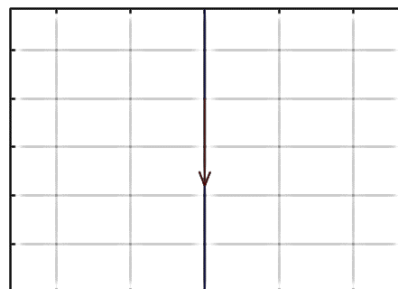
**Left**



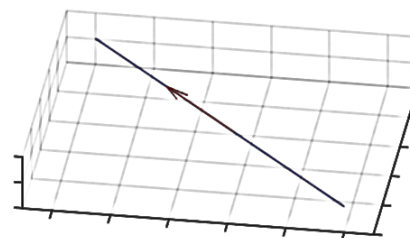
**Up**



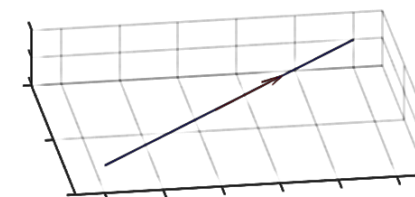
**Down**



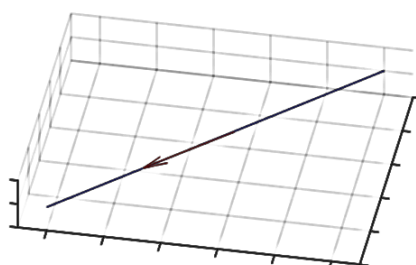
**Forward-Left**



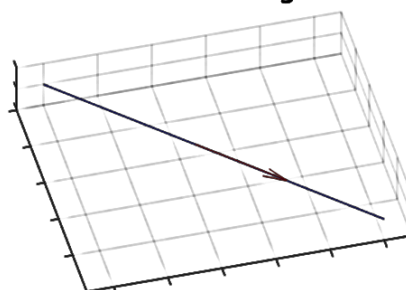
**Forward-Right**



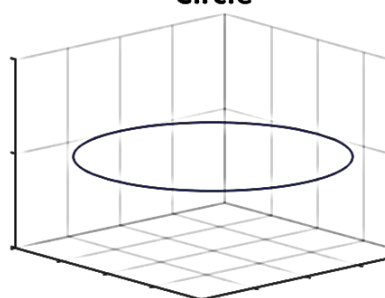
**Backward-Left**



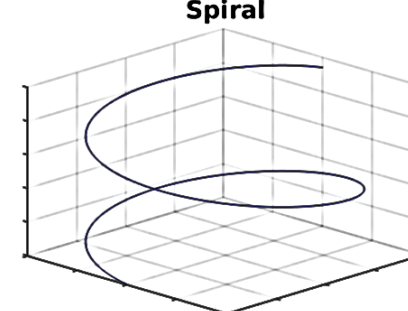
**Backward-Right**

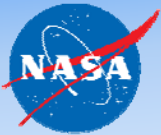


**Circle**

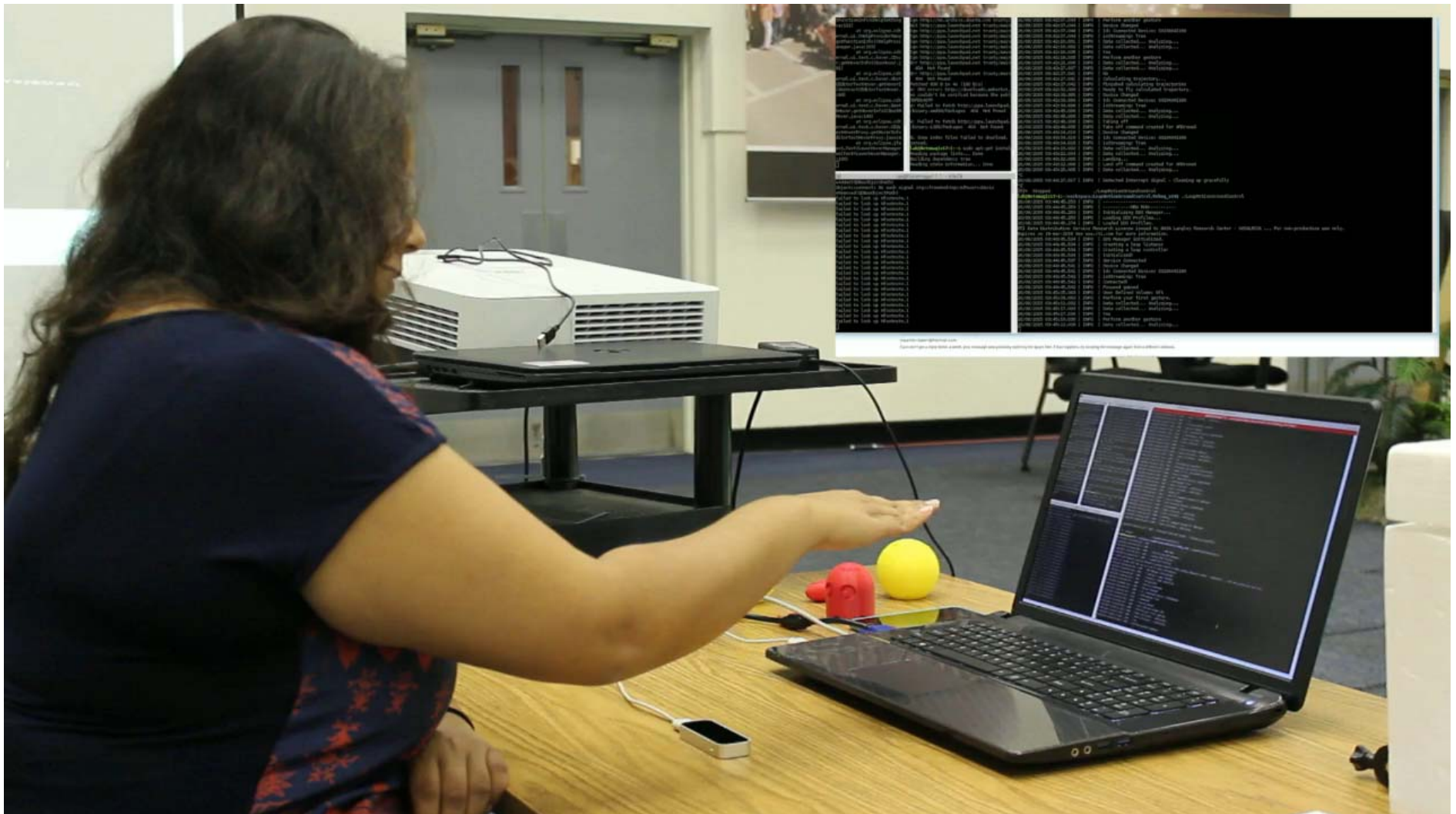


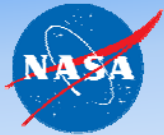
**Spiral**



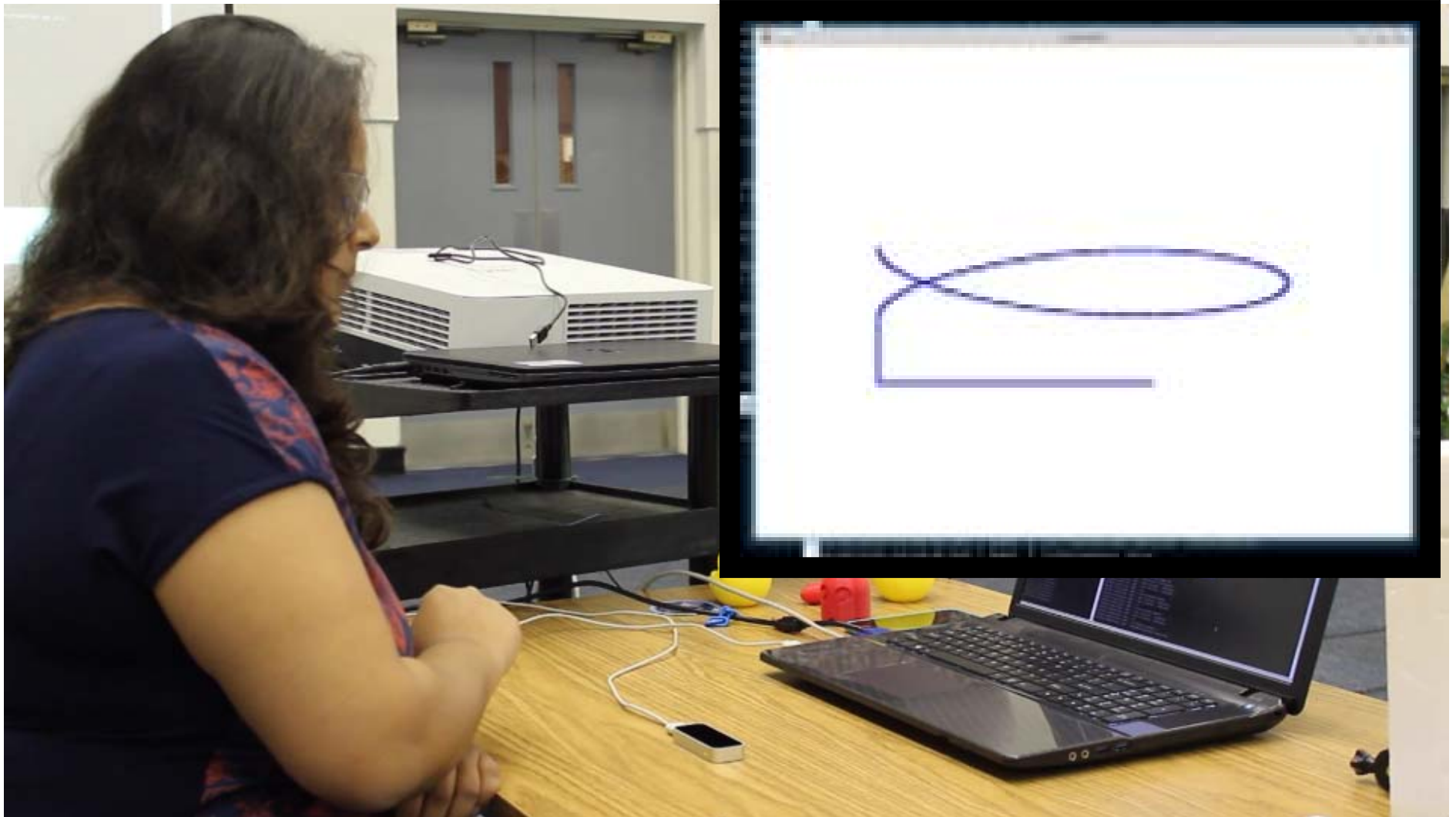


# Human Machine Teaming

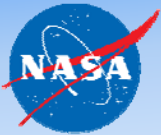




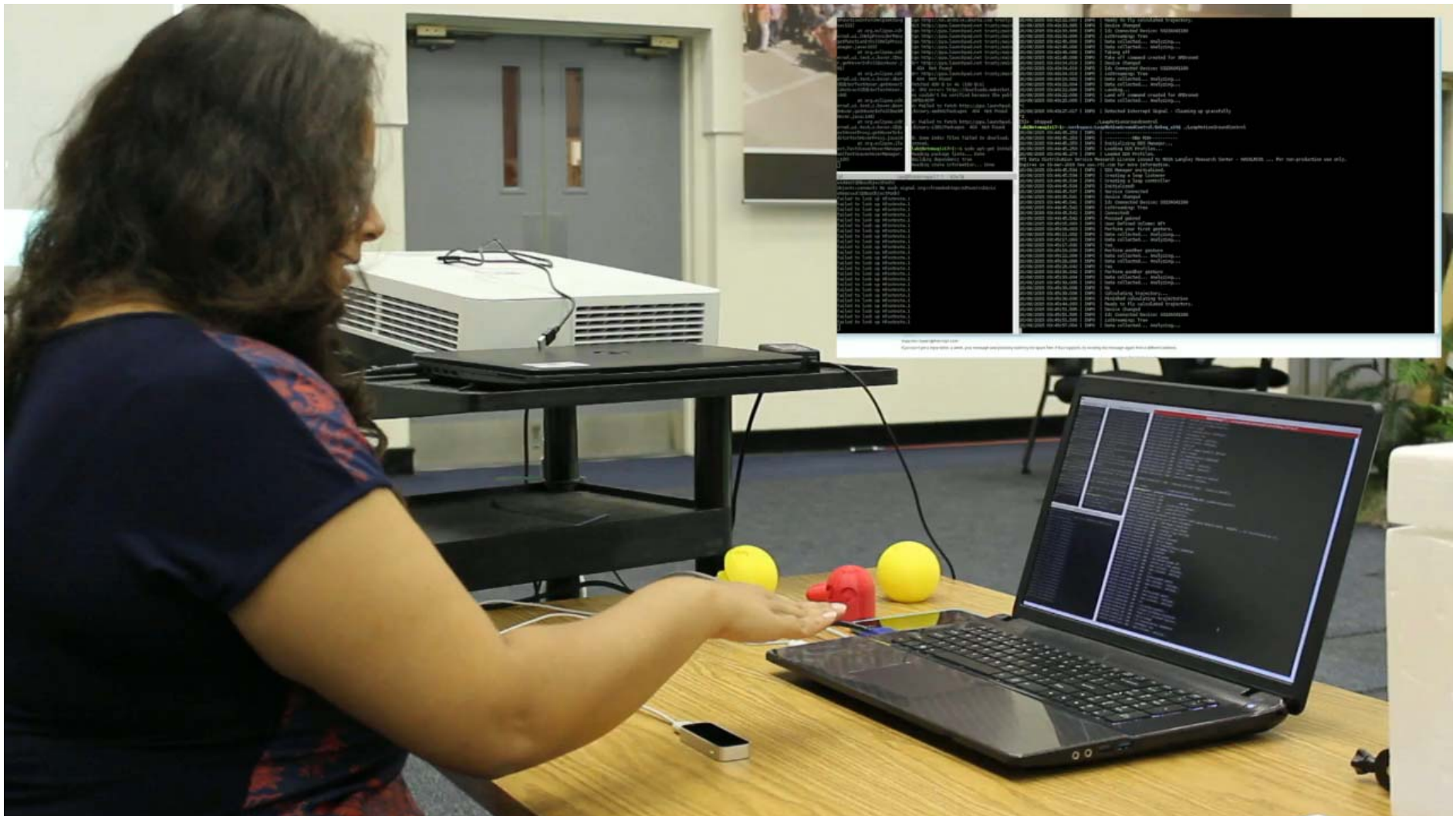
# Human Machine Teaming

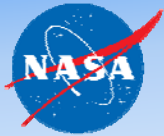






# Human Machine Teaming



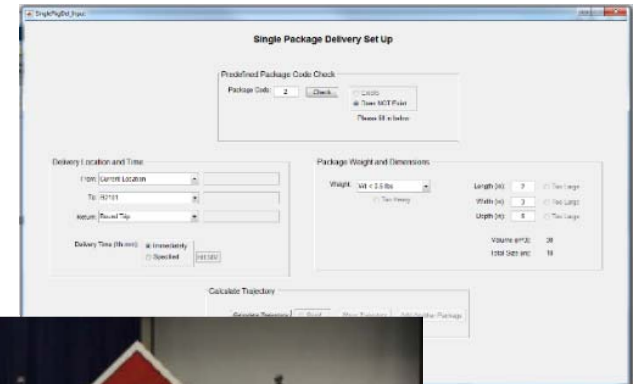


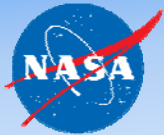
# Voice to Define and Control Missions



Natural method of communication for humans

- Specify package delivery location
- High level vehicle control
  - Takeoff and land
- Using CMU Sphinx4 software package
  - Using a controlled language
  - Not trained to speaker
- Define key words used in verbal descriptions
  - Prof. Nick Roy at MIT
  - Ph.D. student Naomi Schurr



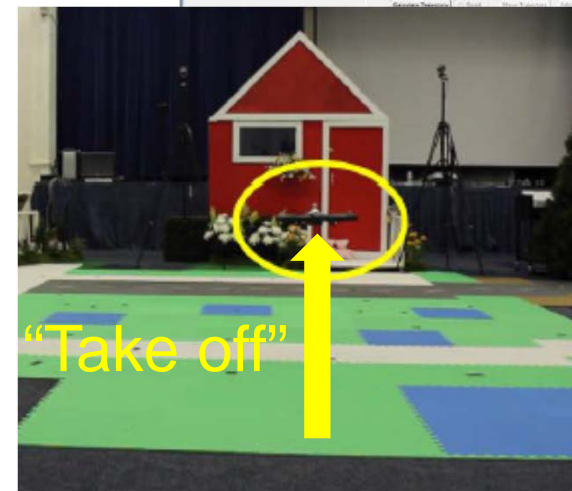
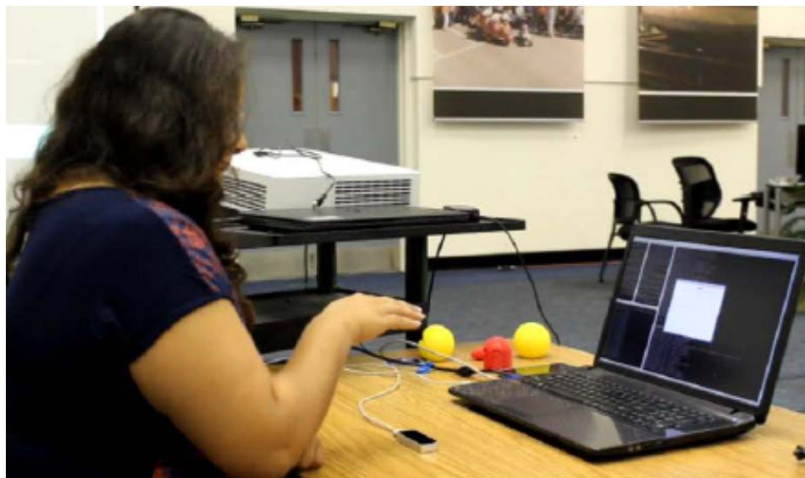
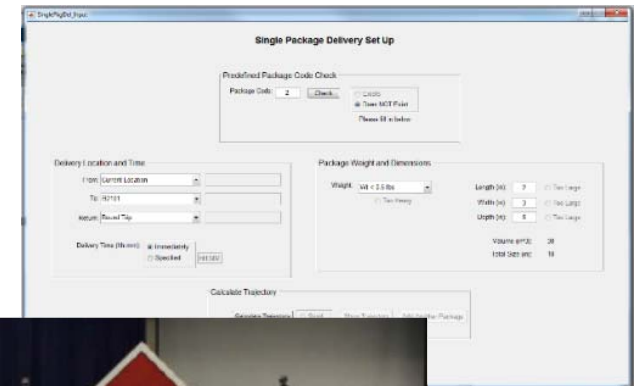


# Combine Communication Modalities



“Put this over there”

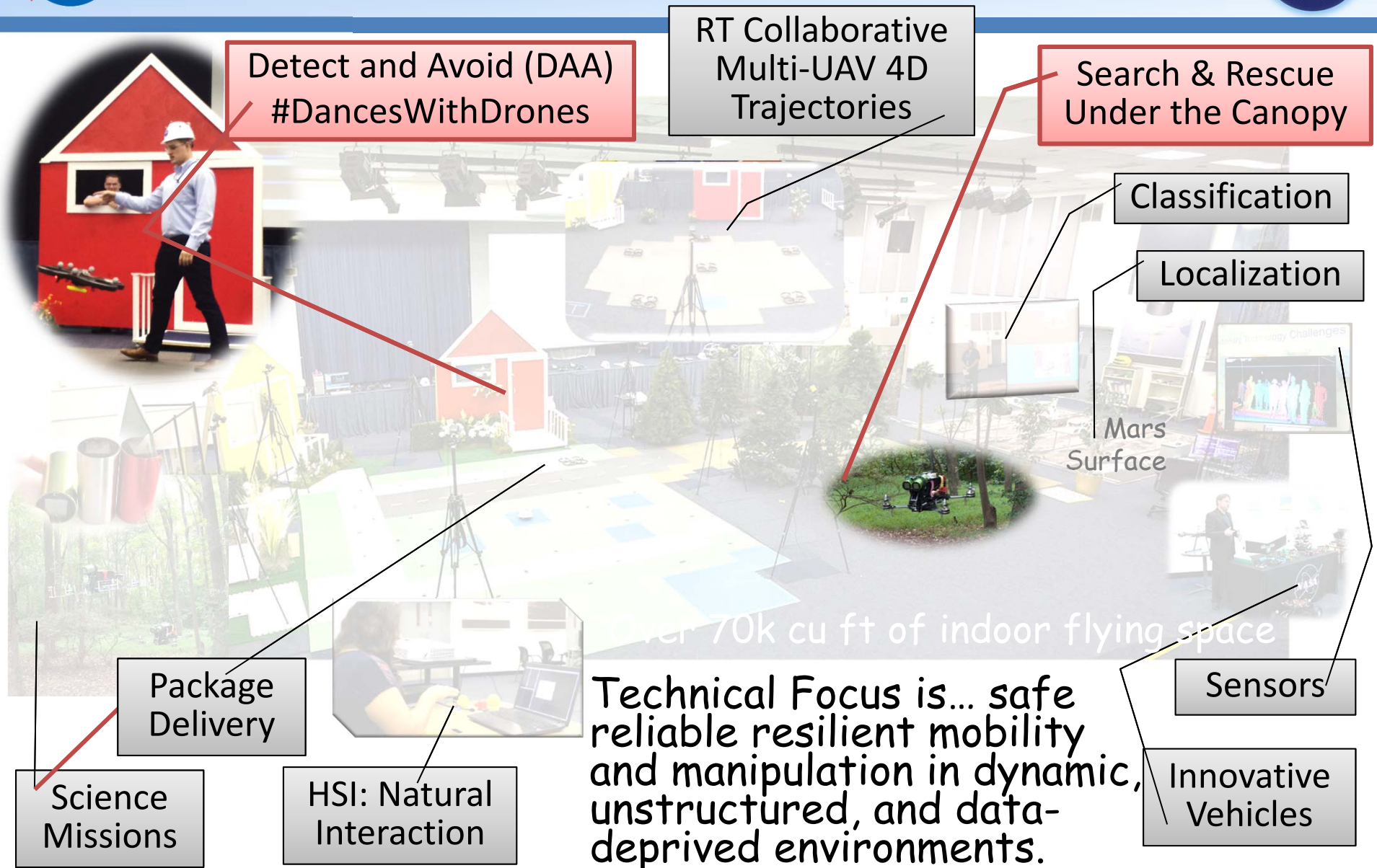
- Understand from context
  - Communicating Commander’s intent
- How team members normally communicate
- Facilitates human-machine teaming

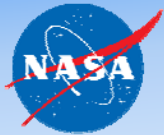






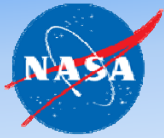
# Autonomy Incubator R&D Overview





# Object Avoidance / Path (re)planning (#DancesWithDrones)





# Obstacle Avoidance under Tree Canopy

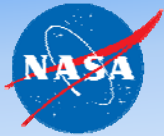


## Pilot Correction Algorithm

- Operator trains avoidance model
- UAV autonomously navigates through cluttered environment



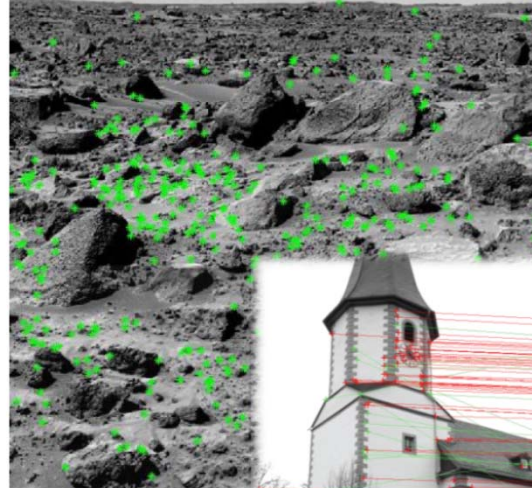




# Data-Deprived Navigation



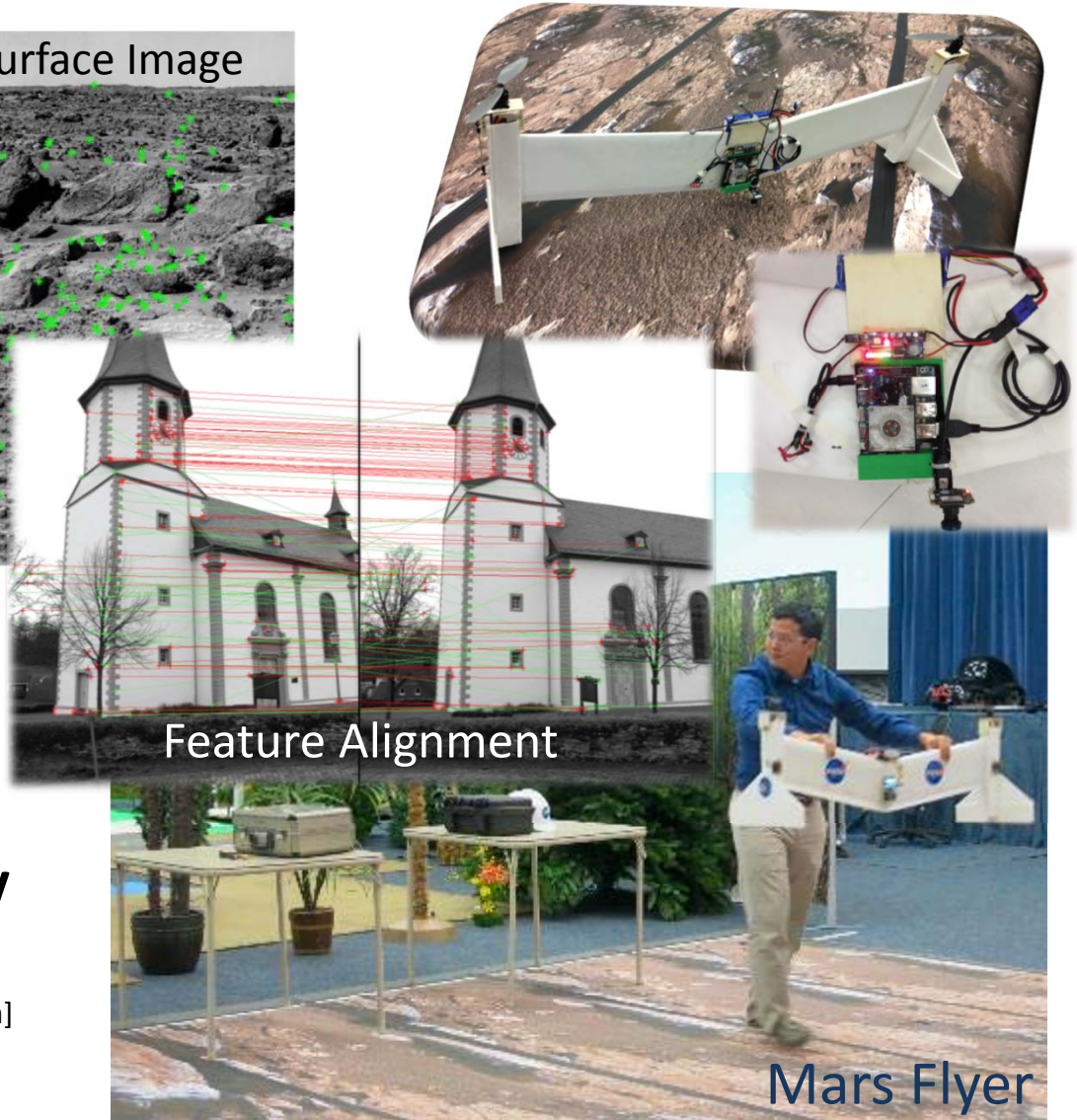
Visual Features Extracted from Mars Surface Image



Fast Semi-Direct Monocular  
Visual Odometry (SVO) with  
**Fault Detection and Recovery**  
for localization and mapping

[C Forster, M Pizzoli, D Scaramuzza]

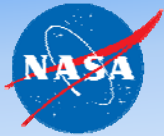
Working with GT and UZH



Feature Alignment

Mars Flyer

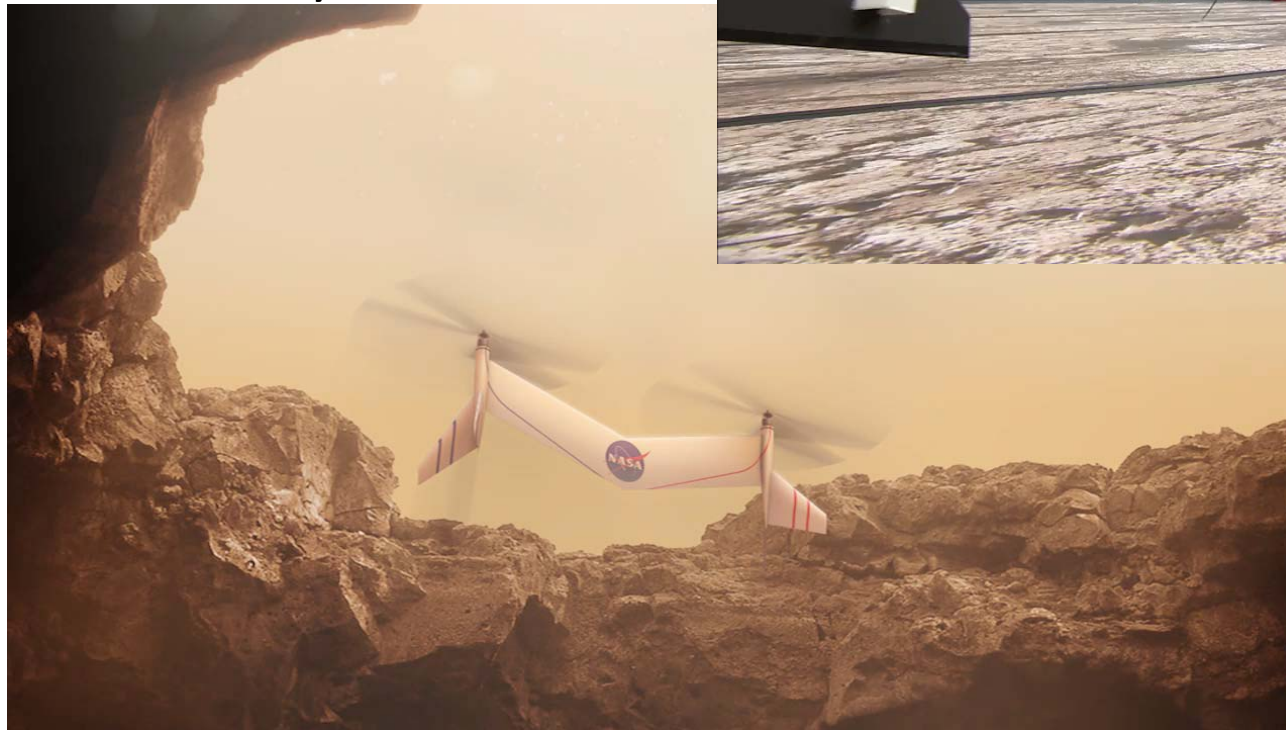




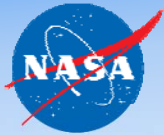
# Data-deprived Navigation



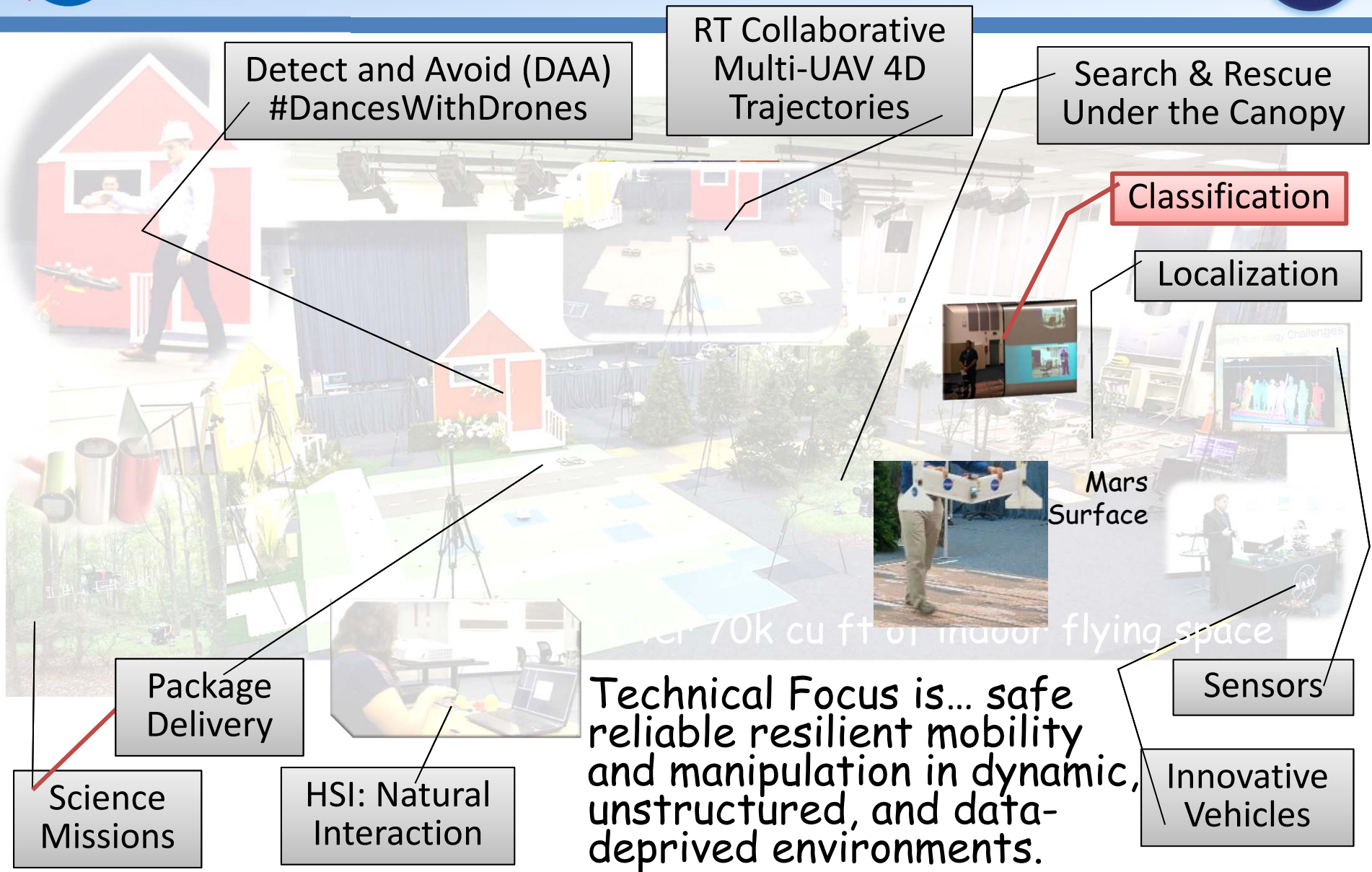
## Mars Electric Reusable Flyer



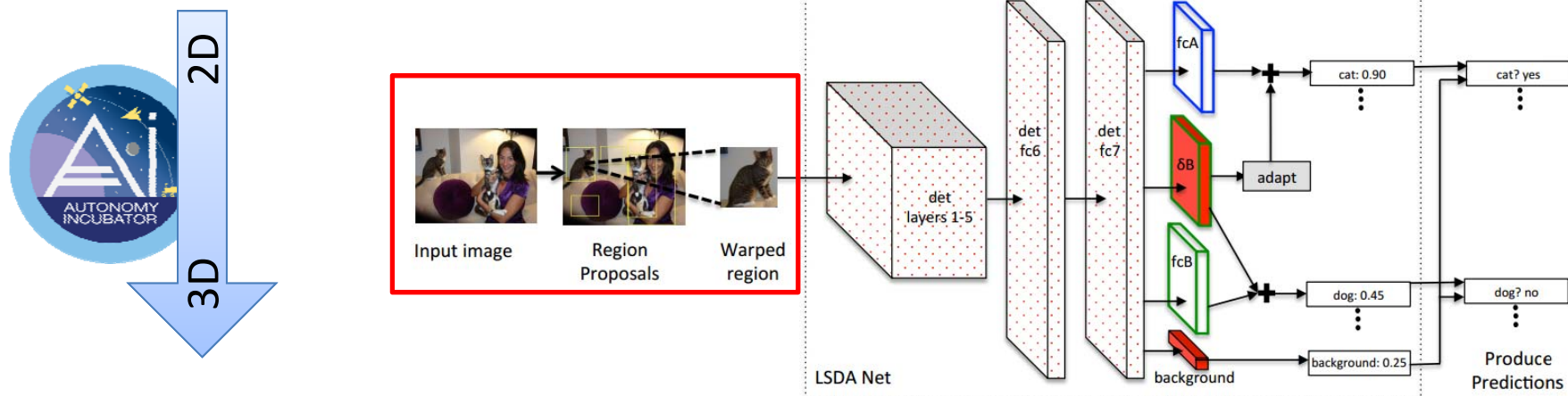
## Robust Visual Odometry



# Autonomy Incubator R&D Overview



- CNN (Convolutional Neural Network)
  - train both a classification and a detection network
- LSDA (Large Scale Detection through Adaptation)
  - train a detector on datasets without bounding box data for all categories



## Alternate Segmentation Approach:

- RGB-D Camera point cloud
- Cluster Extraction (Euclidean, etc.)
- Conversion from point cloud to image

Figure. 2. Detection with the LSDA network. From: J. Hoffman, et al. (2014) "LSDA: Large Scale Detection through Adaptation." NIPS 2014. Retrieved 29 October, 2015, from arXiv:1407.5035



- 3D Space Classification
  - Detect
  - Classify
  - Locate the object in 3D space

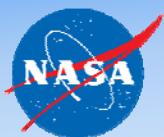


Multiple objects classified:  
kennel, flowers,

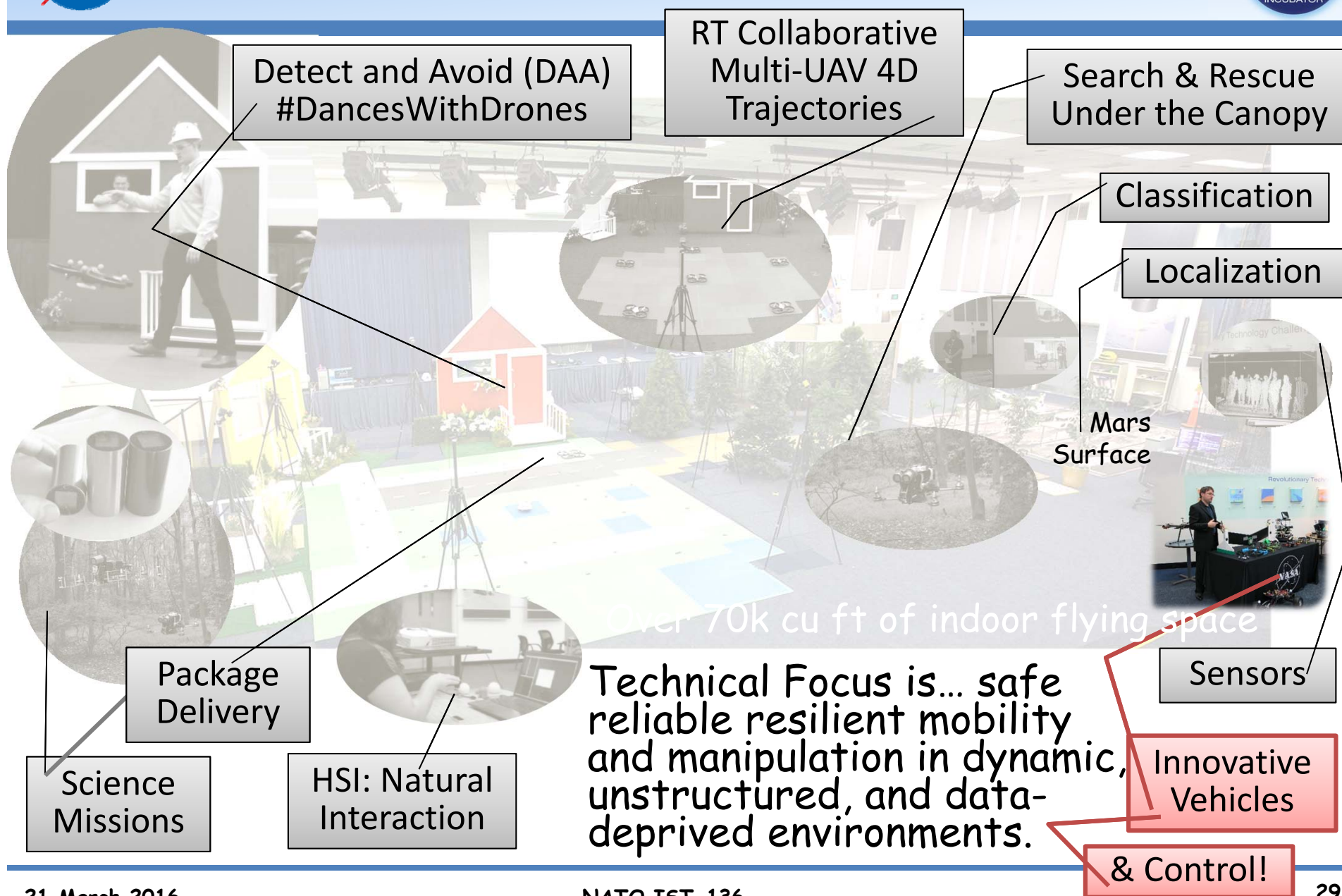


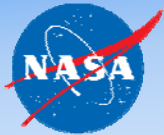
A point cloud visualization  
Object classified as "person"





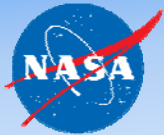
# Autonomy Incubator R&D Overview





# Connected Aircraft

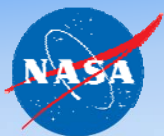




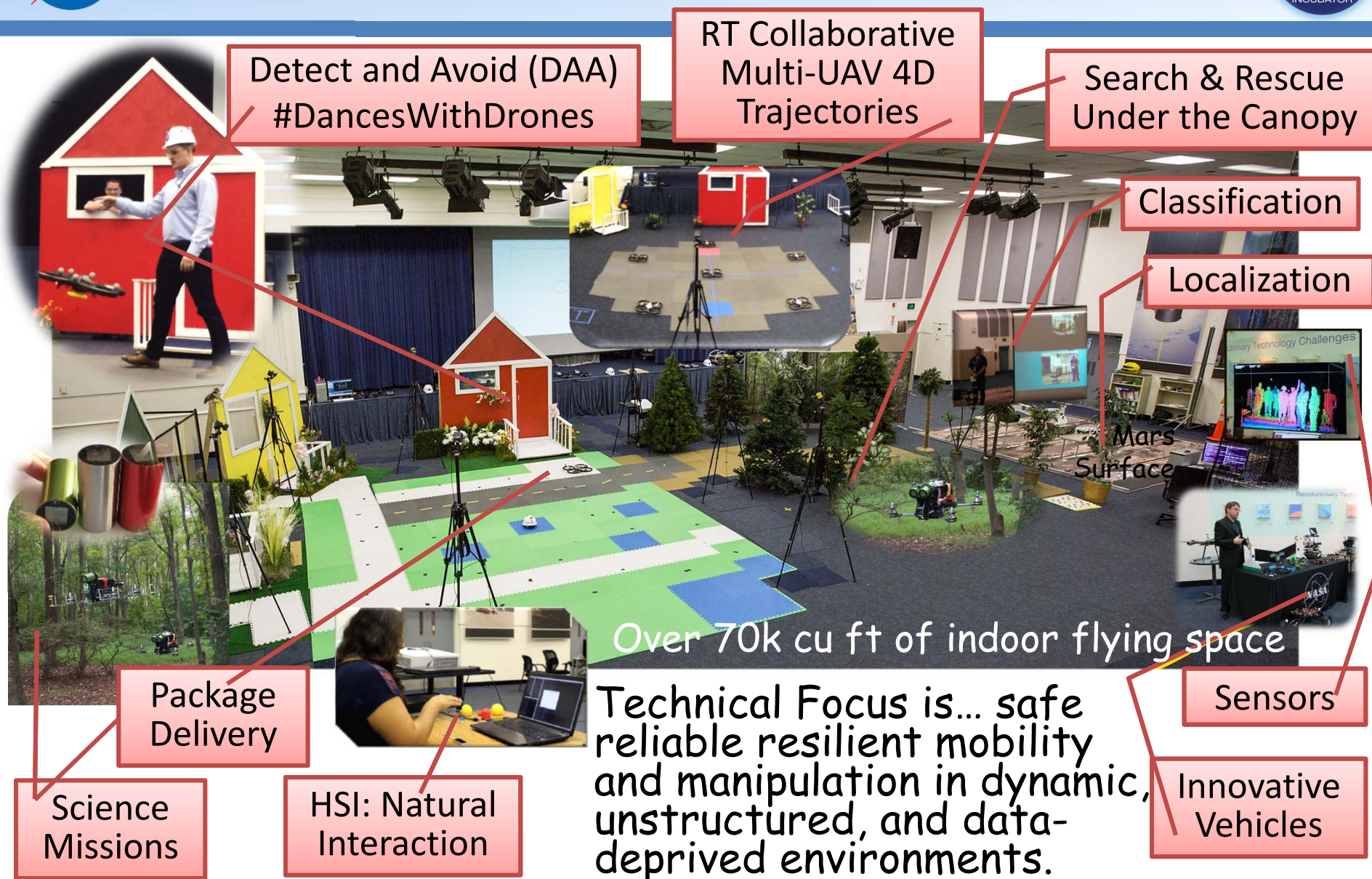
# Agile Control



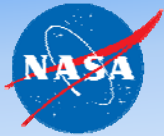




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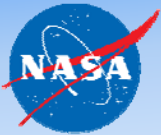




# Safe Reliable Resilient Performance







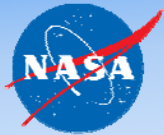
# CERTAIN: Test & Evaluation



## City Environment for Range Testing of Autonomous Integrated Navigation



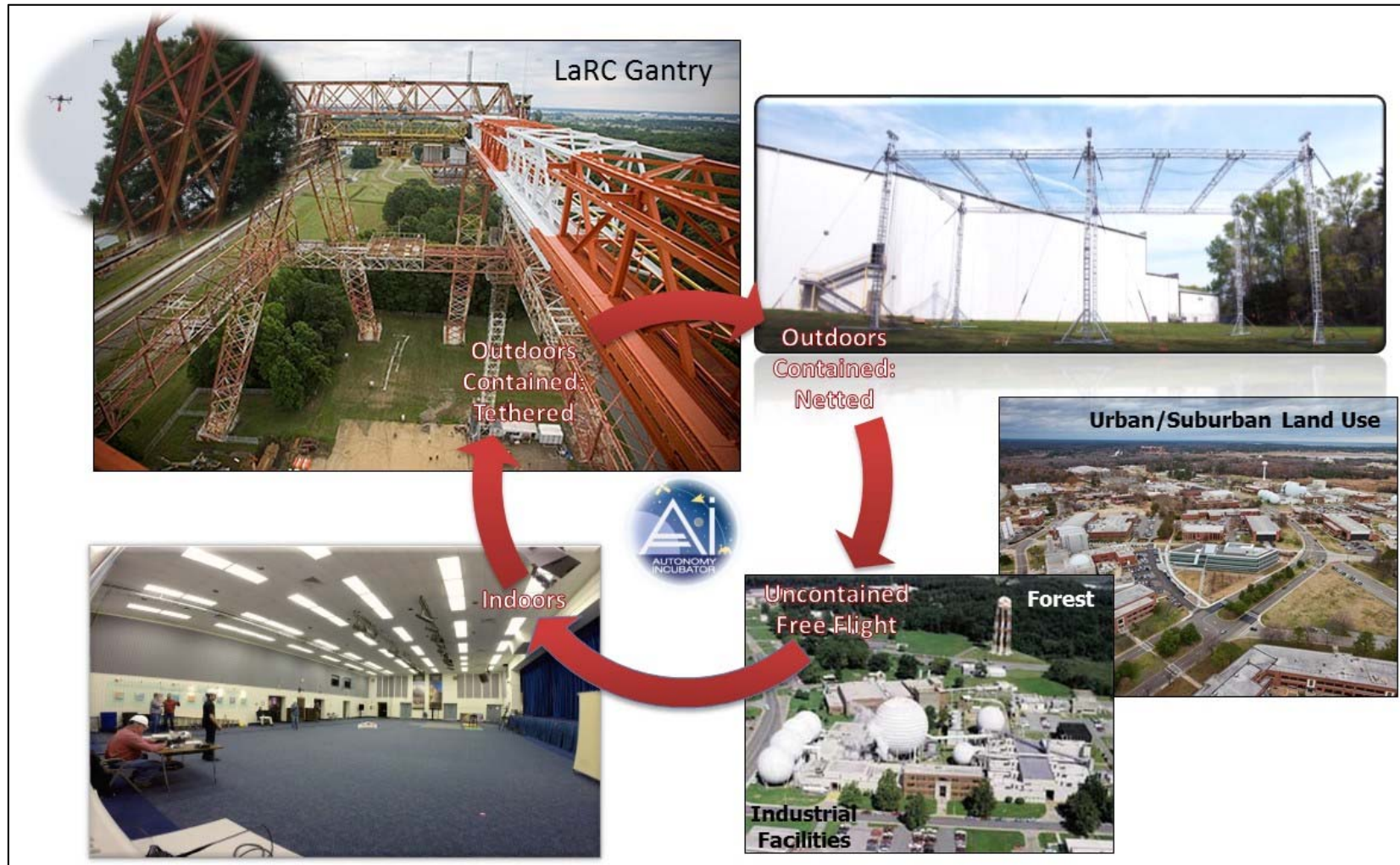


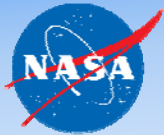


# CERTAIN: Test & Evaluation



## City Environment for Range Testing of Autonomous Integrated Navigation





# Thank You



Blog: <http://autonomyincubator.blogspot.com/>

Twitter: @AutonomyIncub8r

Instagram: @AutonomyIncubator



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