



THE OHIO STATE UNIVERSITY

ELECTROSCIENCE  
LABORATORY

## Using Cognitive Communications to Increase the Operational Value of Collaborative Satellite Networks

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- 1 Motivation and Approach**
- 2 Review of Cognitive Communications**
- 3 Applying Cognition to Collaborative Satellite Networks**
- 4 New Software Library – Observing System Simulations**
- 5 Example Case Study – Deployed Regression**
- 6 Example Case Study – Training a Classification Model**

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- NASA's future space systems will include<sup>1</sup>:

Heterogeneous Networks of Autonomous Small Satellites

- Distributed missions - smaller platforms - but greater numbers
- Resource constraints
  - Low power, restricted duty cycle
  - Limited link availability/reliability, buffer size, reduced power/BW
- Instrument data volume & sensor reconfigurability will expand<sup>2</sup>
- But down-links are not always appropriate (*or possible*)
- Space systems will rely more on inter-satellite communication

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<sup>1</sup>Gilbert J. Clark, Wesley Eddy, Sandra K. Johnson, David E. Brooks, and James L. Barnes, "[Architecture for Cognitive Networking within NASA's Future Space Communications Infrastructure](#)", 34th AIAA International Communications Satellite Systems Conference.

<sup>2</sup>A. G. Schmidt, G. Weisz, M. French, T. Flatley and C. Y. Villalpando, "[SpaceCubeX: A Framework For Evaluating Hybrid Multi-Core CPU/FPGA/DSP Architectures](#)", IEEE Aerospace Conference, 2017

- New mission science goals will depend on collaboration and autonomy
- Introduces a complex decision-making space
- Appealing solution<sup>3</sup>: **Cognitive Communication**
- However, it is insufficient to simply increase link capabilities
- Also necessary to improve the complex communication decision-making:
  - Deciding when to communicate
  - What information is valuable to nodes of the network
  - How to adapt local operations based on new information
- Challenges:
  - Optimizing mission science return (*remote-sensing, etc.*)
  - Increasing the effectiveness of resource-constrained systems
  - Deploying cognitive algorithms on small-satellite platforms

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<sup>3</sup>J. Barnes et. al., "[Machine Learning for Space Communications Service Management Tasks](#)", CCAW 2017

- We developed an open-source C++ space-network simulation library<sup>4</sup>
  - *Purpose:* Efficient simulation of satellite networks with realistic constraints (*power, sensor hardware, communications*)
  - *Key focus:* Simulate sensors that make intelligent decisions based on their own measurements and measurements provided by the network
- We performed simulations to apply cognition to collaborative small sats
- We are investigating:
  - Production of large training data-sets that capture network operation
  - Machine learning techniques to make intelligent communications decisions
  - The applicability of these methods to future cognitive space communication

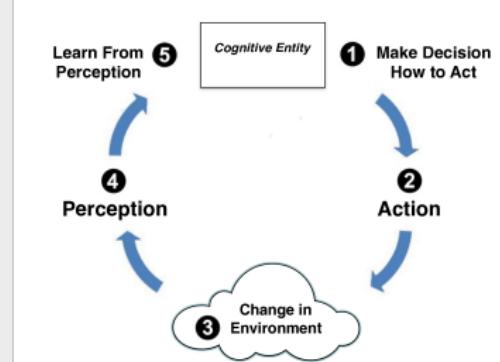
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<sup>4</sup>R. B. Linnabary, A. O'Brien, G. E. Smith, C. D. Ball, and J. T. Johnson, "Open Source Software for Simulating Collaborative Networks of Autonomous Adaptive Sensors", IGARSS, 2019

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## Cognition

- Selecting and carrying out actions based on both:
  - 1 Specific goals
  - 2 Perception of environment
- Learning from experiences
- Interaction with environment



## Cognitive Entity

- An intelligent system that possesses perception, learning, reasoning, and decision making capabilities<sup>a</sup>

<sup>a</sup>G. E. Smith, A. E. Mitchell, C. D. Ball, A. O'Brien, and J. T. Johnson, "Fully Adaptive Remote Sensing Observing System Simulation Experiments", IGARSS, 2018

Communications which are *operationally dependent* on cognition

## Most Existing Work (*Low Level*)

- Cognitive optimization of software-defined radio parameters<sup>5</sup>
- Cognitive satellite digital beamforming<sup>6</sup>
- Bandwidth, Power, Frequency, Modulation

## A New Approach (*High Level*)

- Intelligent routing of information within autonomous network<sup>a</sup>
- Data Contents, Routing, Latency, Sensor Parameters

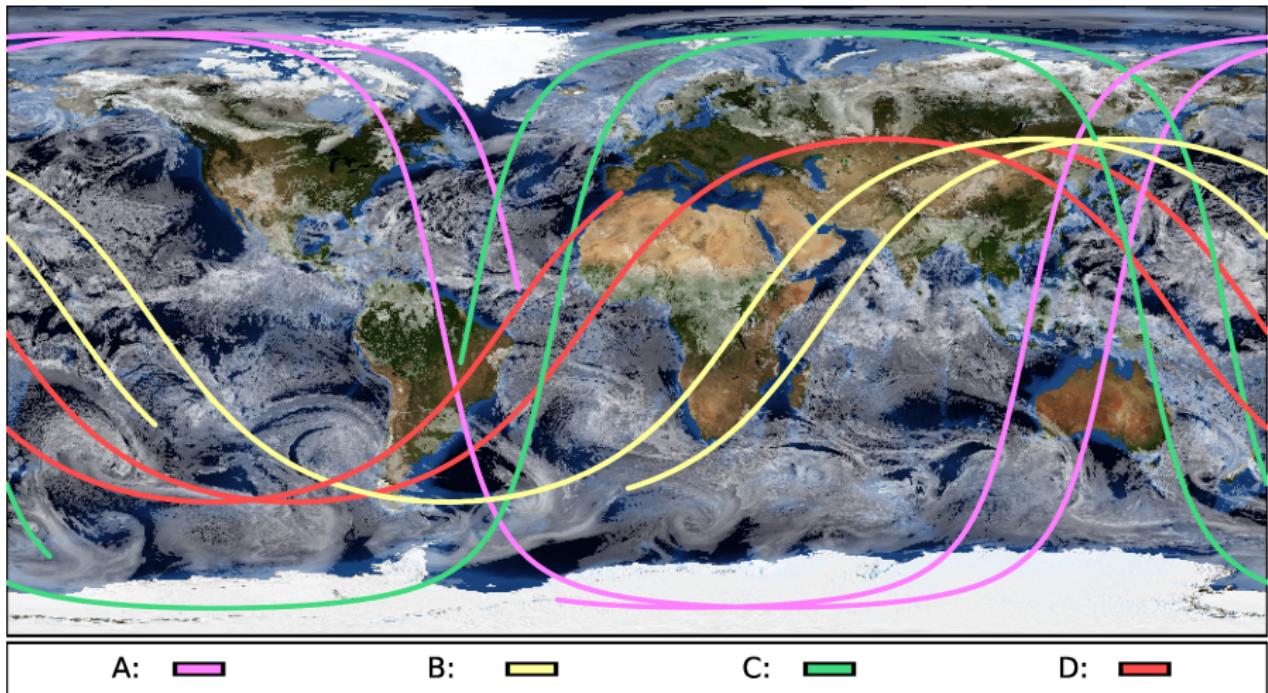
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<sup>5</sup>P. V. R. Ferreira, et al., "[Multi-Objective Reinforcement Learning-Based Deep Neural Networks for Cognitive Space Communications](#)" CCAW, 2017

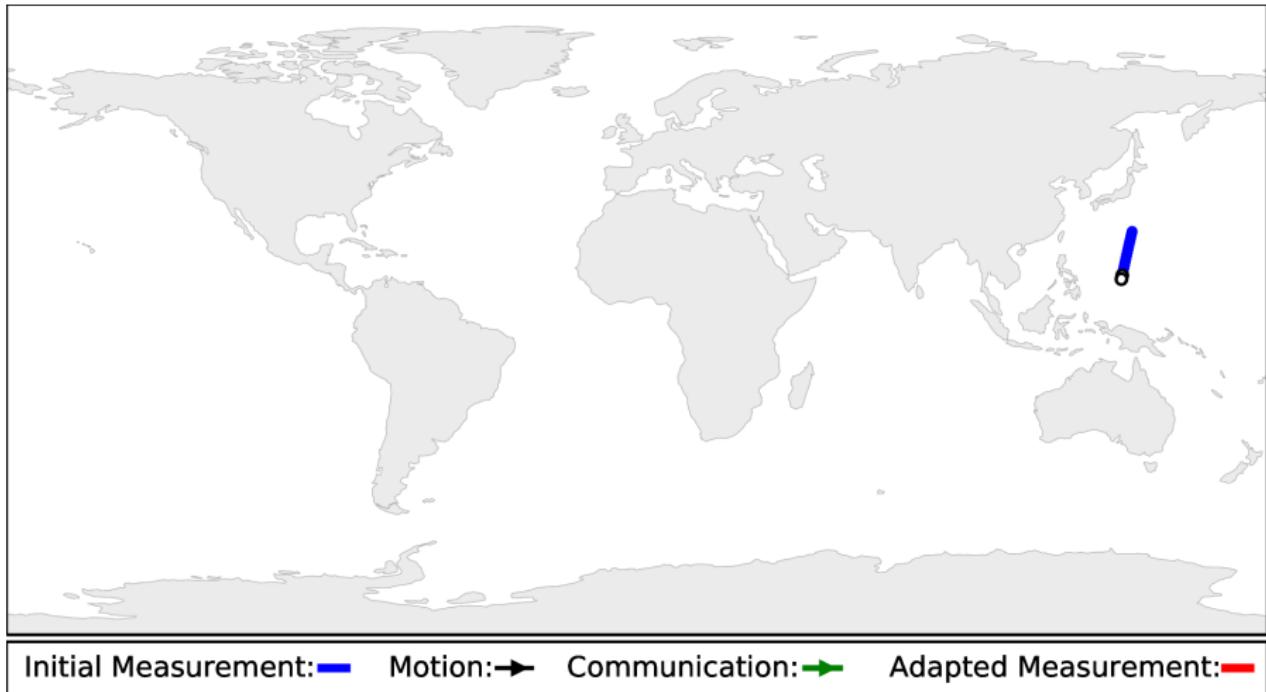
<sup>6</sup>Wenhai Xiong, J. Lu, X. Tian, G. Chen, K. Pham and E. Blasch, "[Cognitive Radio Testbed for Digital Beamforming of Satellite Communication](#)" CCAW, 2017

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Four satellites in separate orbit planes, moving in opposite directions



Satellite A senses atmospheric data and performs *internal processing*

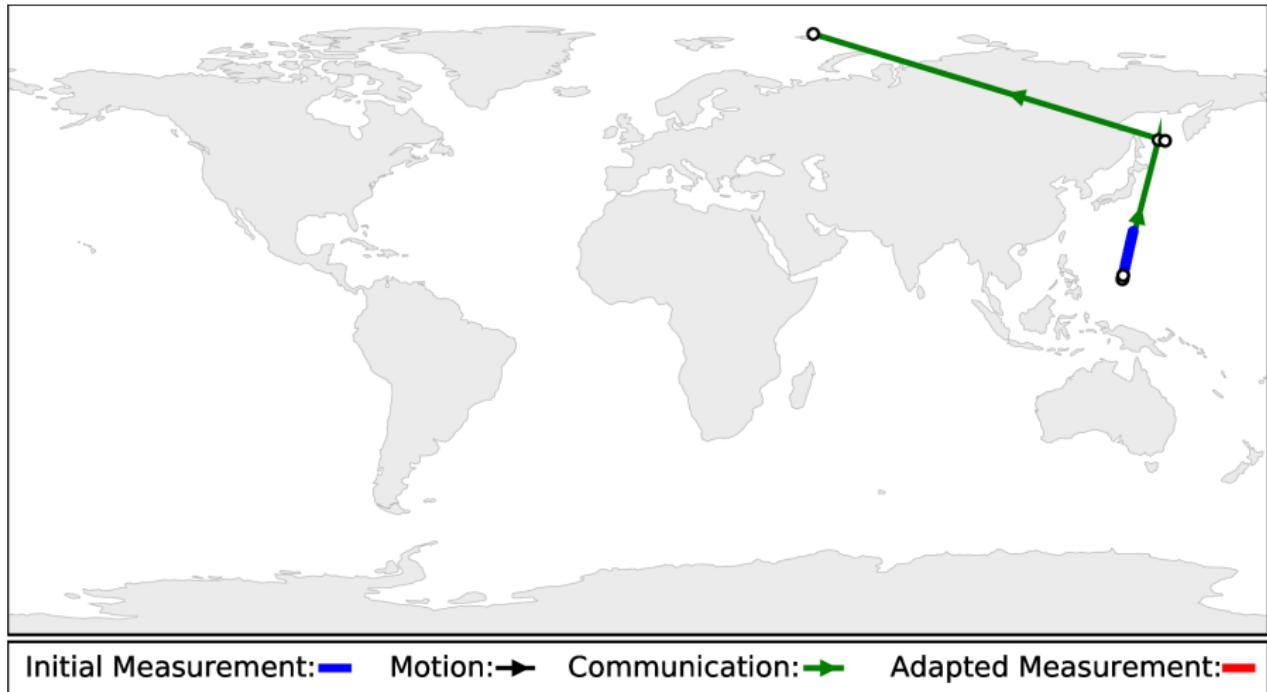


Satellite A forwards a new packet to Satellite B

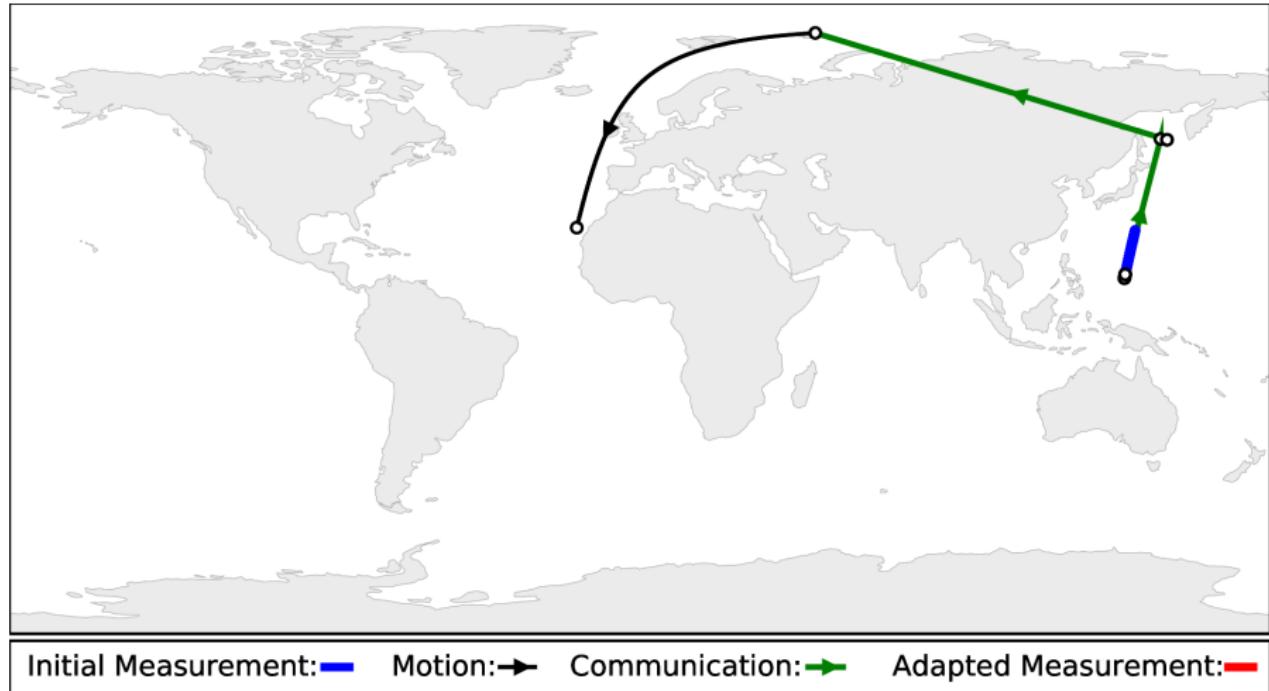


Initial Measurement: — Motion: → Communication: → Adapted Measurement: —

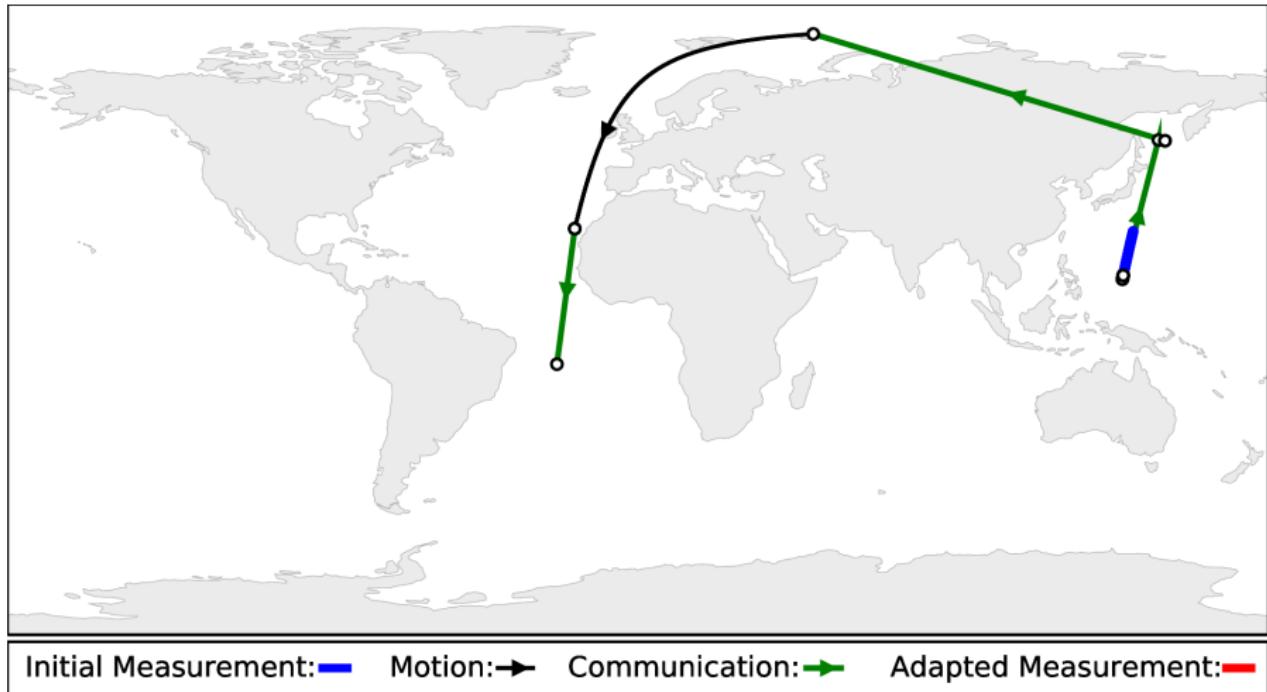
Satellite B forwards the packet to Satellite C



Satellite C holds the packet, *anticipating* an opportunity to transmit

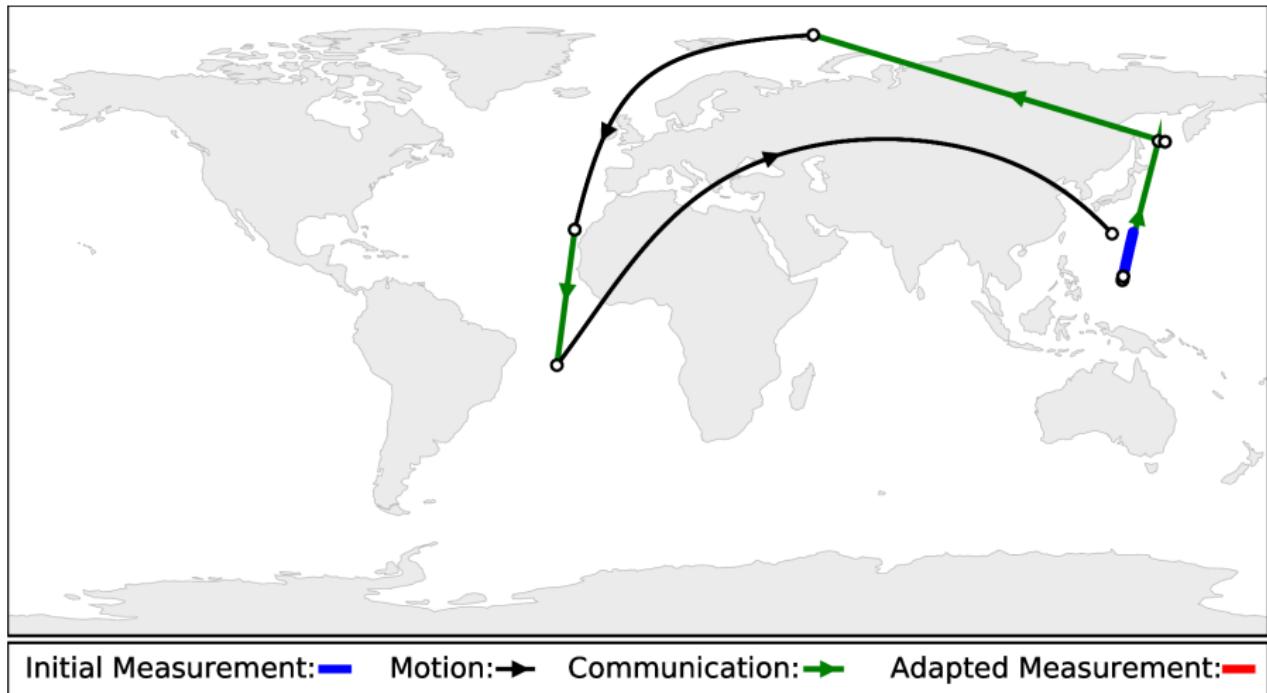


Satellite C forwards the packet to Satellite D

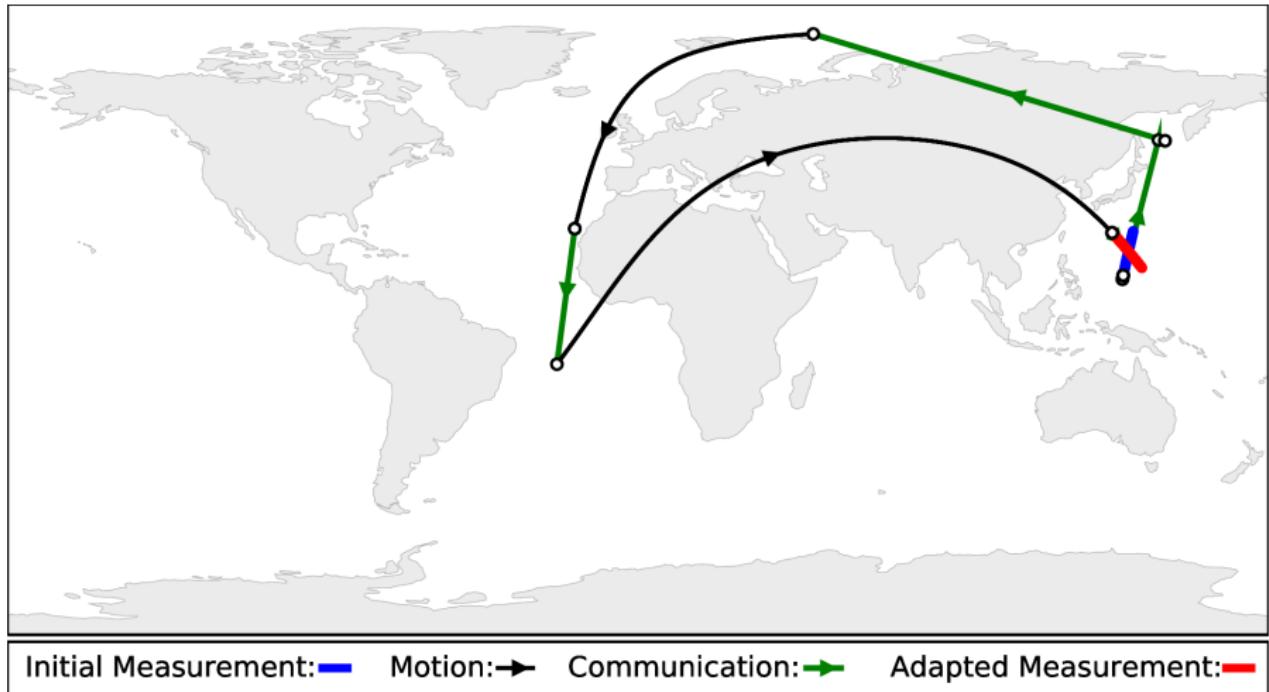


Initial Measurement: Motion: Communication: Adapted Measurement:

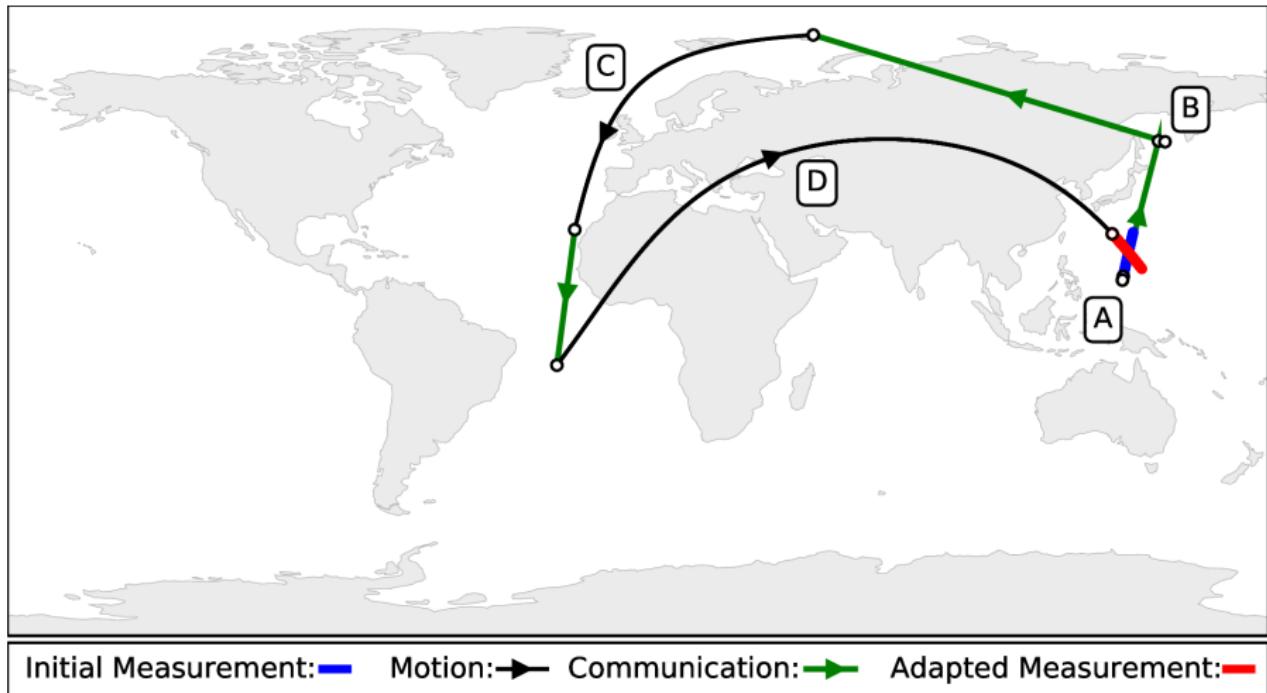
Satellite D processes the packet and adapts its hardware or behavior



Satellite D performs a *follow-up* measurement



A long-term illustration of the collaborative algorithm



Learn From Results  
of Sensor B and  
Actual Route

*Sensor A*  
(Cognitive Entity)

Make Decision on  
Information Content  
and How to Route



Feedback



Communicate



*Sensor B*

*Sensor B Makes  
Measurement  
Influenced by Sensor A*

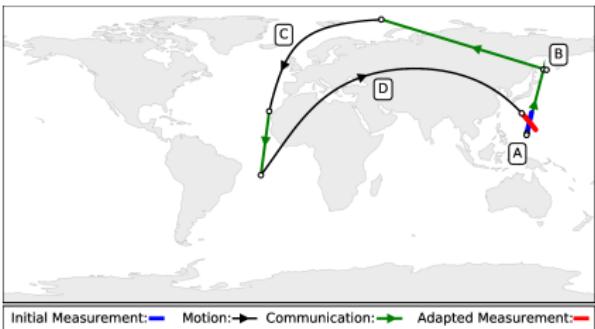
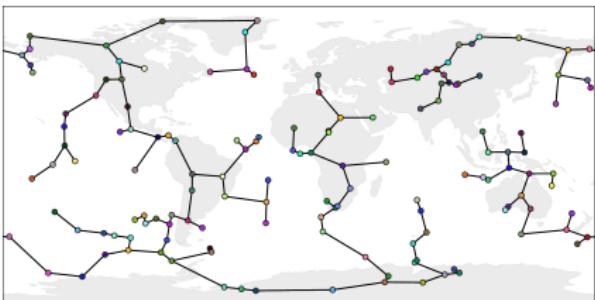
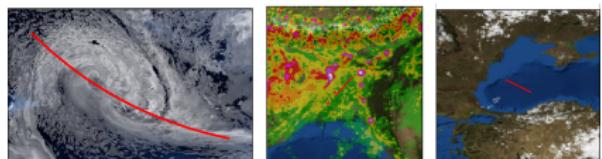
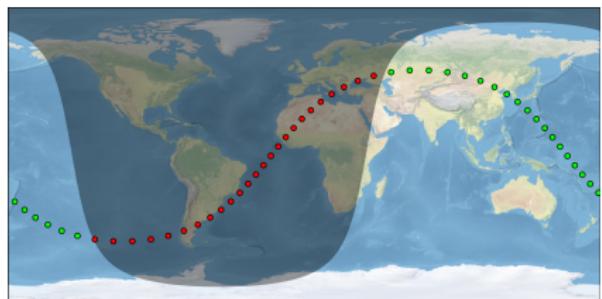
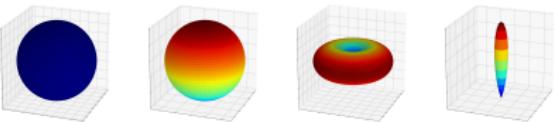
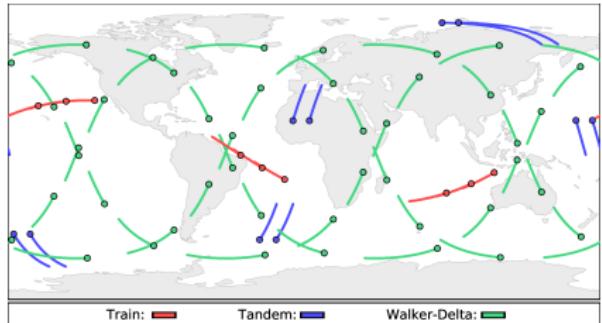
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- COLLABORATE is a software toolset under development for producing training data and implement ML algorithms
  - A C++ development library for observing system simulations
  - A Python package for simulation-data processing (vis./analysis)
- Simulates collaborative networks of satellites
- Focuses on the high-level communication decision space
- Employs network data structures (trees/graphs) to execute predictive route-finding algorithms for efficient communications
- Offers many unique features valuable to future observing system simulation experiments
- Project published to a Git repository, licensed LGPLv3.0.<sup>7</sup>

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<sup>7</sup><https://www.github.com/ryananan/collaborate/>

# COLLABORATE Features



## Advantages

- Latency-tolerant (*data rate & bandwidth*)
- Predictive (*route participants can retain data*)

## Considerations

- Antenna orientation
- Relative velocity (*Doppler*)
- Limited buffer size, channel interruptions

## Assumptions

- Line-of-sight connections with power threshold
- Free-space links
- Established among compatible devices
- No spectrum conflicts (*yet*)

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## Two Main Constellations

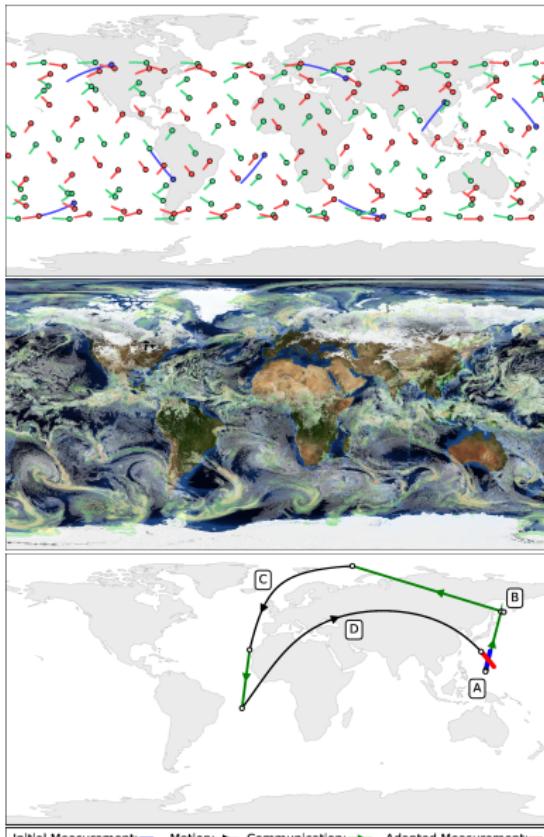
- Cloud Radar
  - Has a *threshold* parameter
- Precipitation Sensor

## Atmospheric Data

- Total cloud optical thickness
- Total precipitation

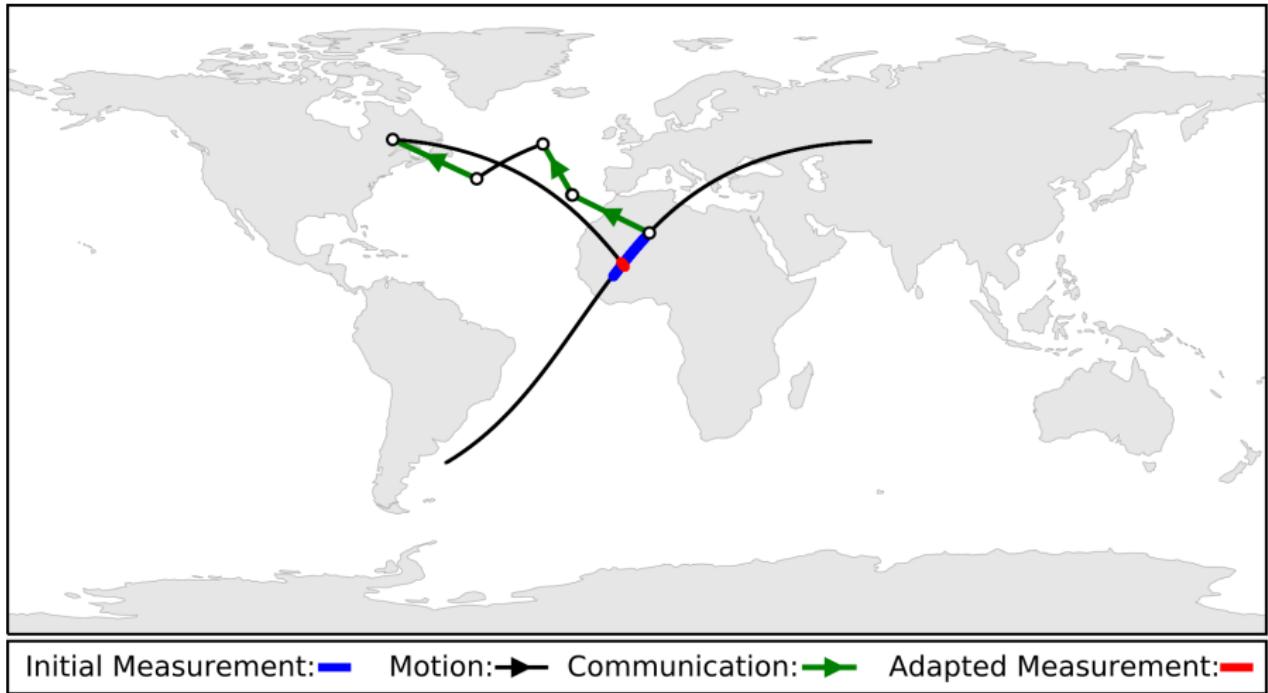
## Exploit Data Correlation

- Goal: maximize precipitation measurements
- Measure cloud thickness at maximum duty-cycle
- Cue precipitation sensors to target regions with high clouds



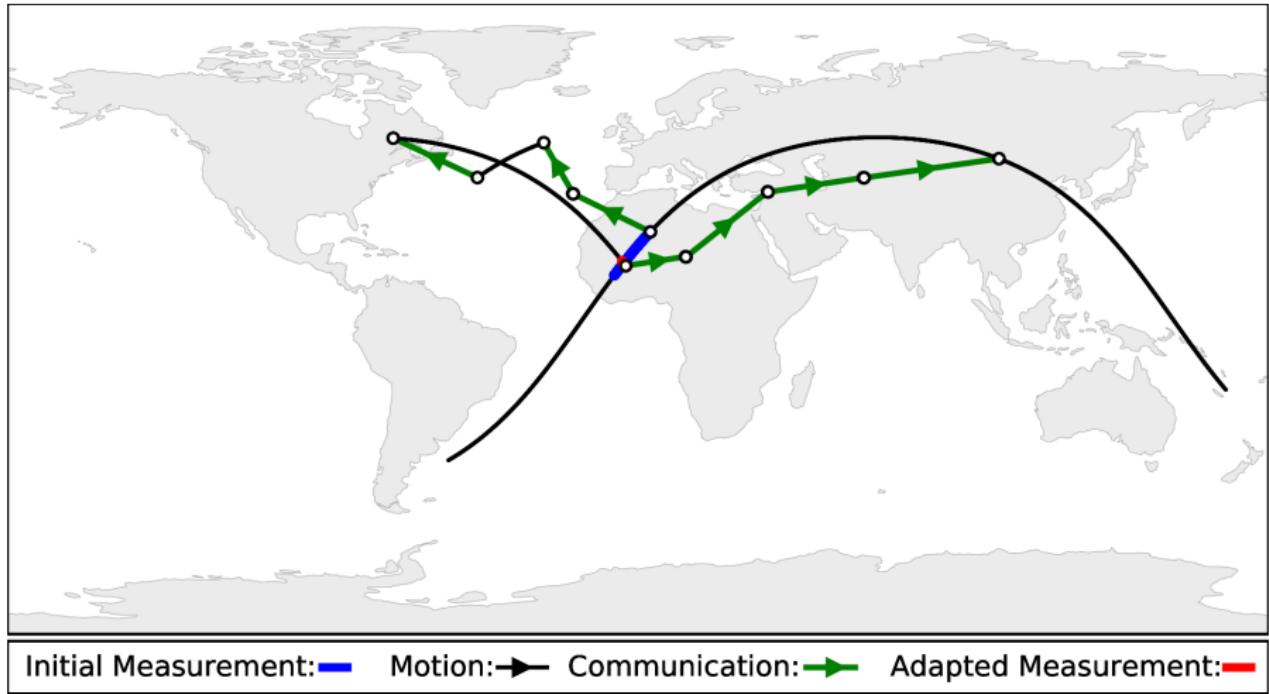
- 1 A cloud radar "A" measures cloud optical thickness
- 2 Cloud thickness exceeds established threshold
- 3 "A" predicts precipitation sensor "B" will visit at time  $\tau$
- 4 "A" forms packet containing *route*, *environment*, and *instructions*
- 5 Packet is forwarded through network before time  $\tau$
- 6 "B" takes follow-up measurement
- 7 "B" forms packet containing *route*, *environment*, and *instructions*
- 8 Packet is forwarded through network ASAP
- 9 "A" modifies internal threshold based on success criteria
- 10 Cycle is repeated

## Feed-Forward Route



- 1 A cloud radar "A" measures cloud optical thickness
- 2 Cloud thickness exceeds established threshold
- 3 "A" predicts precipitation sensor "B" will visit at time  $\tau$
- 4 "A" forms packet containing *route*, *environment*, and *instructions*
- 5 Packet is forwarded through network to "B" before time  $\tau$
- 6 "B" takes follow-up measurement
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- 8 Packet is forwarded through network to "A" ASAP
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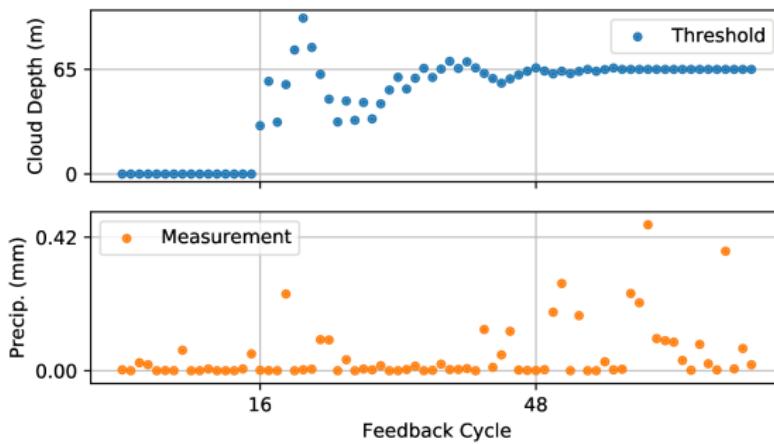
## Feedback Route



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Simulation steps are repeated for several cycles until the sensors in the network have adapted based on feedback from one another

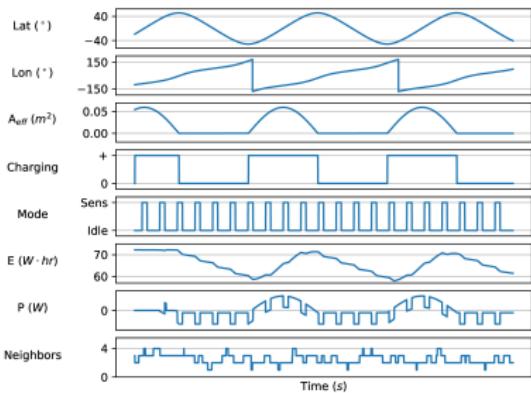
- Cycles 1-16: Measure precipitation anywhere
- Cycles 16-48: Regression
- Cycles 48- $\infty$ : Measure precipitation where clouds  $\geq$  65 meters



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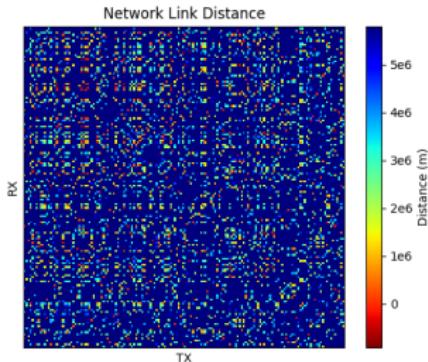
## Simulation Data

- NetCDF format
- Satellite parameters
- Communication channel data
- Network structures
- Sensor measurements

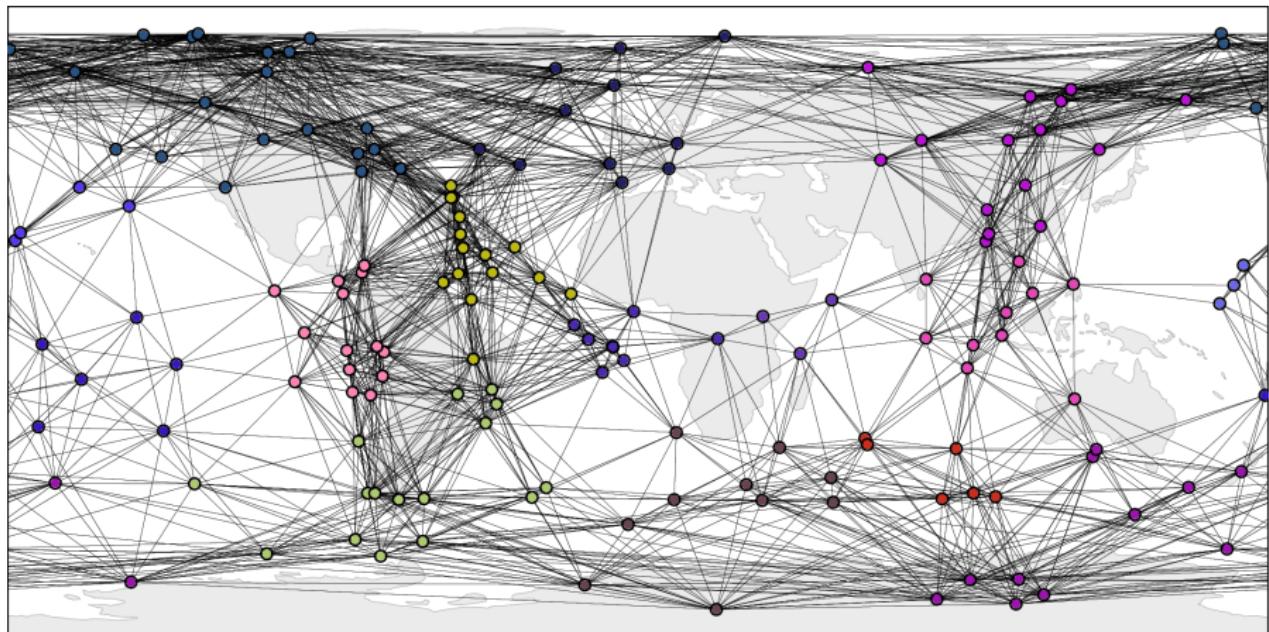


## Machine-Learning Model

- Python
- Scikit-Learn
- *Input*: weighted adjacency matrices
- Spectral clustering (k-means)

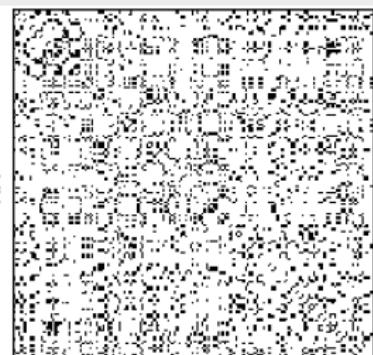


All line-of-sight connections in network (*potential links*)

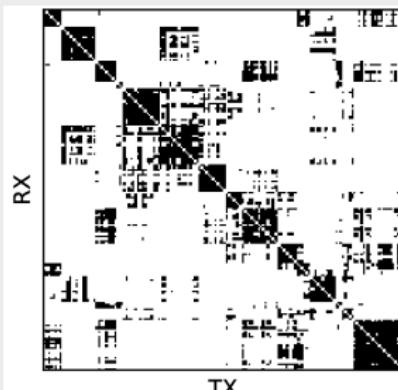


- Classify satellites based on *proximity*
- Extendable to other satellite parameters in the simulation data
- May inform beam direction, channel decisions, TX power, etc.

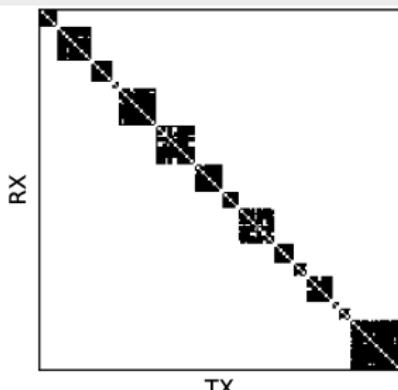
Lines-of-Sight



Sorted Network

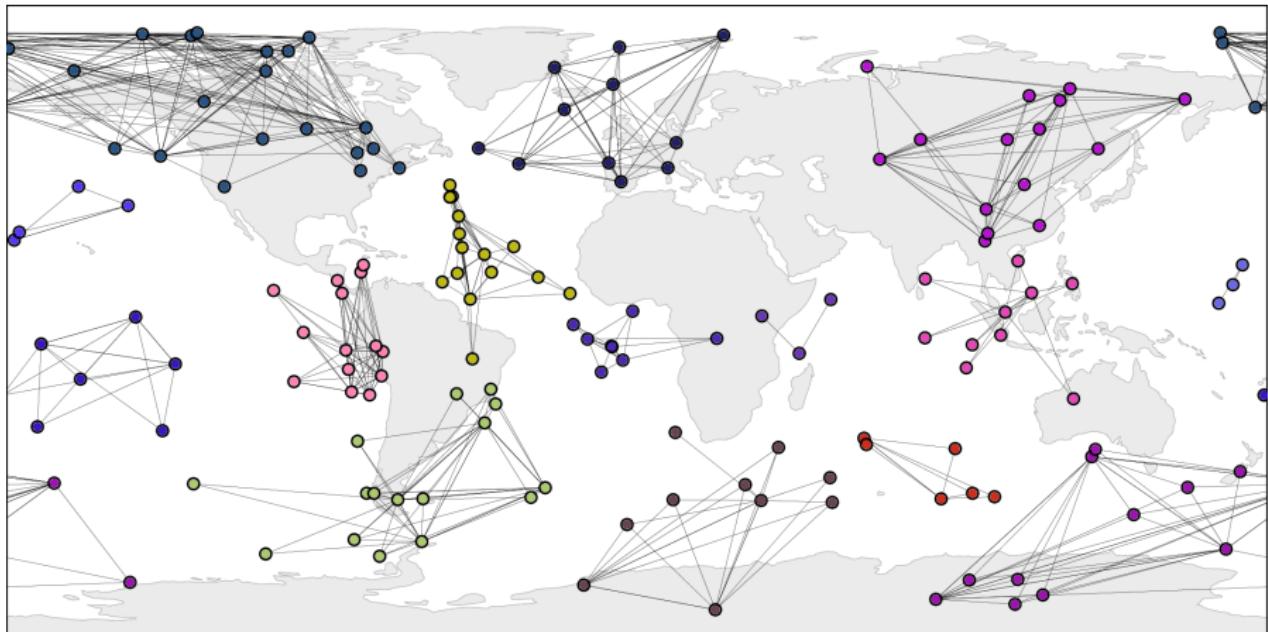


Clusters



## Example - Clustering Results

Isolated satellite network clusters



- Future small-satellites will carry adaptive instruments
- Parameters will be intelligently re-configured
- *Primary purpose collaborative communication* among small satellites:
  - Achieve system-level adaptivity with future instruments
- We introduced our current work:
  - Applying cognitive communications to information flow in a collaborative small-satellite network
  - Using COLLABORATE to investigate cognition as one means of overcoming the complex decision space for such small satellites
- Future goal:
  - Simulate more advanced ML algorithms (employ Neural Network)
  - Add valuable algorithms identified in Python to the C++ routines