



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 include¹:

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 expand²
- But down-links are not always appropriate (*or possible*)
- Space systems will rely more on inter-satellite communication

¹Gilbert 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.

²A. 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 solution³: **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

³J. Barnes et. al., "[Machine Learning for Space Communications Service Management Tasks](#)", CCAAW 2017



- We developed an open-source C++ space-network simulation library⁴
 - *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

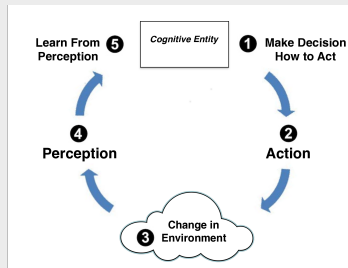
⁴R. B. Linnabary, A. O'Brien, G. E. Smith, C. D. Ball, and J. T. Johnson, "Open Source Software for Simulating Collaborative Networks of Autonomous Adaptive Sensors", IGARSS, 2019



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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^a

^aG. 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 parameters⁵
- Cognitive satellite digital beamforming⁶
- Bandwidth, Power, Frequency, Modulation

A New Approach (*High Level*)

- Intelligent routing of information within autonomous network^a
- Data Contents, Routing, Latency, Sensor Parameters

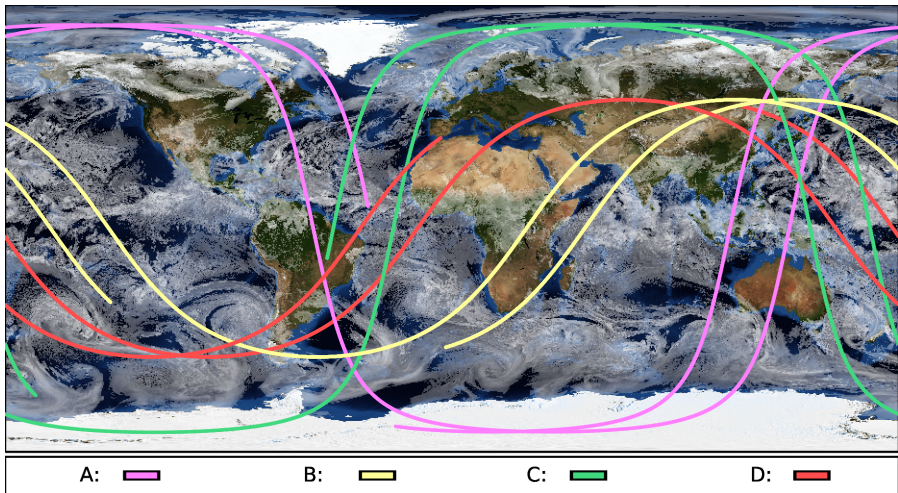
⁵P. V. R. Ferreira, et al., "Multi-Objective Reinforcement Learning-Based Deep Neural Networks for Cognitive Space Communications" CCAAW, 2017

⁶Wenhao Xiong, J. Lu, X. Tian, G. Chen, K. Pham and E. Blasch, "Cognitive Radio Testbed for Digital Beamforming of Satellite Communication" CCAAW, 2017

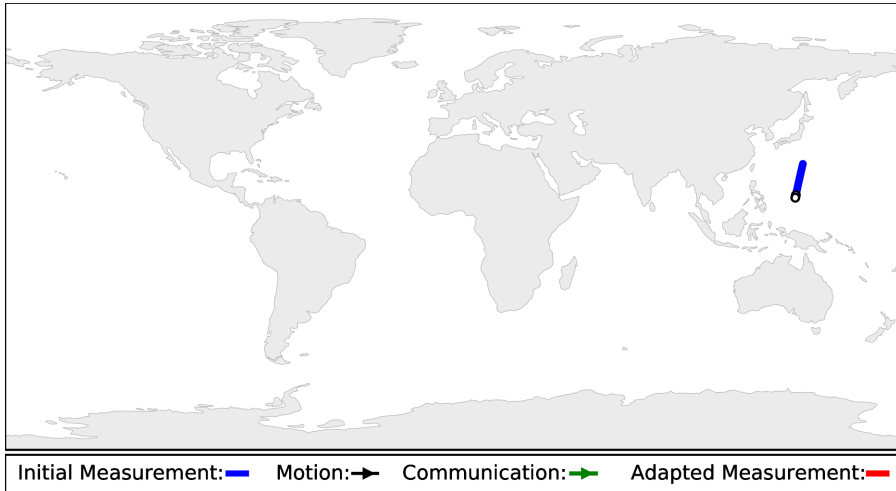


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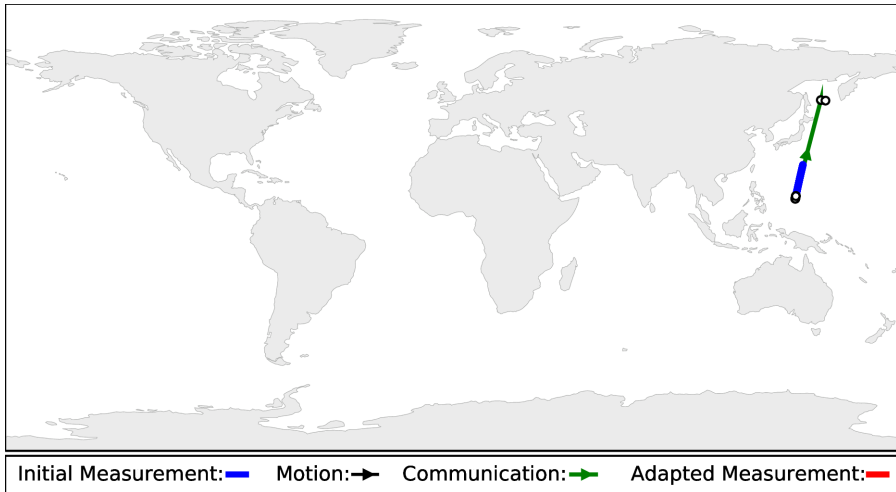
Four satellites in separate orbit planes, moving in opposite directions



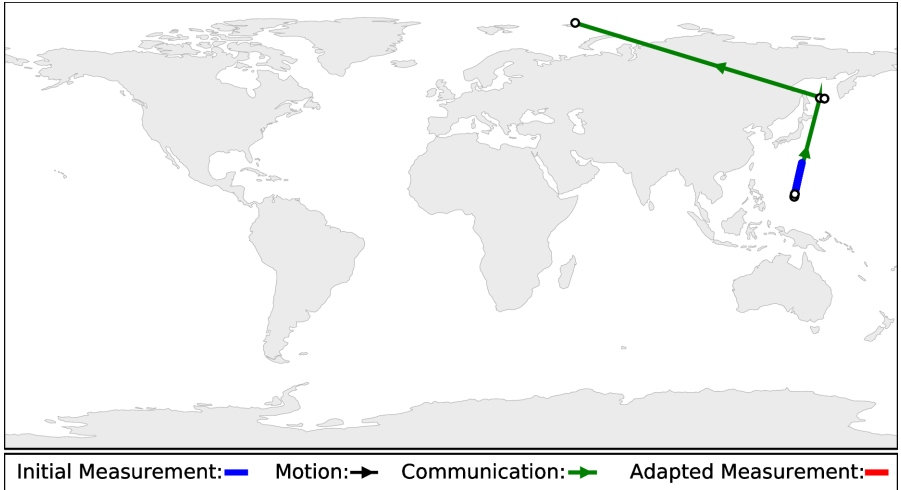
Satellite **A** senses atmospheric data and performs *internal processing*



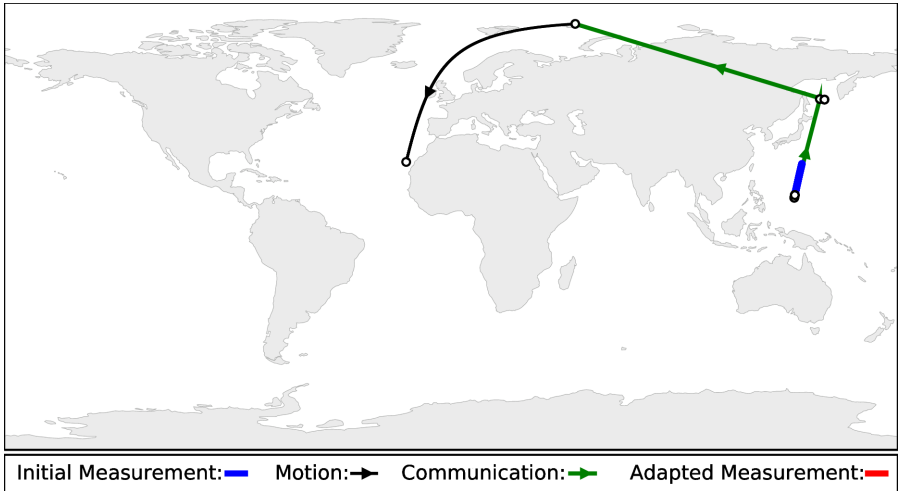
Satellite A forwards a new packet to Satellite B



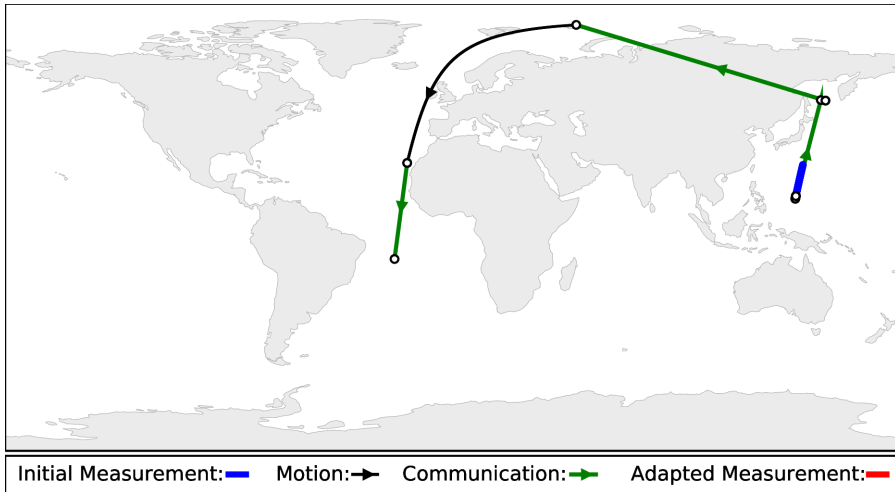
Satellite B forwards the packet to Satellite C



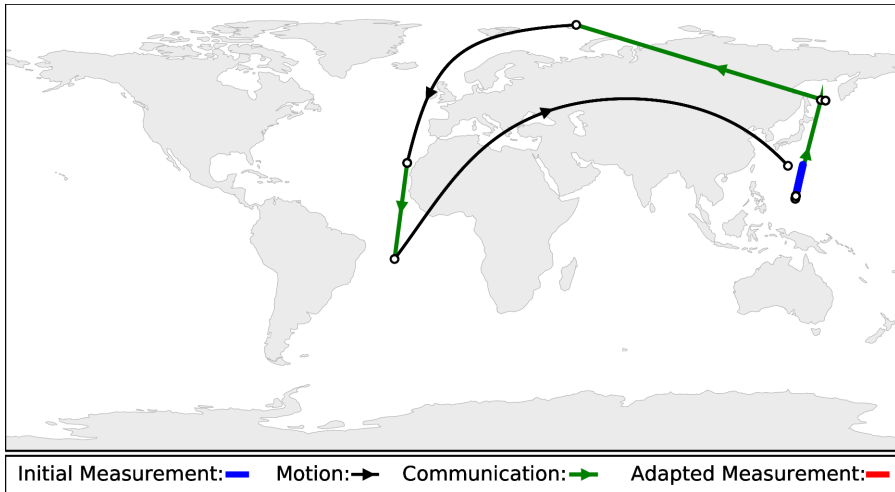
Satellite C holds the packet, *anticipating* an opportunity to transmit



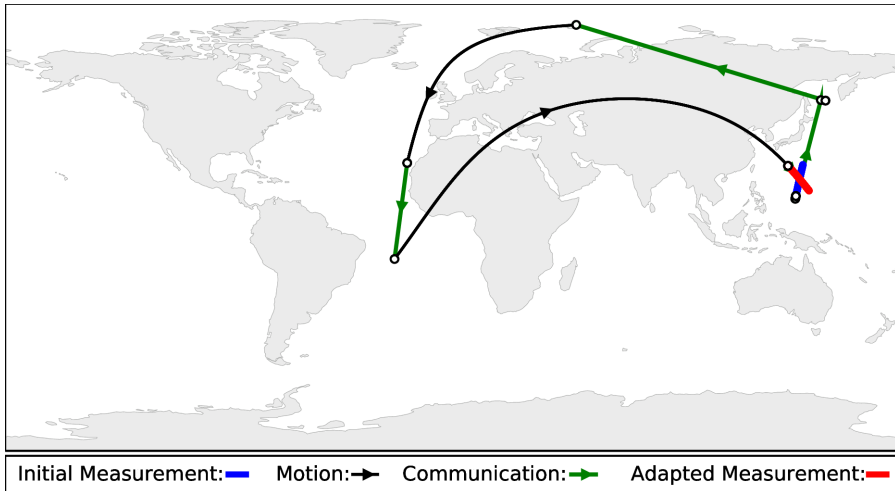
Satellite C forwards the packet to Satellite D



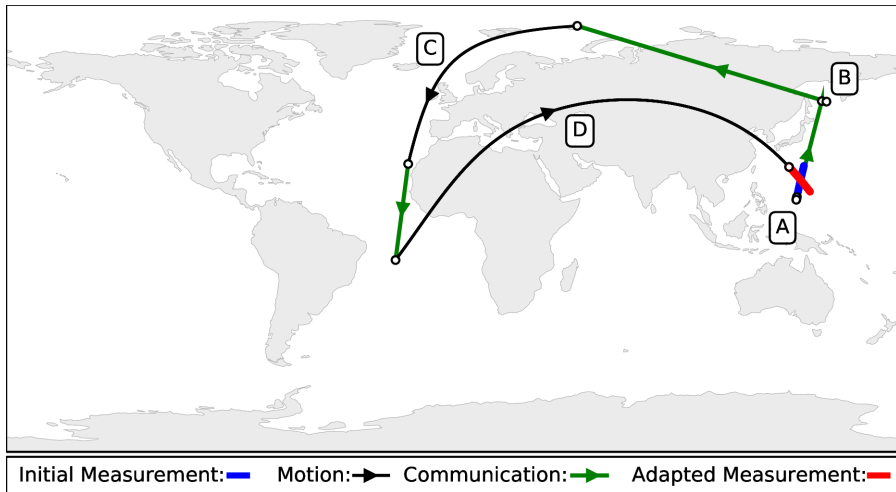
Satellite D processes the packet and adapts its hardware or behavior



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***

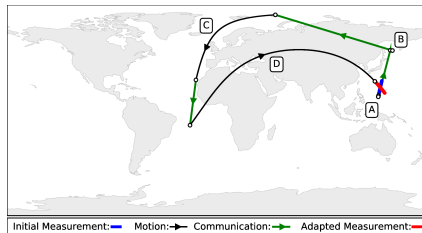
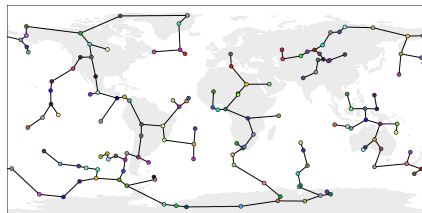
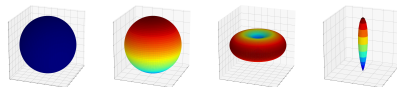
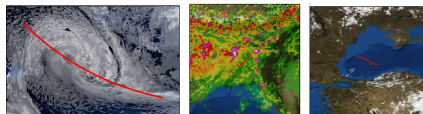
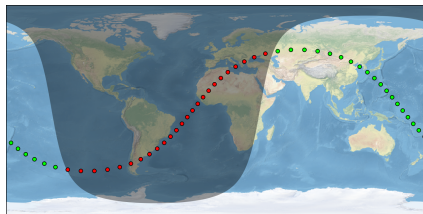
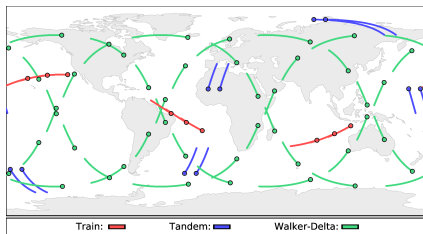


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- COLLABORATE is a software toolset under development for producing training data and implement ML algorithms
 - A C++ development library for observing system simulations
 - A Python package for simulation-data processing (vis./analysis)
- Simulates 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 LGPLv3.0.⁷

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





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

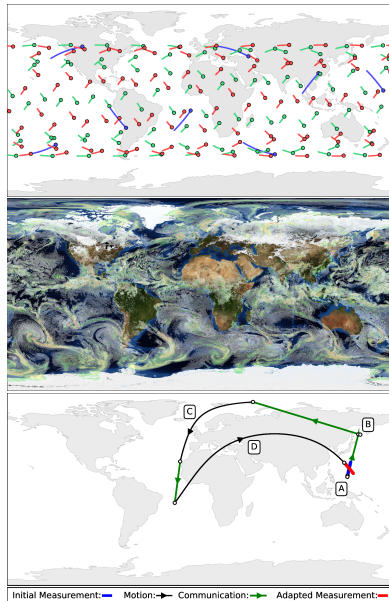
- Cloud Radar
 - Has a *threshold parameter*
- Precipitation Sensor

Atmospheric Data

- Total cloud optical thickness
- Total precipitation

Exploit Data Correlation

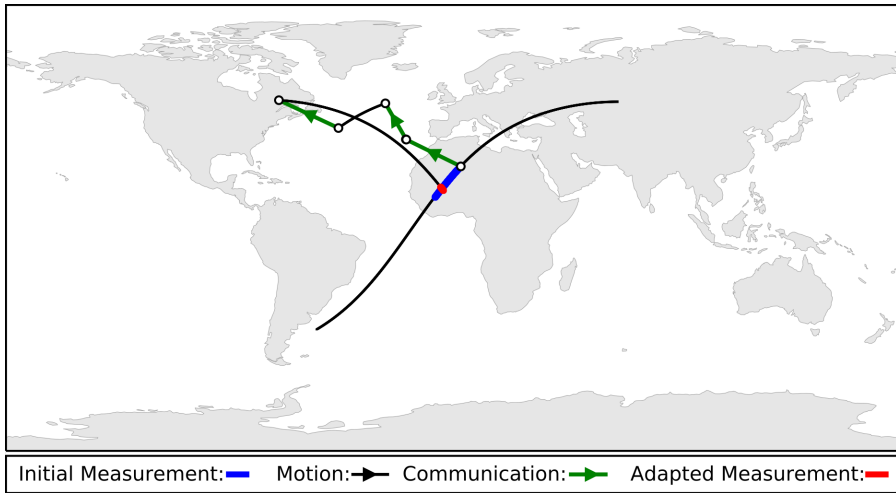
- Goal: maximize precipitation measurements
- Measure cloud thickness at maximum duty-cycle
- Cue precipitation sensors to target regions with high clouds





- 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 τ
- 4 "A" forms packet containing *route*, *environment*, and *instructions*
- 5 Packet is forwarded through network before time τ
- 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

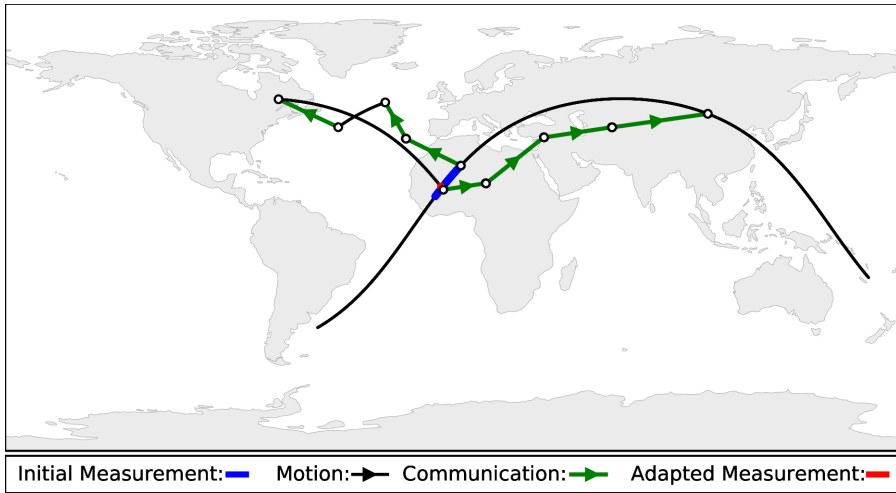
Feed-Forward Route





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- 5 Packet is forwarded through network to "B" before time τ
- 6 "B" takes follow-up measurement
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Feedback Route

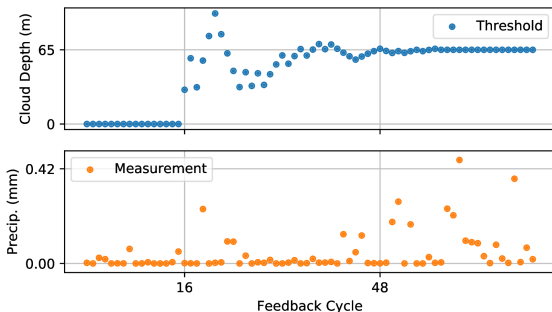




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Simulation steps are repeated for several cycles until the sensors in the network have adapted based on feedback from one another

- Cycles 1-16: Measure precipitation anywhere
- Cycles 16-48: Regression
- Cycles 48-∞: Measure precipitation where clouds ≥ 65 meters





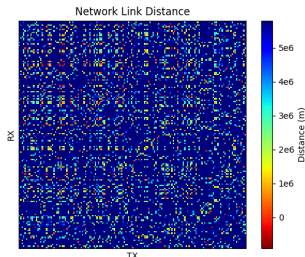
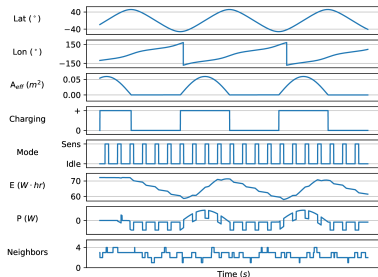
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Simulation Data

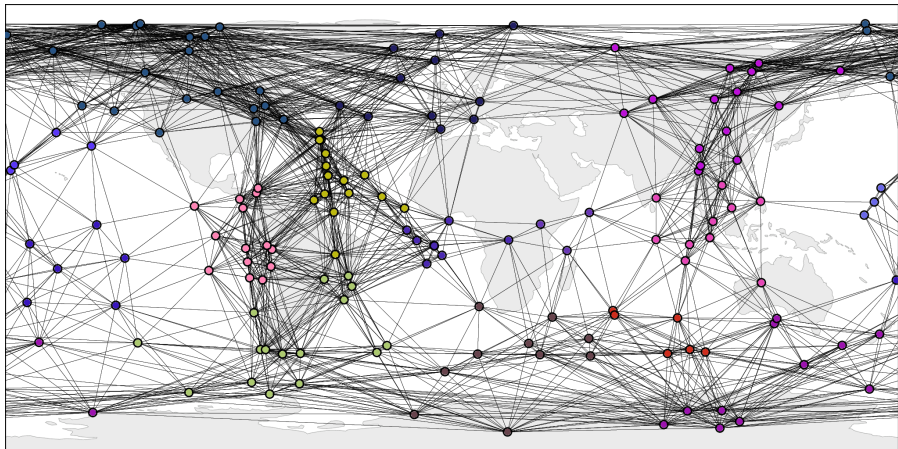
- NetCDF format
- Satellite parameters
- Communication channel data
- Network structures
- Sensor measurements

Machine-Learning Model

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

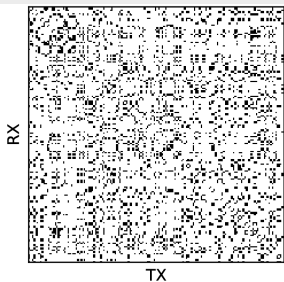


All line-of-sight connections in network (*potential links*)

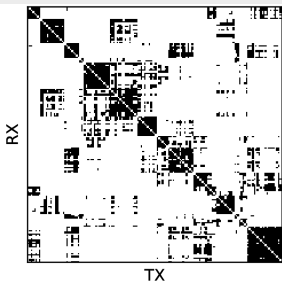


- Classify satellites based on *proximity*
- Extendable to other satellite parameters in the simulation data
- May inform beam direction, channel decisions, TX power, etc.

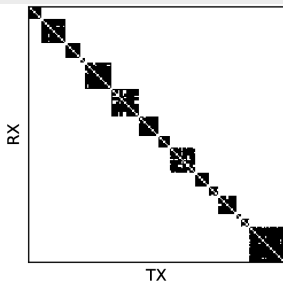
Lines-of-Sight



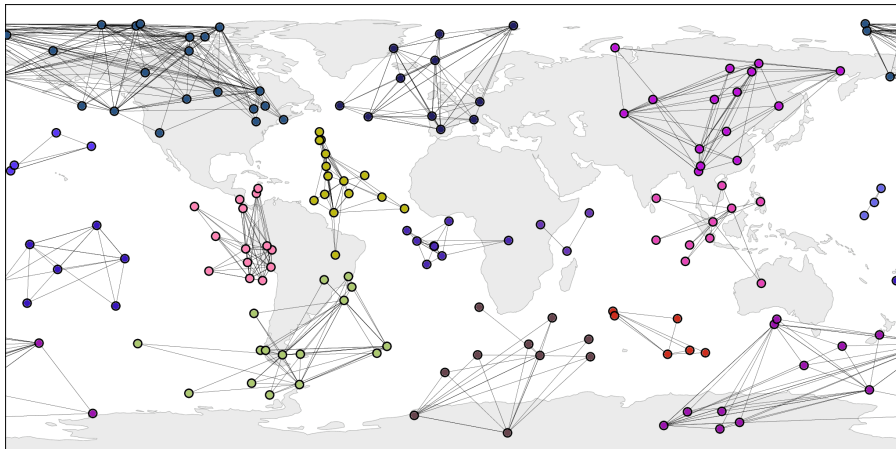
Sorted Network



Clusters



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