



D-SHIELD: Distributed Spacecraft with Heuristic Intelligence to Enable Logistical Decisions

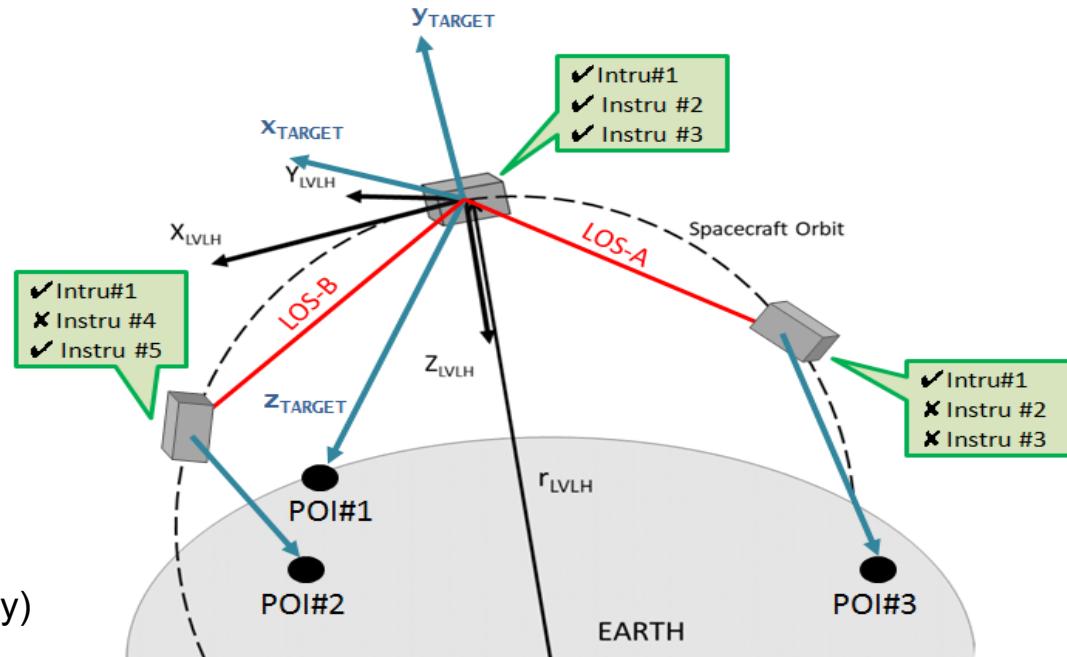
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Background: Motivation

- Multi-payload, multi-spacecraft constellation scheduling for spatio-temporally varying science observations
- Small Sat constellation + Full-body reorientation agility + scheduling autonomy
= More Coverage, for any given number of satellites in any given orbits
- Ground scheduling algorithm allows 2-sat, 1-imager constellation over 12 hours to observe 2.5x compared to the fixed pointing approach. 1.5x with a 4-sat constellation
- Onboard scheduling algorithm allows 24-sat, 1-rainradar constellation to observe ~7% more flood magnitude than ground scheduling



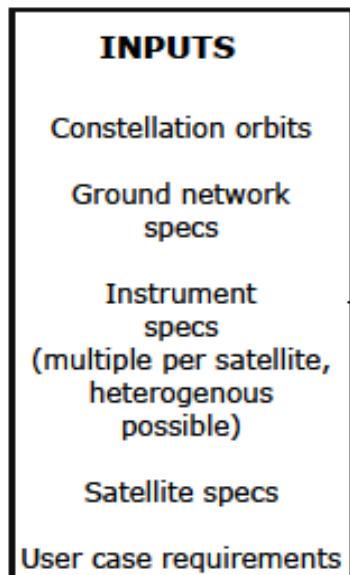
Published Use Cases:

1. Land coverage and coral tracking (COSPAR ASR)
2. Cyclone tracking (IEEE TGRS)
3. Urban Floods (J.Hydrology)



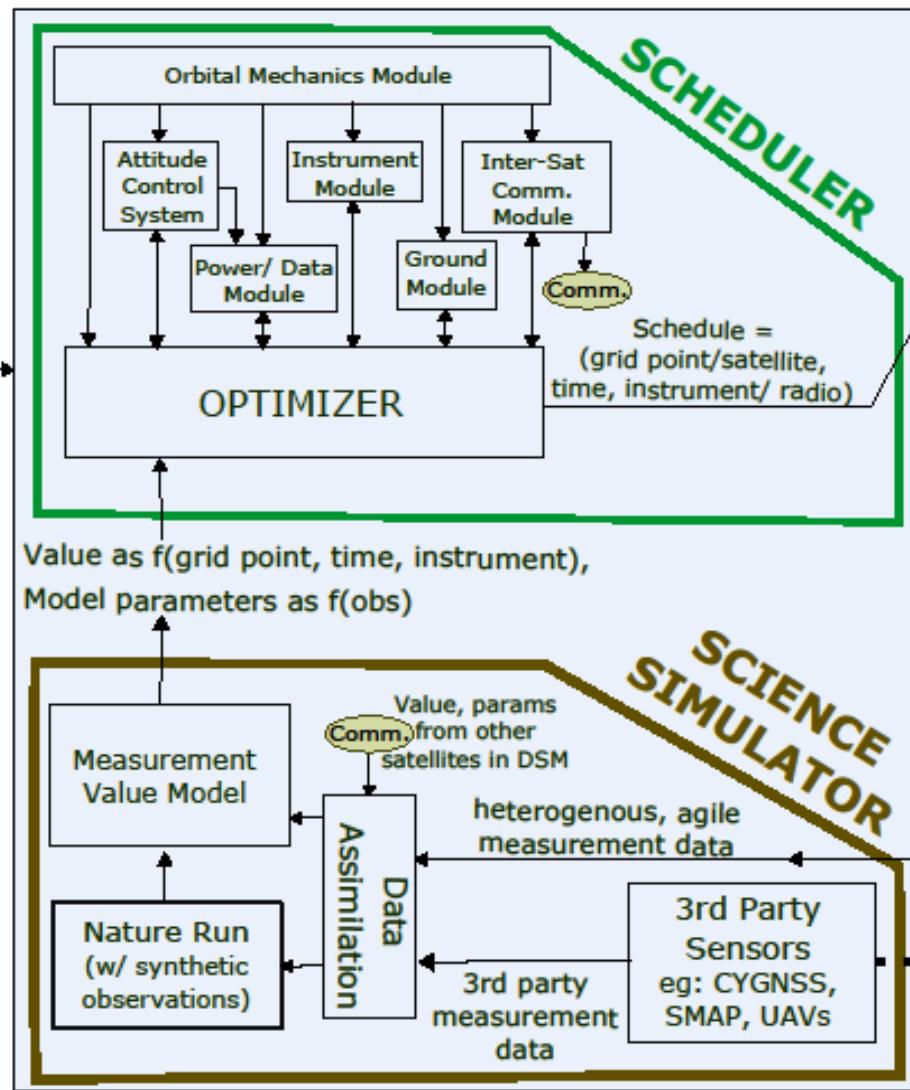
D-SHIELD Proposal

ANALYZER (Input)



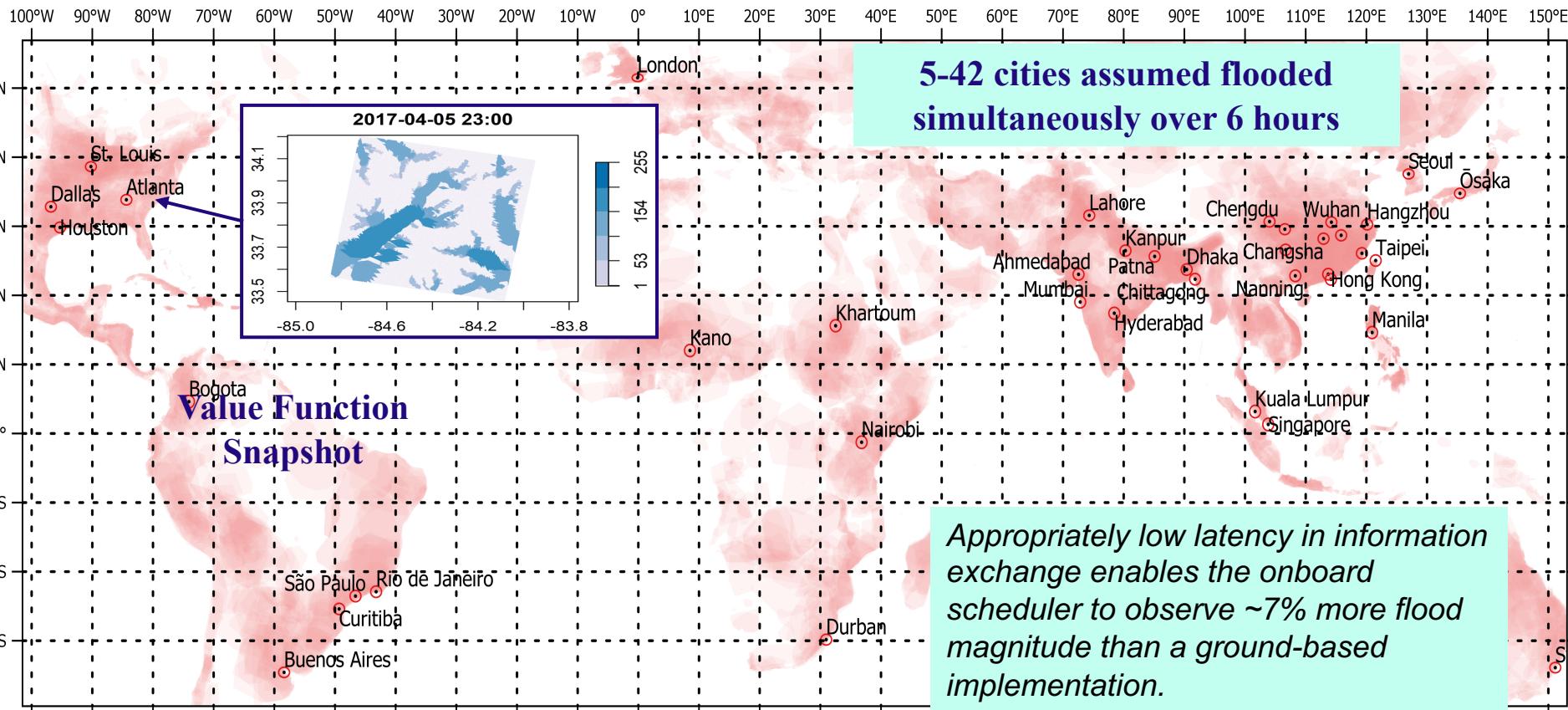
LEGEND

- Measurements
- Uni-directional Data/ Information flow
- Bi-directional Data/ Information flow
- Connector





Science Relevancy Scenario: Urban Floods



Data: Dartmouth Flood Observatory (Brakenridge 2012)

Results: S. Nag, et al, "Autonomous Scheduling of Agile Spacecraft Constellations with Delay Tolerant Networking for Reactive Imaging", ICAPS SPARK Workshop, Berkeley CA, July 2019

Results: S. Nag, et al, "Designing a Disruption Tolerant Network for Reactive Spacecraft Constellations", AIAA ASCEND, Nov 2020



Science Relevancy Scenario: Soil Moisture

Goal: Use a combination of spaceborne radar, radiometers, reflectometers to make spatio-temporal measurements that will reduce soil moisture uncertainty

Traditional Solution: Design a single or constellation of instruments (size, altitude) to address spatio-temporal trade-offs (underscored in conflict with all others)

Radiometric:
Noise sigma
Speckle K_p

Spatial Metrics:
Resolution => Static Uncertainty
Coverage => Global Understanding

Temporal Metrics:
Revisit => Dynamic Uncertainty
Revisit => Global Understanding

Baseline = SMAP Conical Scanning:

-30dB sigNEZ ; 450m along track (AT) resolution ; 3 day global coverage+revisit

Alternative = Science-based Intelligent Planning of Stripmap SAR:

-30dB sigNEZ ; optimized* spatial resolution at the cost of speckle, coverage, revisit ~ to be addressed by more looks + measurements using constellation + intelligent agility

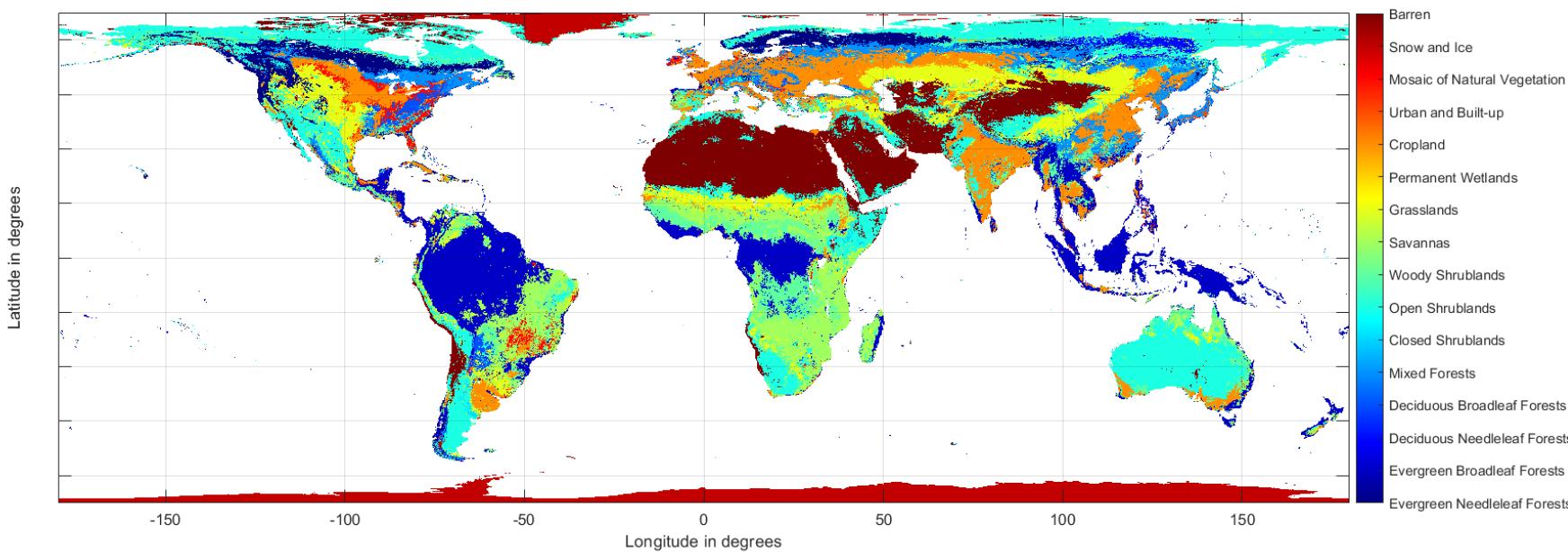
* ~7m AT and >250m CT resolution

Goal: Measure to reduce Soil Moisture Uncertainties

Sources of variation over the global 9km tile grid:

1. Soil type and vegetation
2. Season and solar conditions
3. Precipitation
4. Saturation of Soil

International Geosphere–Biosphere Programme (IGBP) 16 classes distilled into 5 relevant for Soil Moisture: Forest, Shrubland, Cropland, Grassland, Bare



Ignoring water, wetland, urban, frozen

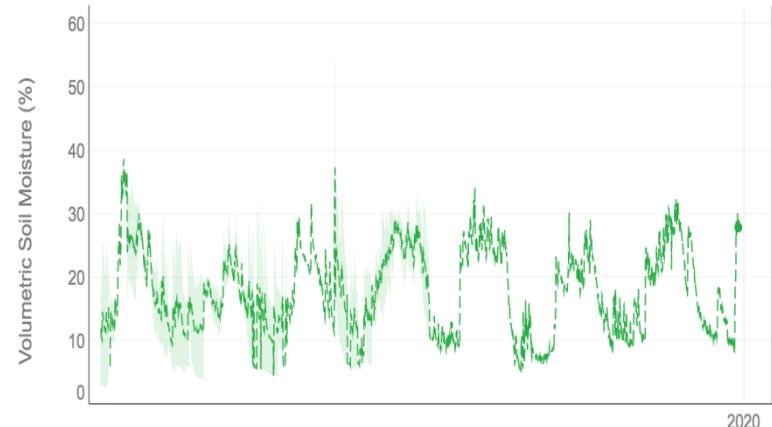
Goal: Measure to reduce Soil Moisture Uncertainties

Sources of variation over the global 9km tile grid:

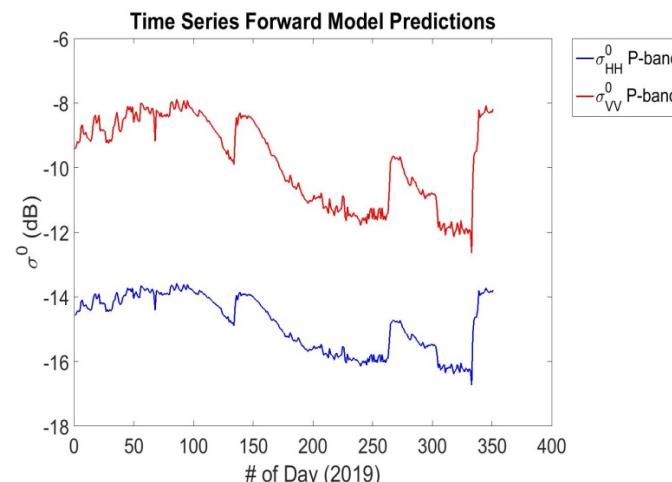
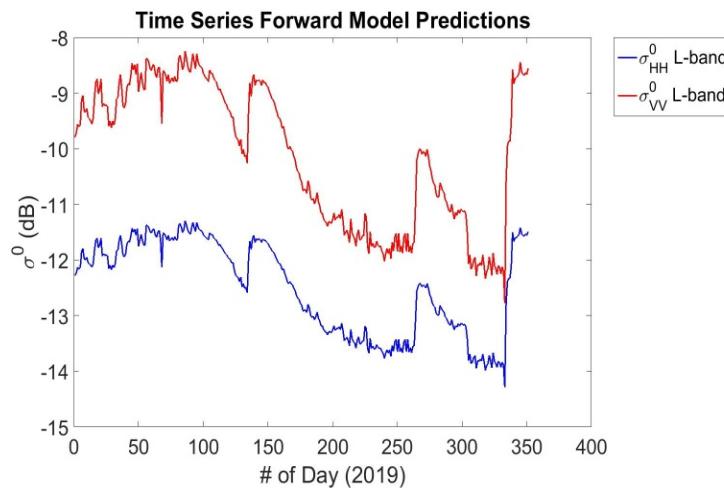
1. Soil type and vegetation
2. Season and solar conditions
3. Precipitation
4. Saturation of Soil

Will be accounted for in the speckle noise model of the science simulator

Time series at SoilSCAPE Node ID 1503 (Point)



Time Series radar cross section (RCS) prediction for Walnut Gulch at L:1.57GHz, P:430MHz, VWC = 0.29kg/m², 40deg incidence, 0.02m roughness



Goal: Measure to reduce Soil Moisture Uncertainties

Sources of variation over the global 9km tile grid:

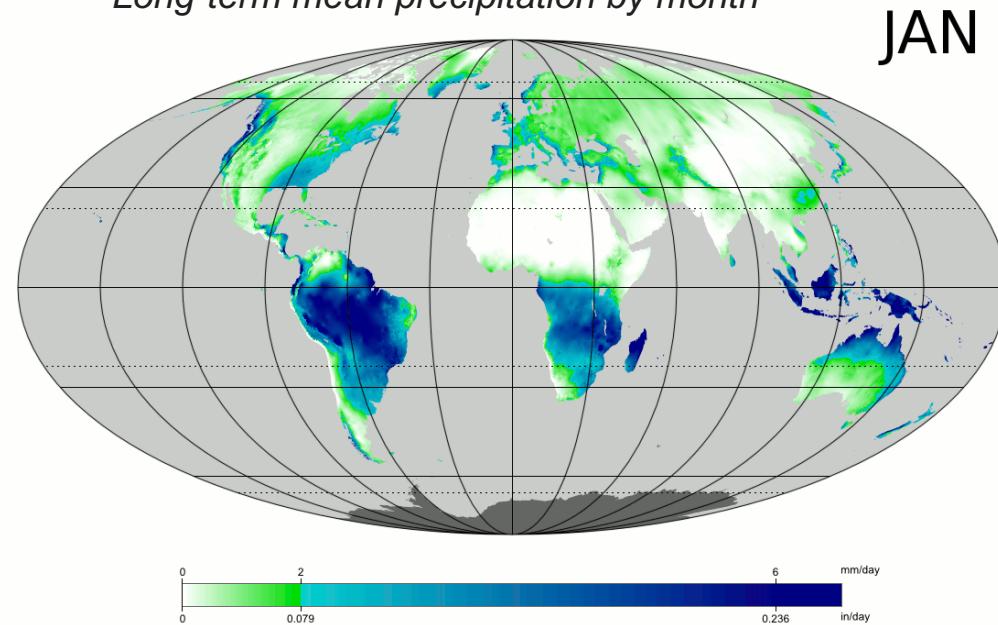
1. Soil type and vegetation
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Hourly precipitation forecast from GEOS FP in Cubed-sphere grid C720 resolution (12 km) and ~30km lat-lon. Using PRECTOT - Total precipitation ($\text{kg m}^{-2} \text{s}^{-1}$) ...

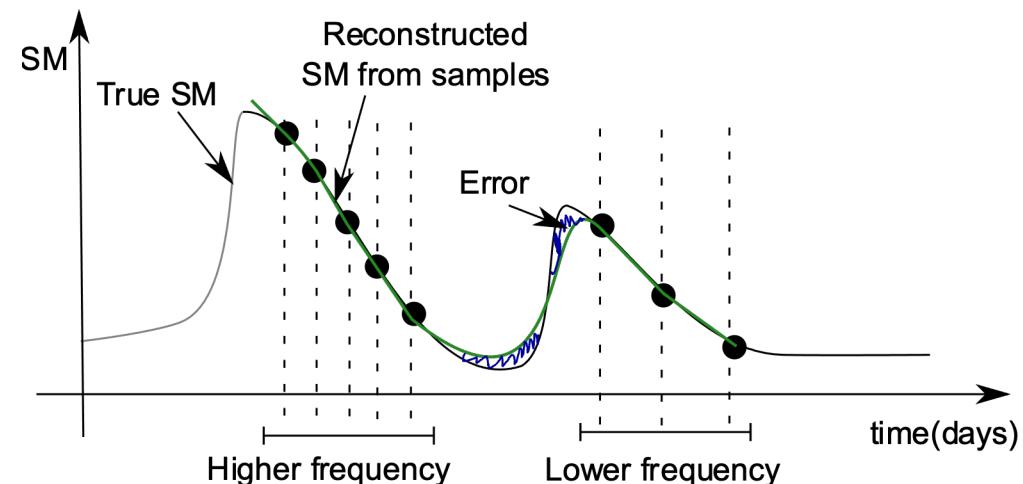
SMAP saturated pixel product globally available every 3 days. Interesting pixels are those that are not saturated and there has been rain recently...

Long-term mean precipitation by month



Addressing Temporal Resolution / Science Needs

Temporally close measurements (just as neighboring pixels) can be combined to reduce speckle noise. Use inverse modeling to find maximum ΔT up to which SM dynamism does not prevent meaningful integration => $\Delta T=2\text{hours}$

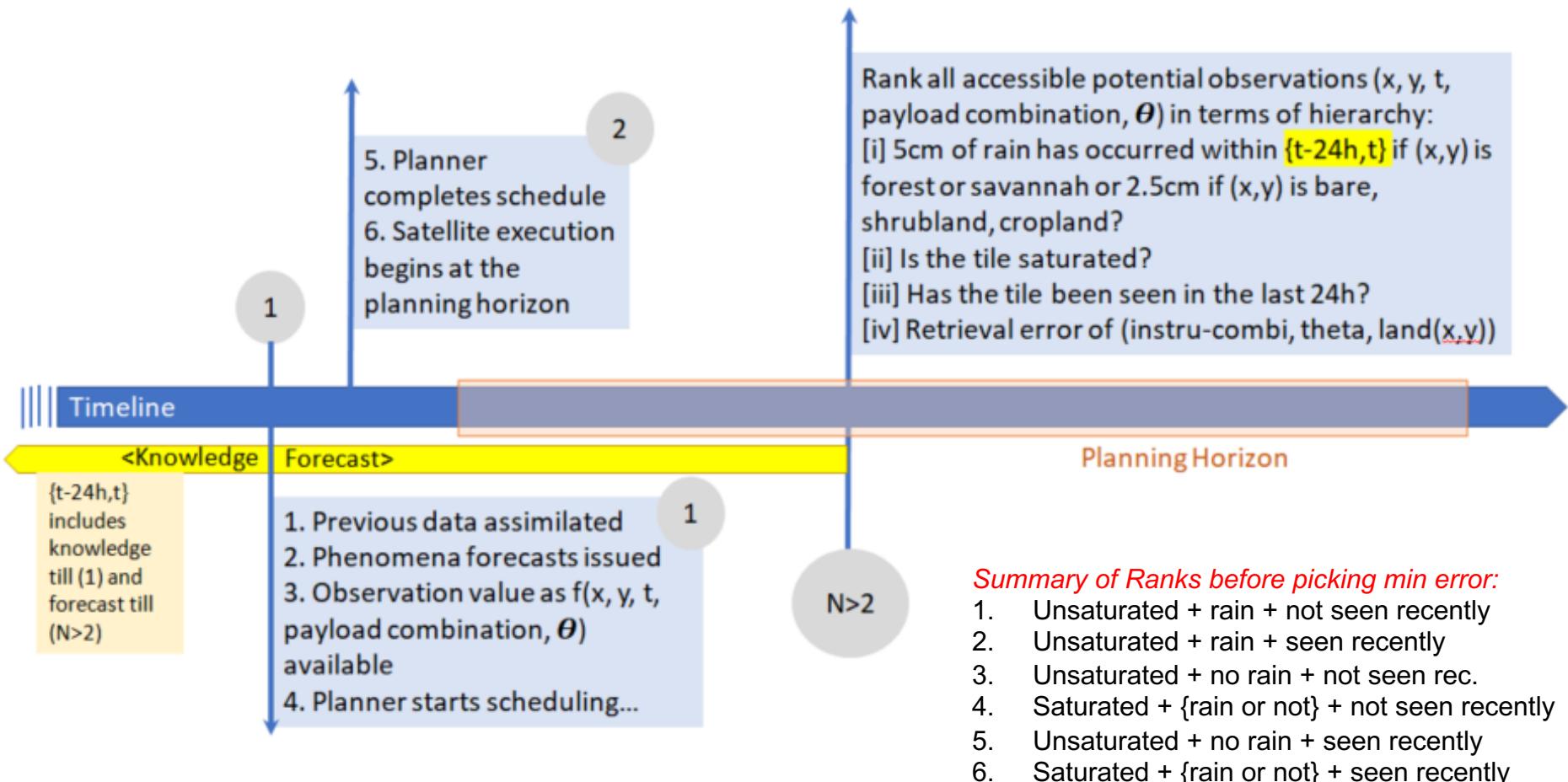


| Shrubland, wet period | | | | | | | | |
|-----------------------|-------------------------|-------|-------|-------|-------|--------|--------|--------|
| Table B Coding: | | Sat#1 | Pay#1 | Sat#1 | Pay#2 | Sat#2 | Pay#2 | M.E.SM |
| Code | Meaning | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 0 | No operation | 0 | 0 | 0 | 0 | 1 | 0.0039 | |
| 1 | 35+/-5 deg inc, 1 obsvs | 0 | 0 | 0 | 0 | 2 | 0.0048 | |
| 2 | 45+/-5 deg inc, 1 obsvs | 0 | 0 | 0 | 0 | 3 | 0.032 | |
| 3 | 55+/-5 deg inc, 1 obsvs | 0 | 0 | 0 | 0 | 4 | 0.0038 | |
| 4 | 35+/-5 deg inc, 2 obsvs | 0 | 0 | 0 | 0 | 5 | 0.0048 | |
| 5 | 45+/-5 deg inc, 2 obsvs | 0 | 0 | 0 | 0 | 6 | 0.0319 | |
| 6 | 55+/-5 deg inc, 2 obsvs | 0 | 0 | 1 | 1 | 0.0038 | | |
| | | 0 | 0 | 1 | 2 | 0.0041 | | |
| | | 0 | 0 | 1 | 3 | 0.0161 | | |
| | | 0 | 0 | 1 | 4 | 0.0038 | | |

1000+ rows of combinatorics for 2 sats

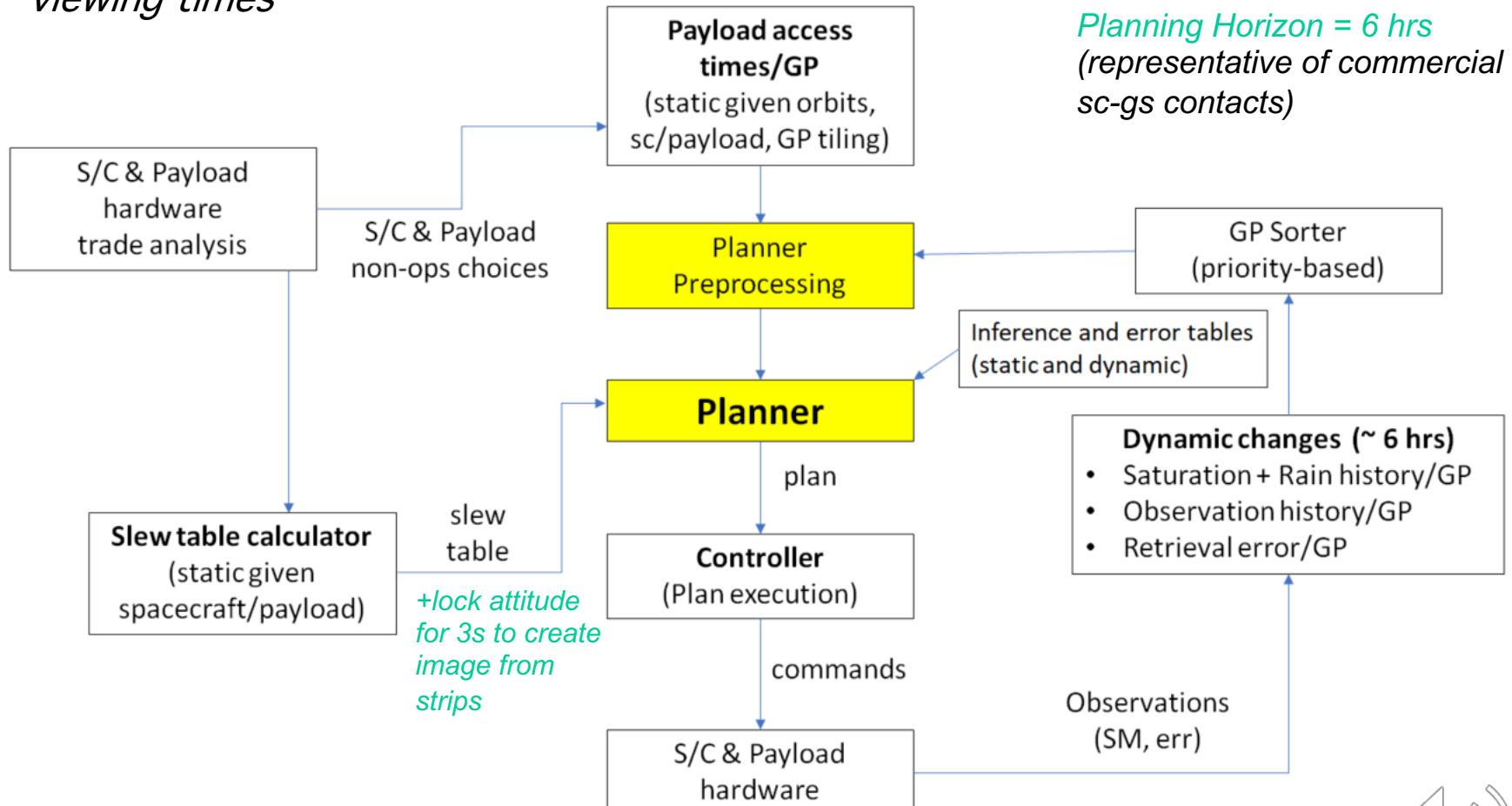
Concept of Operations

Preliminary rules as a strawman for the science simulator:



Observation Planner: Optimizing the Schedule

Planner-centric View to *decide* what to look at, *when* to look at it and *how* to look at it i.e. Choose command <instrument, viewing angle> *for all available viewing times*





Planner: Optimizing the Observation Schedule

Search space size:

- 24 hours (4 x 6-hour plans), 1-s increments (86.4k s)
- 2 instruments (L-band, P-band)
- 62 viewing angles/instrument
- 41,500 Access Time Points (TP)
- 1,662,486 Ground Positions (GP)

Pre-processing for choice flattening (reduces space by 65%)

Uses Constraint Satisfaction Problem (CSP) Algorithm to find solution

Timepoint (TP) choices:

| Time | Inst/angle | GP | science value |
|-------|--------------------|----|---------------|
| 1311: | L.32: [3165] | | score: 0.925 |
| | L.34: [3445, 3446] | | score: 0.925 |
| | P.32: [3165] | | score: 0.9 |
| | P.33: [3165] | | score: 0.9 |
| | P.34: [3445, 3446] | | score: 0.9 |
| | P.35: [3445, 3446] | | score: 0.9 |

Gridpoint (GP) choices:

| GP | Time | Inst/angle(s) | # obs | science value |
|-------|--------|---------------|--------|---------------|
| 3165: | 1311: | [L.32] | 1 obs, | score: 0.925 |
| | 1311: | [P.32] | 1 obs, | score: 0.9 |
| | 1311: | [L.32, P.32] | 2 obs, | score: 0.9 |
| | 12597: | [L.33] | 1 obs, | score: 0.575 |
| | 12597: | [P.33] | 1 obs, | score: 0.2 |
| | 12597: | [L.33, P.33] | 2 obs, | score: 0.2 |

sciencevalue= 1-retrievalerror/0.04 (after ranking for seen, rain, saturation)

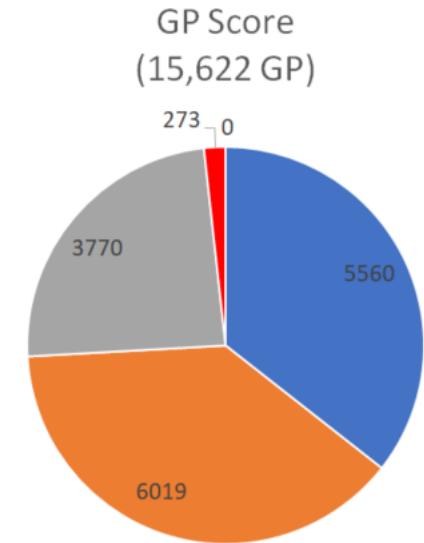
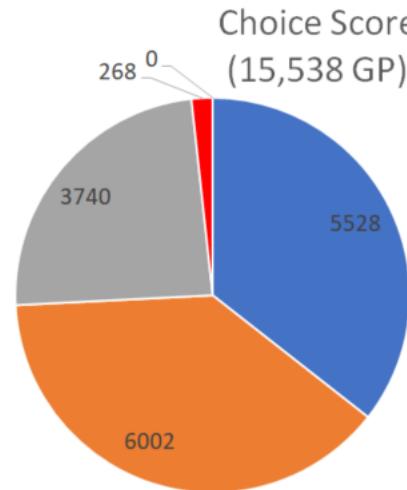
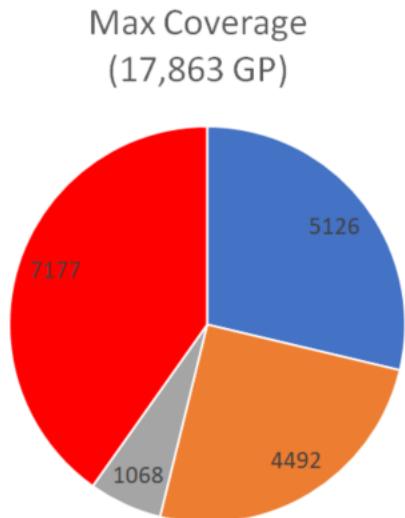
Planner: Optimizing the Observation Schedule

Local and Global Heuristics are ongoing topics of research:

1. Max Coverage maximizes number of GPs seen but does not use science value
2. Choice Score maximizes science value without accounting for GPs seen
3. GPscore maximizes product of GPs and science value (*current POR*)
4. Other options: max GP choice rank, max RareGP (TBD with improved science simulator)

For the first horizon of 6 hours by a single satellite

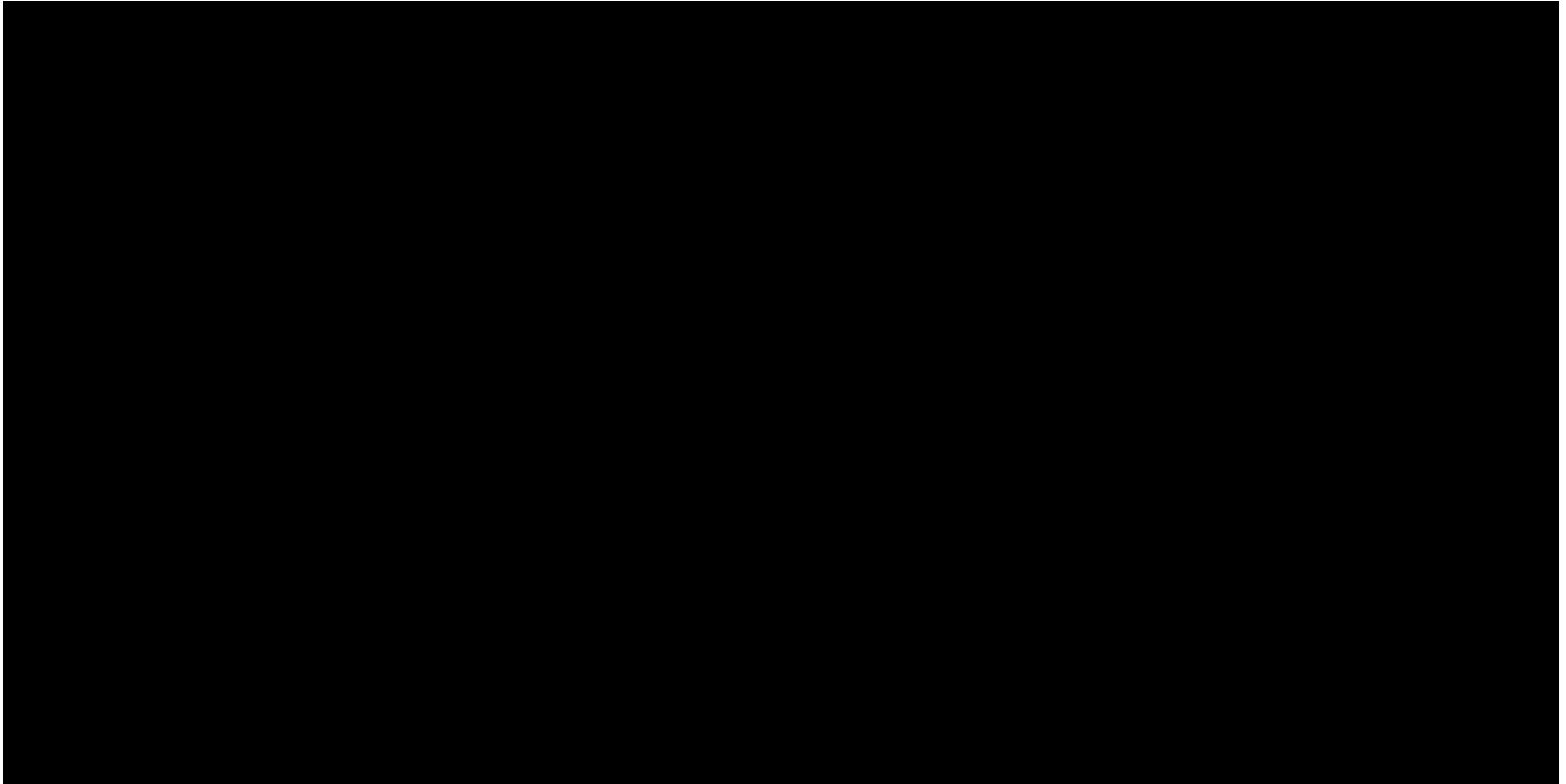
Distribution of GP scores for 3 different heuristics



Sciencevalue = $1 - \text{retrievalerror}/0.04$ as a % ■ 90-100 ■ 80-89 ■ 50-59 ■ 20-29



Planner: Optimizing the Observation Schedule



Over 24 hours by single sat:

Interesting land cover GPs = 1.662m

Rainy, unsat. GPs = 307.9k-309.8k

Total observed GPs = 53.4k (3.2%)

Rainy, unsat. observed GPs = 15.6k (~5%)

For 1 horizon of 6h:

Interesting land cover GPs = 637k

9.8k variables, 3.8mins to solve

Adding all constraints and heuristics

~16k GP, 3.2k variables, 43s to solve

Very prelim Planner: Single Sat has 15% SMAP temporal coverage at 60x AT spatial resolution





Thank you!

Questions? Sreeja.Nag@nasa.gov

Publicly available Source Code and Docs:

The screenshot shows the GitHub interface with several repositories listed:

- EarthObservationSimulator**: A repository with a 3D cube icon.
- orbitpy**: A repository with 3 C++ files, Apache-2.0 license, 0 stars, 5 issues, 0 pull requests, and 0 updates.
- instrupy**: A repository with 1 Python file, Apache-2.0 license, 0 stars, 1 issue, 0 pull requests, and 0 updates.
- eosim**: A repository with 1 Python file, Apache-2.0 license, 0 stars, 0 issues, 0 pull requests, and 0 updates.

The screenshot shows two documentation pages side-by-side:

OrbitPy Documentation (https://orbitpy.readthedocs.io/en/latest/)

- Welcome to OrbitPy's documentation!**
- Install**
 - Requires: Unix-like operating system (currently tested in Ubuntu 18.04.03, MacOS, CentOS-8), `python 3.8`, `pip`, `gcc`, `gfortran`
 - 1. Make sure the `instrupy` package (dependency) has been installed. It can be installed by running `make` in the `instruments/instrupy/` directory.
 - 2. Navigate to the `orbits/oc/` directory and run `make`.
 - 3. Navigate to the `orbits/orbitpy/` directory and run `make`. `runtest` to run tests and verify OK message.
- InstruPy Documentation (https://instrupy.readthedocs.io/en/latest/)**
- Welcome to InstruPy's documentation!**
- Instrument specifications** → **InstruPy** → **Typical data metrics during the access**
- Satellite states during access** → **InstruPy**
- The high-level function of the InstruPy package is shown in the figure. The package ingests access data and satellite states (time, position, velocity) information, and instrument specifications (for which the access data is generated), and outputs typical data metrics of observations made during the access.