LINCC - developing software for large-scale analysis of time domain data

LINCC







The LINCC Frameworks Project

LSST Interdisciplinary Network For Collaboration And Computing

- A collaboration between UW, CMU, LSSTC, U Pitt, and NOIRLab to build software systems for key LSST science
- PIs: Andy Connolly (UW), Rachel Mandelbaum (CMU)
- Director of Engineering: Jeremy Kubica (CMU)
- Science software infrastructure: combining user algorithms & code, astro packages, and industry tools to build scalable science analysis packages







LINCC Frameworks mission is to enable scientists by developing scalable and productionised software/algorithms in collaboration with broader community.

We want to:

- be engineering and algorithmically focused,
- collaborate with other software efforts (projects may be contributions to existing code bases),
- leverage existing tools (build on top of the Rubin Science Platform and standard community tools/libraries), and
- coordinate with community to avoid unnecessary duplication of effort.



Additional venues for collaboration:

- Workshops Work with LSST Science Collaborations to identify areas of need.
 - Data to Software to Science Workshop (March 2022)
- **Incubators** Scientists work with team to get their science applications working (open proposal process).
- **Tech talks** showcase work done by the broad Rubin software and archives community that's designed to enable LSST science
- LINCC Frameworks members joining LSST science collaborations.

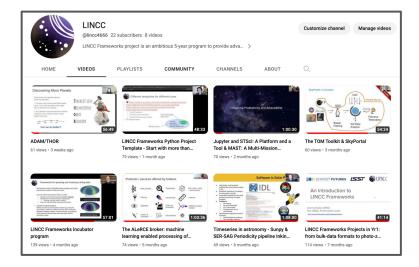


- Incubators provide support for researchers to work directly with LINCC Frameworks team to apply new tools to research problems.
- Goal: Establish long-term software development collaborations that serve both the selected teams and LINCC Frameworks.
- 2nd proposal deadline was June 15, 2023, next one is expected October
- Incubator #1 Solar System Simulation
 - Meg Schwamb at Queen's University Belfast

• Incubator #2 - Supernova Classification

• Kaylee De Soto at Penn State (Session 3A)







• Talks that showcase the work done by the broad Rubin software and archives community.

2nd Thursday of the month, 10 am Pacific at https://ls.st/lincc-talks

- So far: Brokers, in-kind contributors, data centers, LINCC engineers...
- Future: Addition of talks by Roman software group & results from incubators



- Astropy contributions (<u>multiband periodograms</u>, implemented in the latest Astropy 5.13)
- Tape (Timeseries Analysis & Processing Engine)
 - Enable scaling of external functions (via dask)
 - Internal functions filling the gaps that currently exist
 - E.g., <u>structure function calculations</u>
 - Widely useful algorithms
 - E.g., color correction, custom ``ubercal`` of individual objects
 - Aggregating **brokers** results
- <u>LSDB development</u> (how to crossmatch, work with large catalogs, integrate with time domain)



• This decade is marked by many projects producing large datasets. Rubin's Year #1 dataset will be O(100TB) in size.

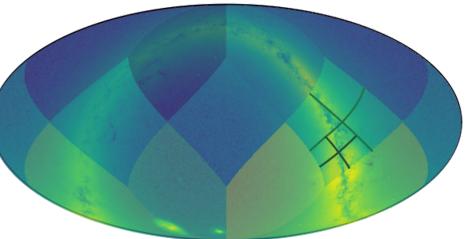


- A major use case for these data is whole-dataset science (statistics, searching, mapping). Examples: variable star classification, time-domain analyses, dust distribution, Milky Way, large-scale structure, etc.
- Few ready-to-use tools exist today for such work, and the way data are distributed hinder it. Barrier to entry is very high.
- Objectives: Develop tools that enable scientists to conduct large-scale analytics on multiple datasets (LSDB). Develop formats that enable dataset providers to expose their dataset to such analyses (HiPSCat).



- A hierarchical data storage scheme, where the sky is hierarchically split into HEALPix tiles until each tile has roughly a
- similar number of objects (rows).
- These tiles are stored as Parquet files within a directory tree that encodes their location on the sky.

Enables: fast spatial lookup, distributed analytics, distributed joining and cross-matching. Based on 10+ yrs of thinking/experience/experimentation (LSD + AXS).



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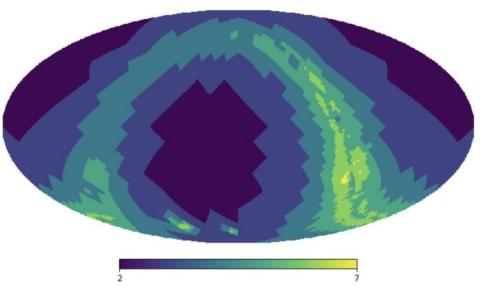
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Details: https://dirac.us/teg

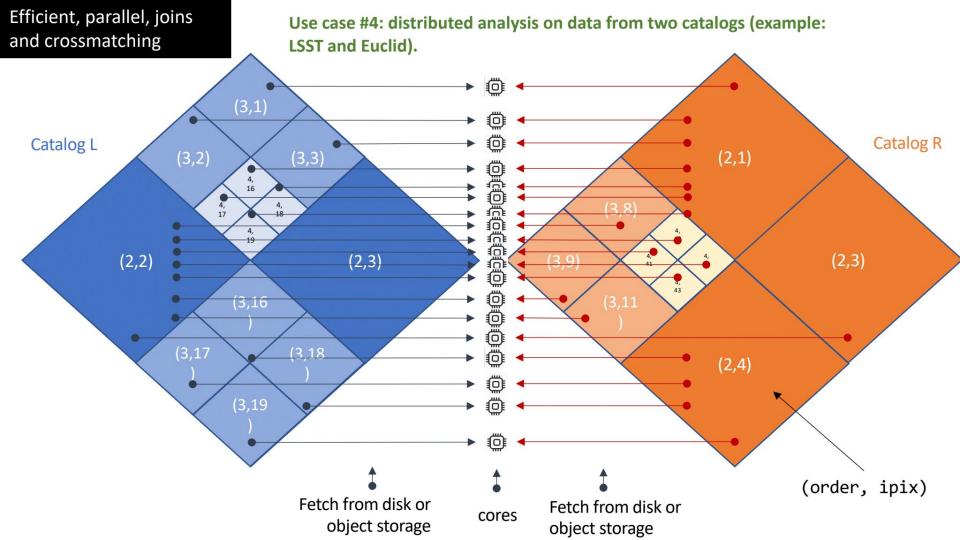


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The Tool: Large Survey Database (LSDB)

- HiPSCat is a valid Parqued (partitioned) dataset. Parquet tools can use it out-of-the-box.
- But a tool with full HiPSCat awareness can enable spatial queries, cross-matching, timeseries, and efficient multi-dataset joining. **Enter LSDB.**
- LSDB: Pandas-like analysis of astronomical datasets with trillions of observations using thousands of cores.
- Build on existing tools and ecosystem. Presently Dask (may look at Ray).

```
img = gaia
    .query("pm > 10")
    .crossmatch(ztf)
    .join(ztf_sources)
    .for_each(varstar_classify)
    .query("pRRLy > 0.95")
    .skymap()
hp.mollview(img)
```



+ New item from the last few

months

= Existing item

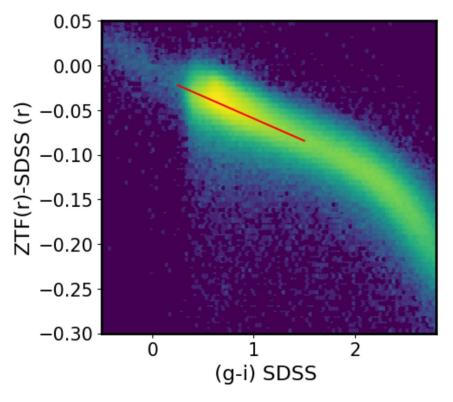
* Work underway

Software Aspects			
API	Infrastructure	Performance/Scalability	
 + Object/Source Refactor + Loading from parquet files, HiPSCat directories*, and dictionaries + Filtering Operations (dropna, prune) + Source Binning, Insertion 	 + Uses Python Project Template + Tutorial/Documentation on readthedocs = Unit Testing, 85% coverage (and rising!) 	+ Have run O(million) ZTF lightcurves through analysis suite functions	

Scientific Aspects		
Analysis Suite	Scientific Usage/Testing	Community Interaction
 Stetson J Function + Structure Function Refactor + Custom-User Analysis Functions 	+* Time Domain Minimal Viable Product +* TinyGP/JAX Code	=* Fast Periodic Detection (Dr. Tansu Daylan) =* light-curve (Konstantin Malanchev)



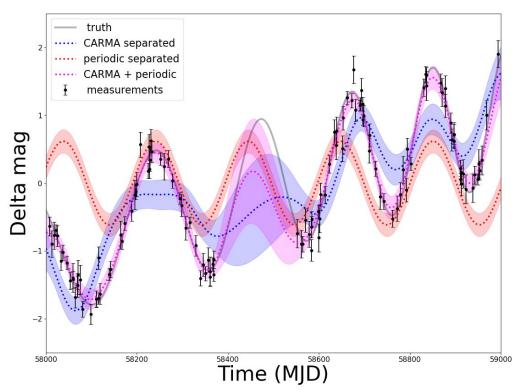
Examples: color-correction



- Enable analysis of long light curves
- Primarily by cross-matching and ``re-normalizing" PanSTARSS, ZTF + (SDSS)
- Enable estimation of custom color correction and/or apply already estimated factors from literature
- Enable estimation of errors from the spread of stars of similar magnitude/color
- Your input appreciated

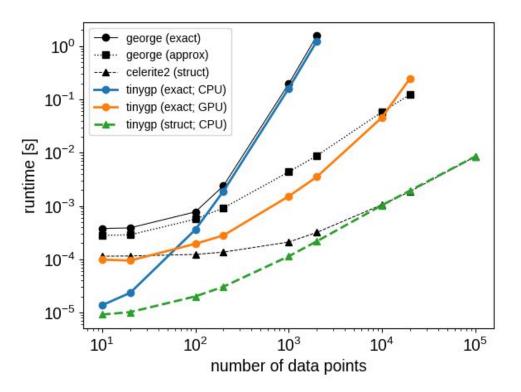


- JAX just in time compilation & autograd (e.g., plenary talk by Dan-Foreman Mackey at AAS)
- Enables fast Gaussian process estimation
- Compilation per given length of lightcurve
 - when operating on large number of lightcurves, we want to send all objects of same length to a single dask worker to avoid recompilation





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- <u>LINCC</u> building software systems for key LSST science
- Supporting community through <u>incubators</u>, <u>workshops</u>, <u>talks</u>
- <u>LSDB</u> crossmatching library, via HiPSCat and parquet files
- <u>Time domain</u>
 - Astropy contributions
 - Scaling up codes (via dask)
 - Functions useful for community



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