

Discovering undiscoverable asteroids

The power of DECam, its archive, and novel algorithms

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DATA INTENSIVE RESEARCH IN
ASTROPHYSICS AND COSMOLOGY

**ASTEROID
INSTITUTE**

A PROGRAM OF B612

Solar System @ UW



Joachim Moeyens,
UW grad
THOR algorithm



Tom Wagg,
Hybrid catalogs & NEO
estimates



Zach Langford,
UW undergrad (-> U. Penn)



Bryce Kalmbach
LSST postdoc @ UW



Ari Heinze
HelioLINC+ codes



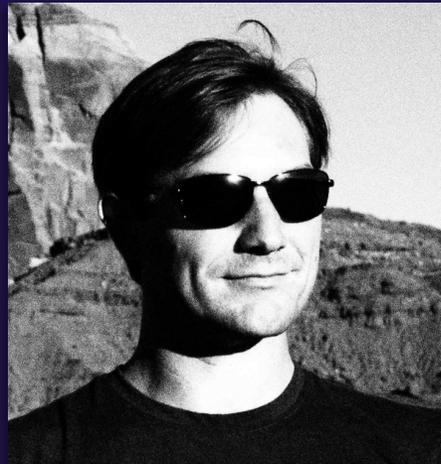
Pedro Bernardinelli,
2021 DiRAC
Postdoctoral Fellow



Stephen Portillo, 2018 DiRAC Fellow
Barycentric KBMOD stacking
-> Concordia U. Prof.



Lynne Jones et al.
Rubin Performance Scientist



Siegfried Eggl, HelioLinc 3D
UW -> UIUC Prof.



Steven Stetzler
UW grad
DECat Processing & sub-
threshold detection
algorithms



Aidan Berres
UW undergrad (-> UIUC grad)



Petter Whidden, Hayden Smotherman
-> NY Times -> Northrop Grumman



Maria Chernyavskaya
UW undergrad (-> NAU grad)



Sarah Greenstreet, DiRAC Scientist
Lead of the SSSC NEO+ISO WG
UW -> NOIRLab

+ B612 Asteroid Institute collaborators ...



21 Lutetia - 132 × 101 × 76 km
Rosetta, 2010



Dactyl
([243] Ida I)
1.6 × 1.2 km
Galileo, 1993

243 Ida - 58.8 × 25.4 × 18.6 km
Galileo, 1993

25143 Itokawa
0.5 × 0.3 × 0.2 km
Hayabusa, 2005



2867 Steins
5.9 × 4.0 km
Rosetta, 2008



4179 Toutatis
4.6 × 2.3 × 1.9 km
Chang'E 2, 2012



5535 Annerfrank
6.6 × 5.0 × 3.4 km
Stardust, 2002

9969 Braille
2.1 × 1 × 1 km
Deep Space 1, 1999



951 Gaspra
18.2 × 10.5 × 8.9 km
Galileo, 1991



433 Eros - 33 × 13 km
NEAR, 2000



253 Mathilde - 66 × 48 × 44 km
NEAR, 1997

1P/Halley - 16 × 8 × 8 km
Vega 2, 1986

19P/Borrelly
8 × 4 km
Deep Space 1, 2001

9P/Tempel 1
7.6 × 4.9 km
Deep Impact, 2005

81P/Wild 2
5.5 × 4.0 × 3.3 km
Stardust, 2004

103P/Hartley 2
2.2 × 0.5 km
Deep Impact/EPOLI, 2010

Finding Asteroids

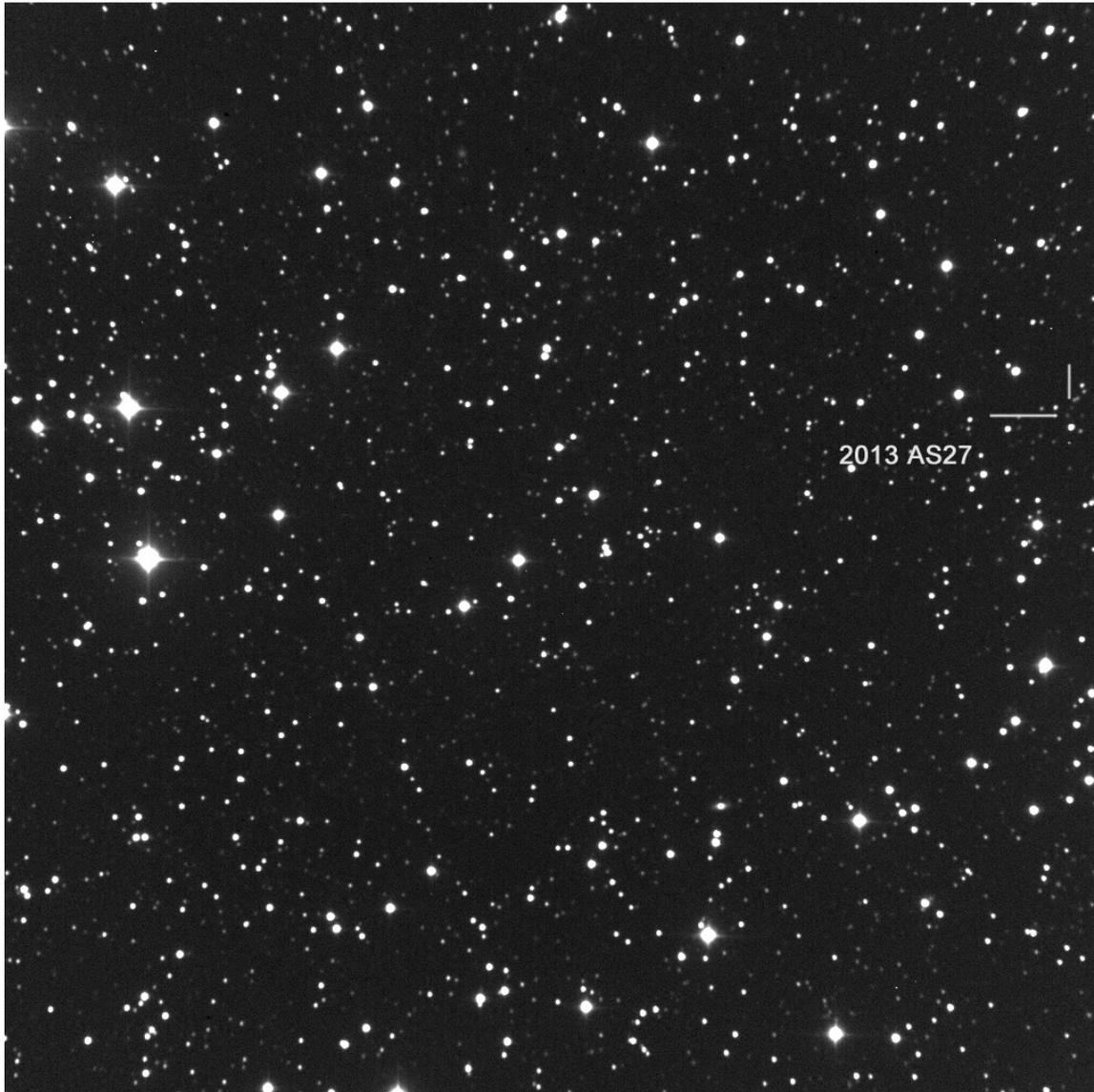


Credit: a sequence of images of asteroid (1078) Mentha by the UK Spaceguard Centre (<https://spaceguardcentre.com>)

Identifying “tracklets”

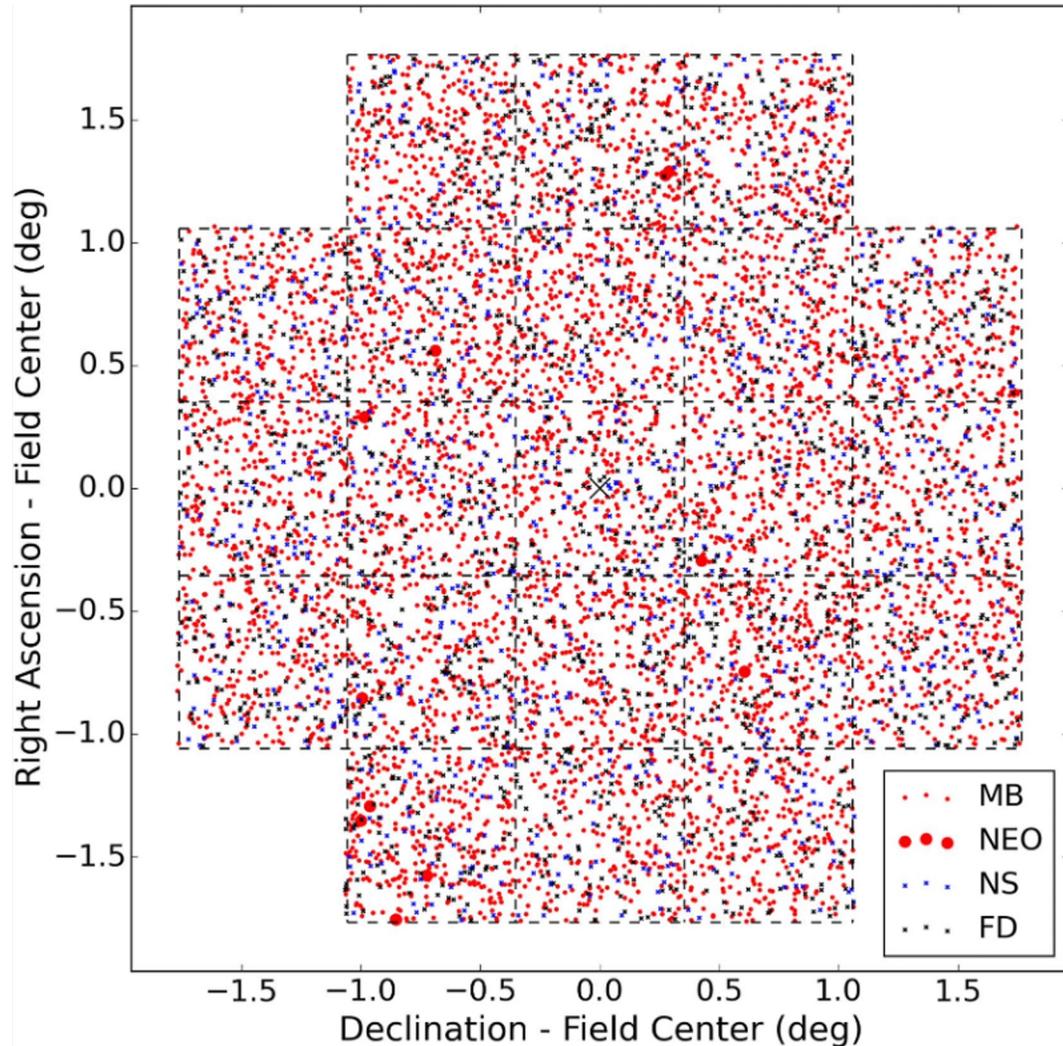
In shallow imaging, two closely-spaced images can be sufficient to unambiguously identify the asteroid.

More observations are needed to determine the orbit; the asteroid should typically be re-observed one or two nights later.

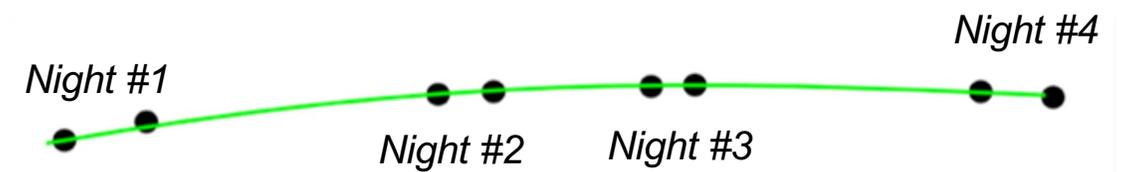


*Photo credit: Gary Hug via the Planetary Society
(<http://www.planetary.org/explore/projects/neo-grants/images/Hug-2013AS27-Discovery-Images.html>)*

DEEP SURVEYS: FULL OF ASTEROIDS!



Deep sky images have high object densities. Tracklets become probabilistic.



We construct algorithms to “link” *possible* combinations from different nights.

Ari Heinze and Siegfried Eggl’s HelioliNC Advanced

Matt Holman et al. (2019)

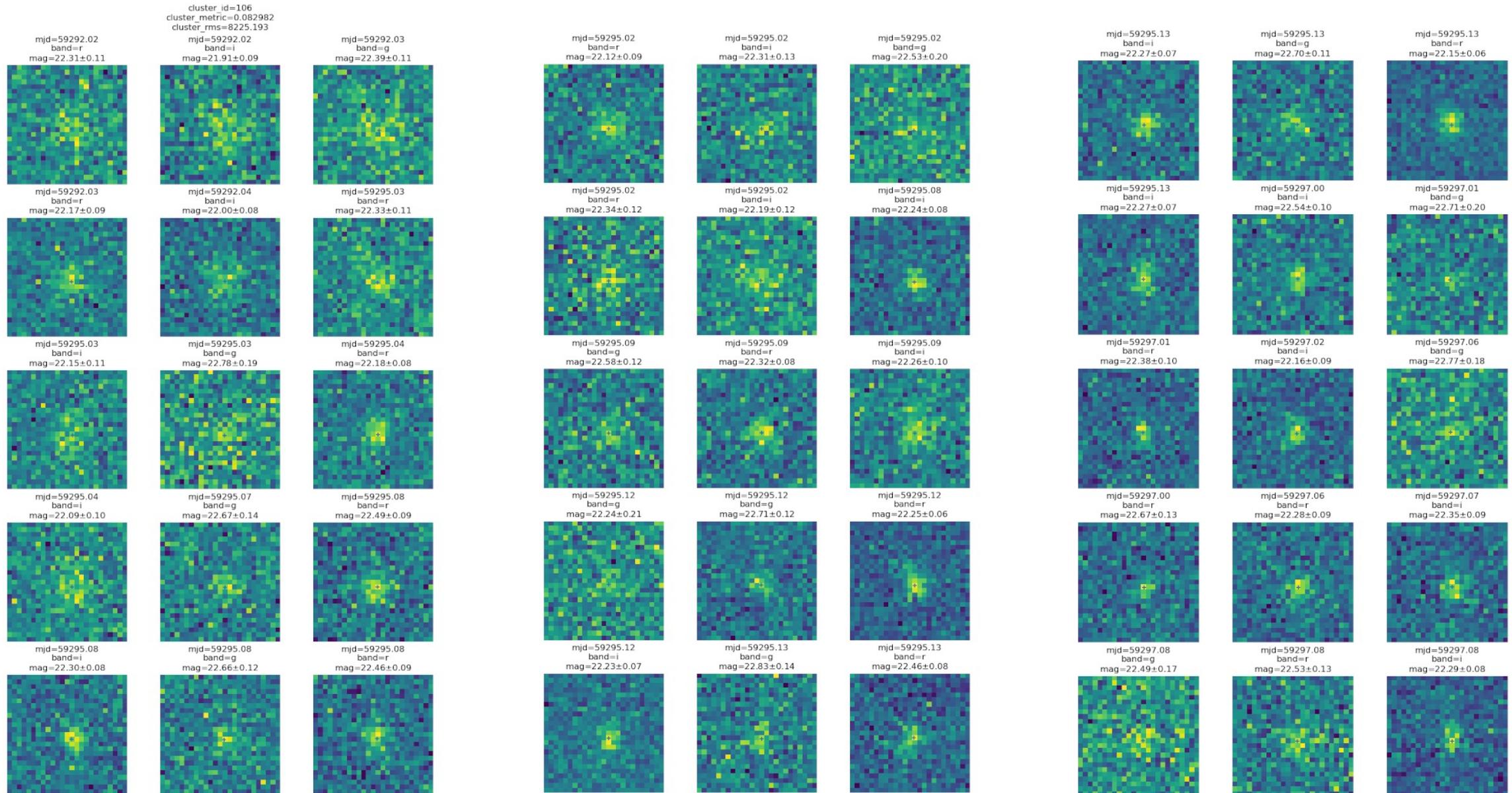
Optimized LSST Implementation: HeliLinC+



Ari Heinze
HeliLinC+ codes

- Science performance
 - Full-sky LSST tests over a two-week linking window: **97% completeness.**
 - Full-sky LSST test for interstellar objects: **96% completeness.**
 - LSST NEO discovery performance: **~70% completeness (not tuned yet).**
 - In all cases, **purity >90%**, without using orbit determination χ^2 as a filter.
- Scaling and performance
 - **Scales as $O(N \log N)$** with the number of tracklets; MOPS scaled as $O(N^3)$
 - The constant factor is \sim small: a few heliocentric distances explore most of the parameter space
 - The code is completely in C++ (working on a small Python wrapper). **Fast.**
- Results: We may be able to replace a small cluster with a \sim single machine for LSST-scale workloads and we are robust to unanticipated differencing issues (!).

Running on real LSST-like data being acquired on DECam. Also tested w. ATLAS.

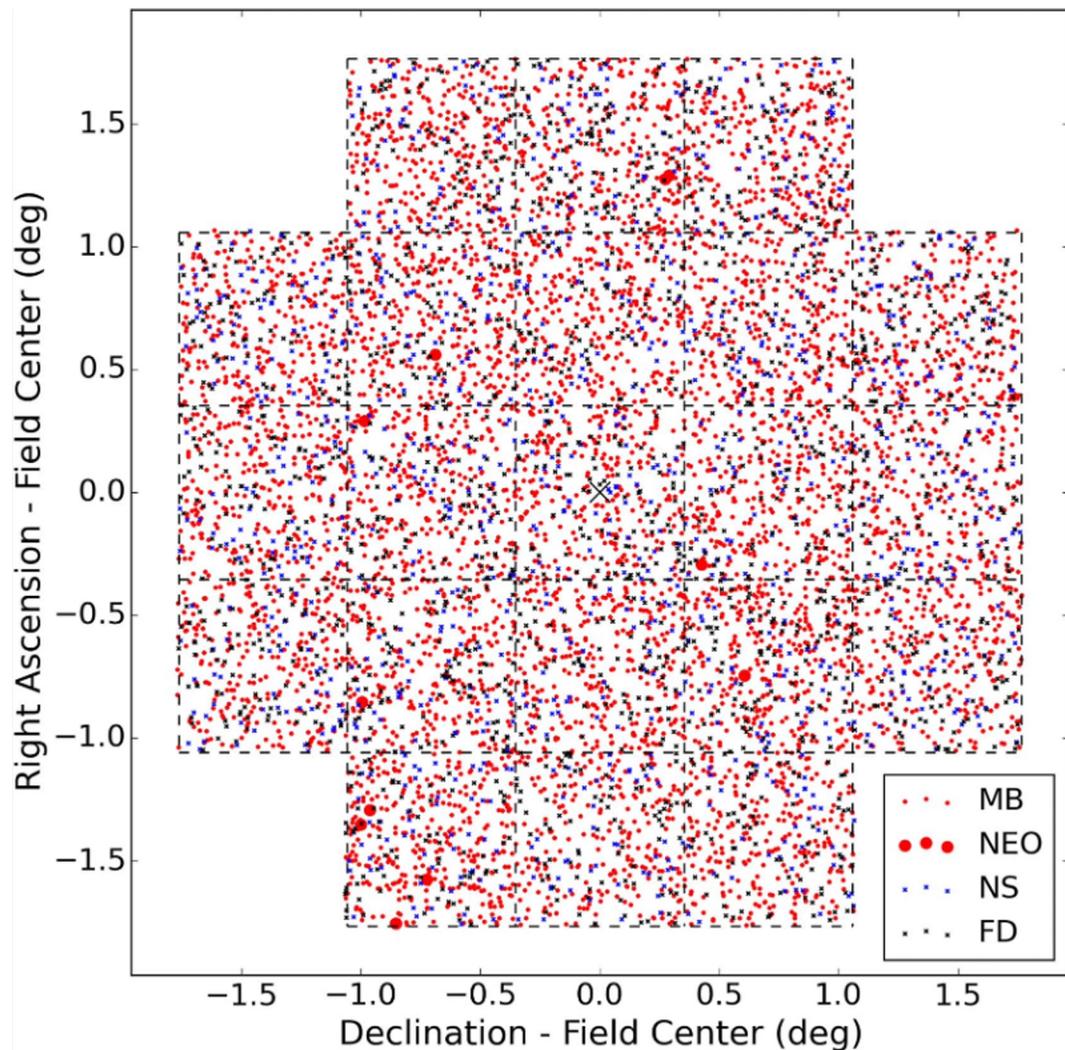


Cutouts by *Steven Stetzler (UW)*; Data: *DECam*, PI: *Melissa Graham (UW)*

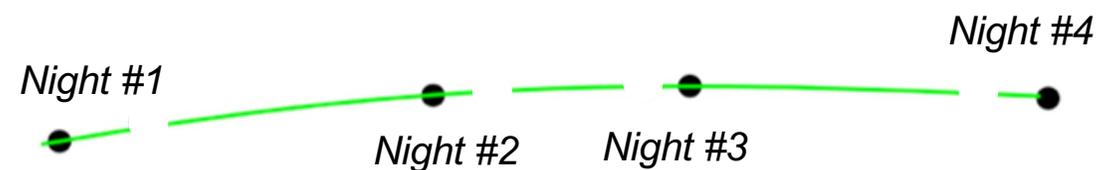


Can we do better?

CAN WE DO BETTER?



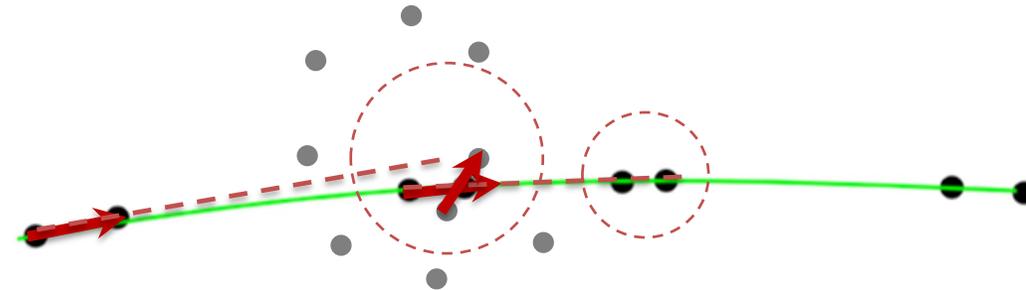
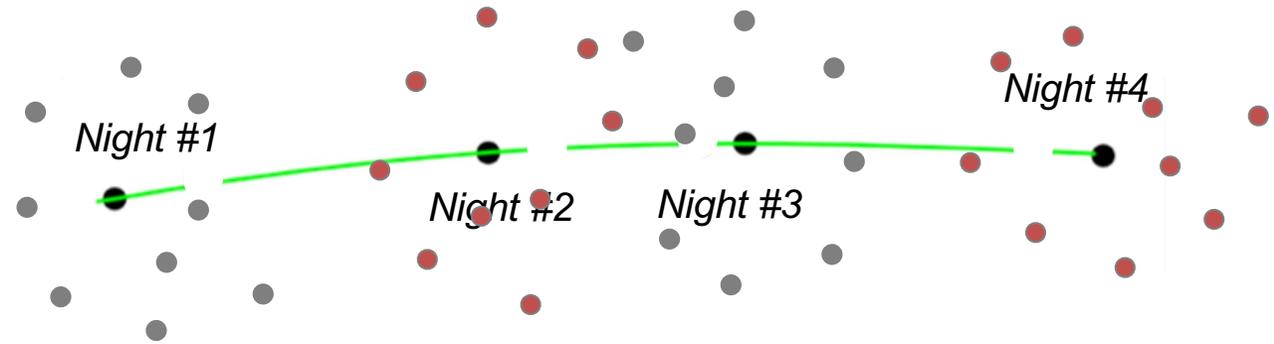
Every time we have to take a pair of images, we half the amount of sky we could observe each night...



Could we link asteroids with just a *single* observation each night?

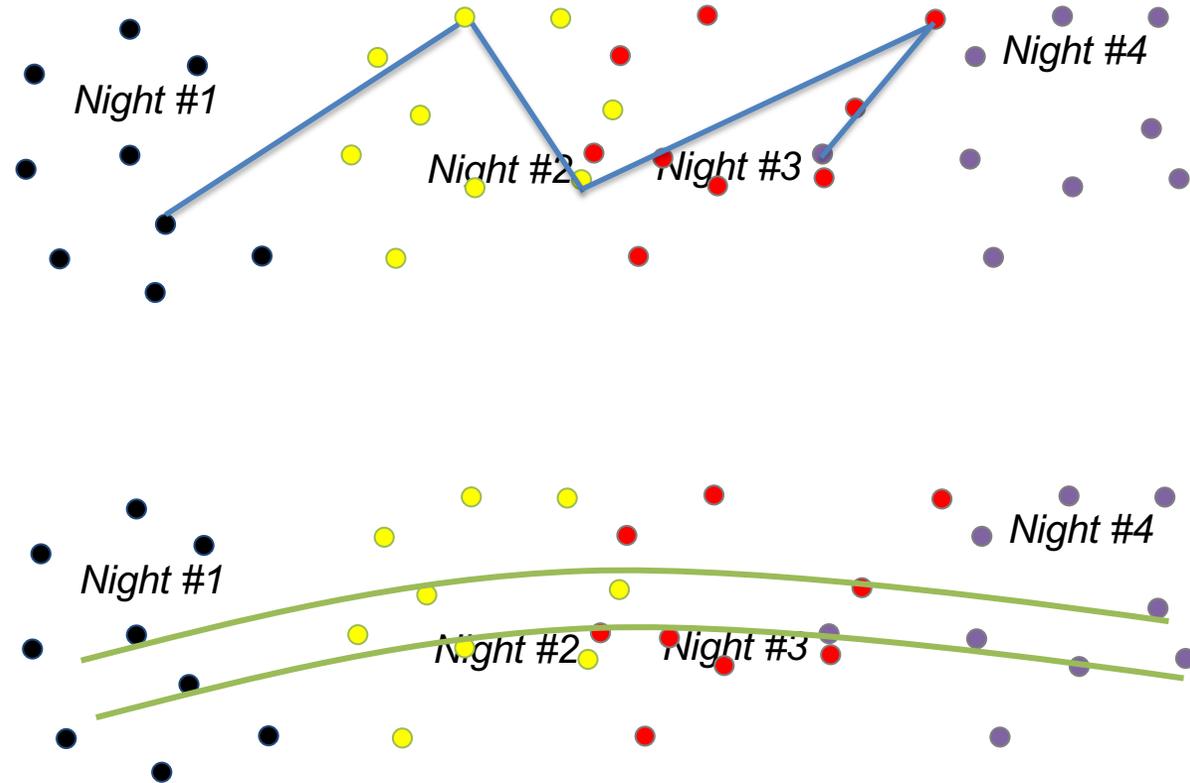
Why is “the linking problem” hard? (and how to solve it)

- If the linkage is known (the observations that go together), it’s trivial to verify it (fit an orbit).
- If the linkage is unknown, naïve algorithms would scale as $O(N^k)$, where N is the number of detections within a distance an asteroid can plausibly move to next observation (**1-6 deg for ~6 days**), and k is the number of detections needed to confirm the discovery ($k \sim 5$). Computationally tough, $O(10^{20})$ for LSST.
- “Tracklets” simplify this problem by dramatically reducing the number of choices.
- But the cost is a cadence constraint – need two observations to create tracklets.



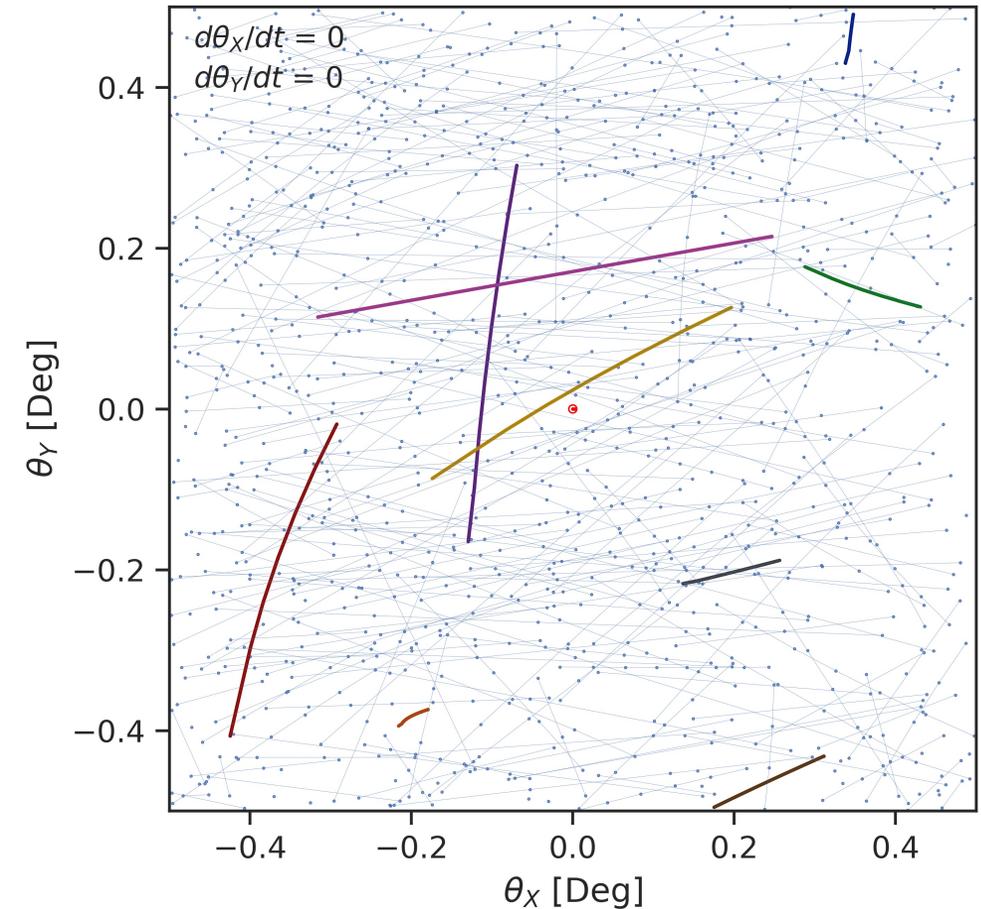
Why is “the linking problem” hard? (and how to solve it)

- The naïve algorithm is obviously too naïve. Most combinations seem clearly implausible.
- Why? They’re dynamically impossible, even at a coarse scale (admitting large observational errors). In fact, there are many, many, more implausible than plausible trajectories in the dataset of all combinations.
- So let’s turn things around: rather than testing each combination for whether it admits an orbit solution, let’s **test each plausible orbit** and check if it passes “near” actual observations.
- This looks formidable at first (covering a 6D phase space!) but turns out it can be done quite sparsely: a few “test orbits” will surface most plausible candidates for a link.



THOR: Tracklet-less Heliocentric Orbit Recovery

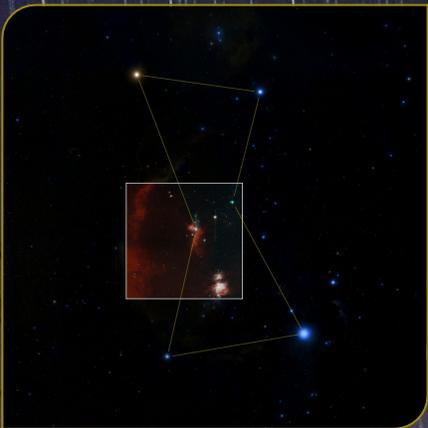
- Define a "test orbit", a trial orbit that may (or may not) contain a body.
 - We typically use state vectors (r, v) , rather than Keplerian elements
- Define a small "bundle" of orbits around the test orbit, $(\Delta r, \Delta v)$. Objects on these orbits will stay close to a particle on the test orbit over short timespans (month).
- Map the particles from the bundle onto the sky. A rough approximation is sufficient.
- Transform all observations in that region into the frame that follows the test orbit particle (assuming distance r).
- Result:
 - In the transformed frame, the test orbit particle will cluster at $(0, 0)$
 - Particles on orbits from the bundle will (approximately) trace lines
 - This is, effectively, an expansion of the trajectory relative to a nearby orbit
 - They can now be picked up by Hough transform or similar.
- Cover the observable areas of the phase space, find all asteroids.



Note: You can think of this as "shift-and-stack" in catalog space, nonlinear, over long timespans.

Zwicky Transient Facility

Systematic Exploration of the Dynamic Sky



- > 1000 images/night, 576 mpix
- > 300 M detected sources/night
- > 2 billion objects, 75-250 mea/obj/year
- > 1 M alerts/night



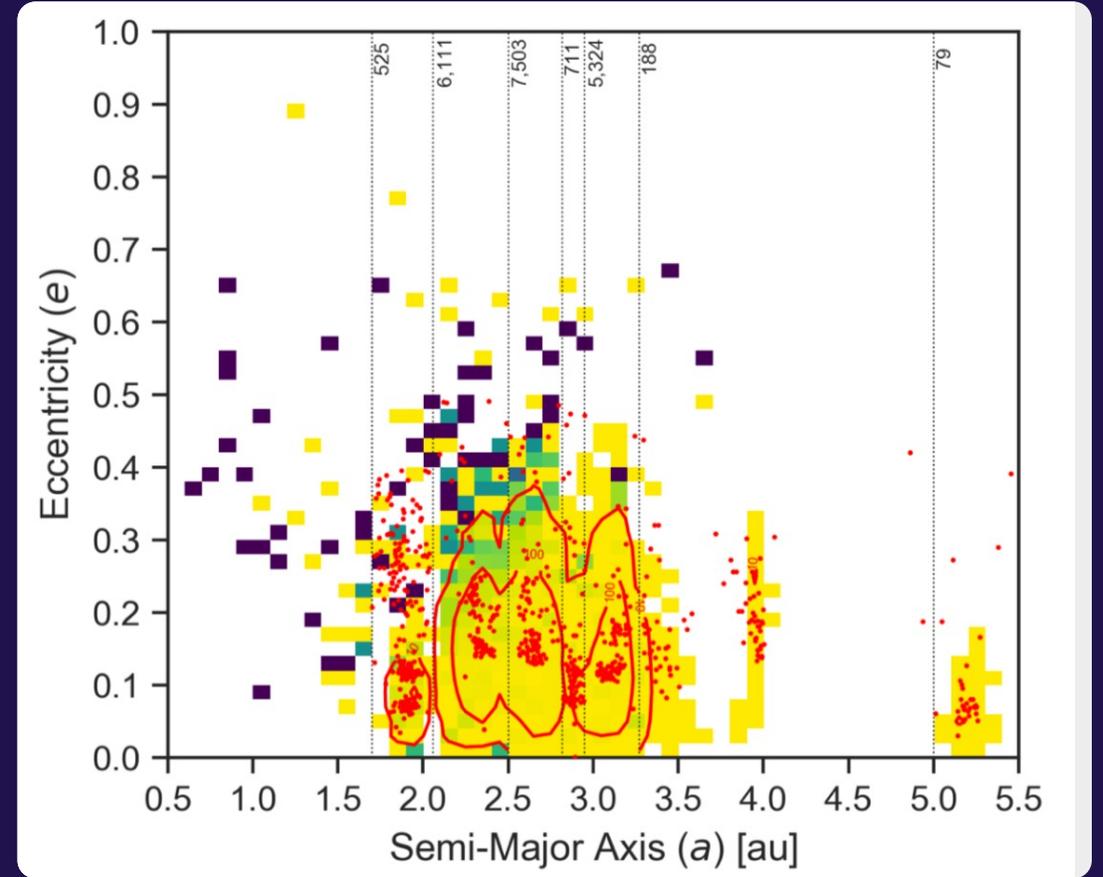
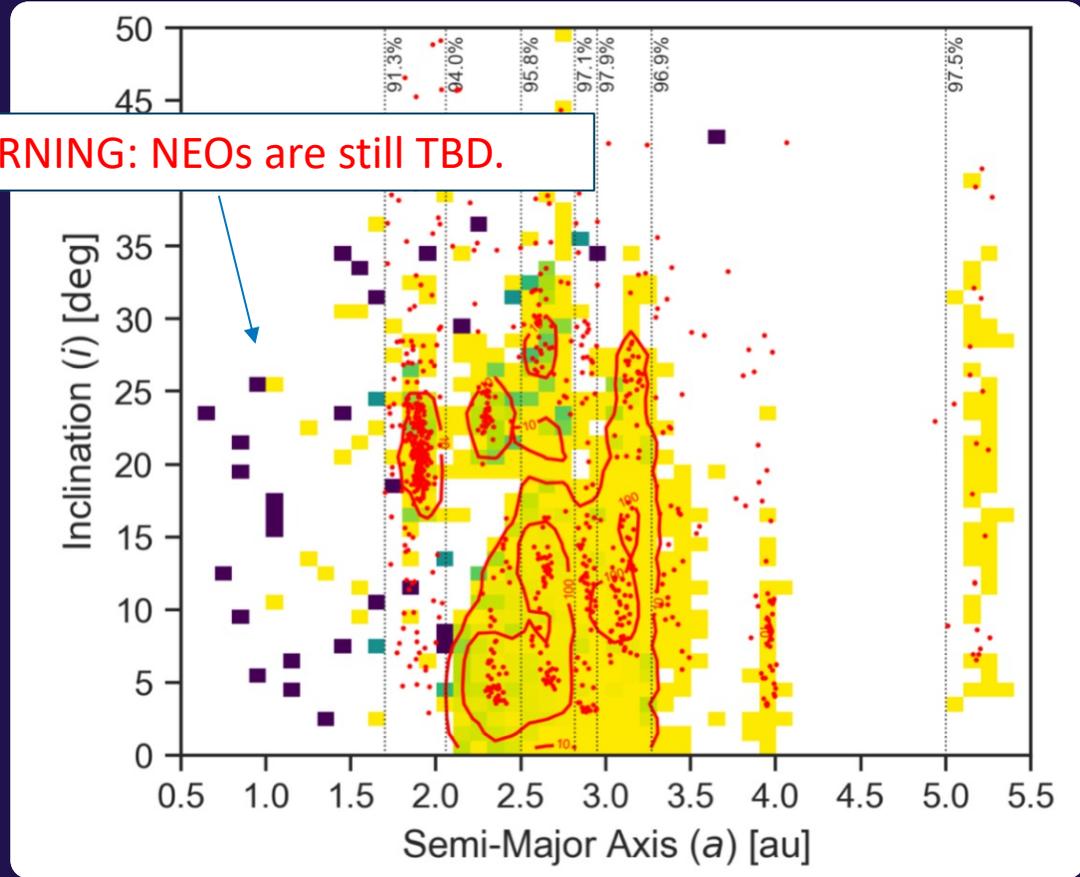
Public-Private Partnership.
40% obsv. on LSST-like cadence,
immediately public.

<http://ztf.caltech.edu>



Running THOR on ZTF Data

Results: recovered 95.6% of known objects. Found a dozen new ones. Also pre-covered a comet.

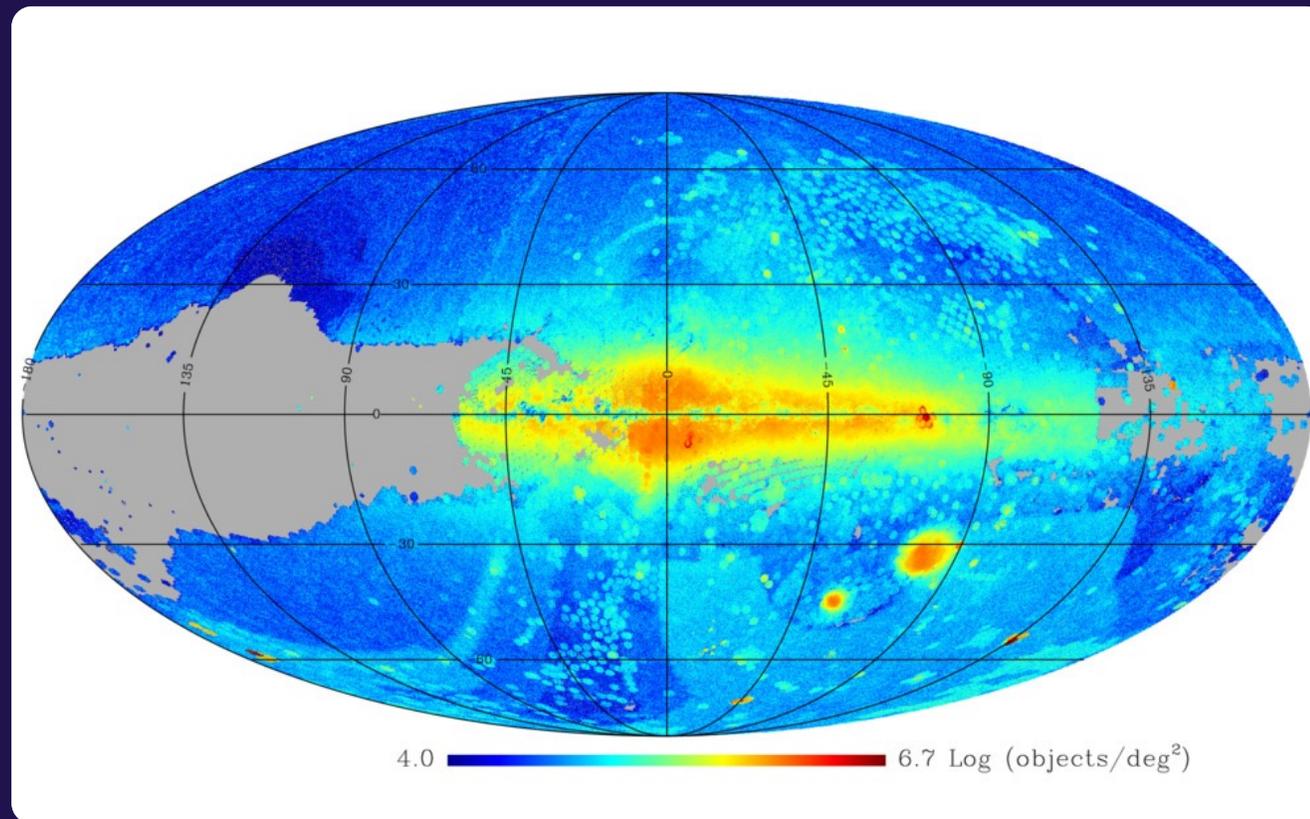


Just 44% findable by classical MOPS, 66.4% by ZMODE.

Moeyens et al. (2021)

More interesting: NOIRLab Source Catalog (NSC)

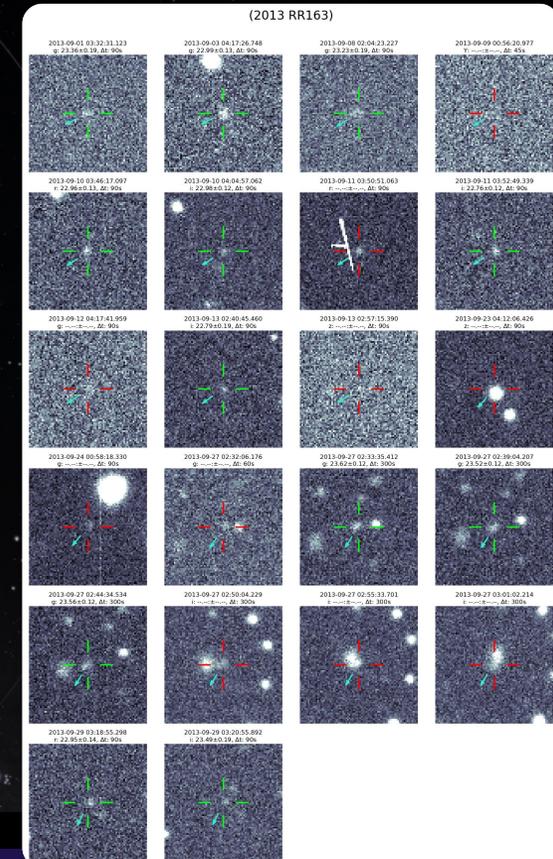
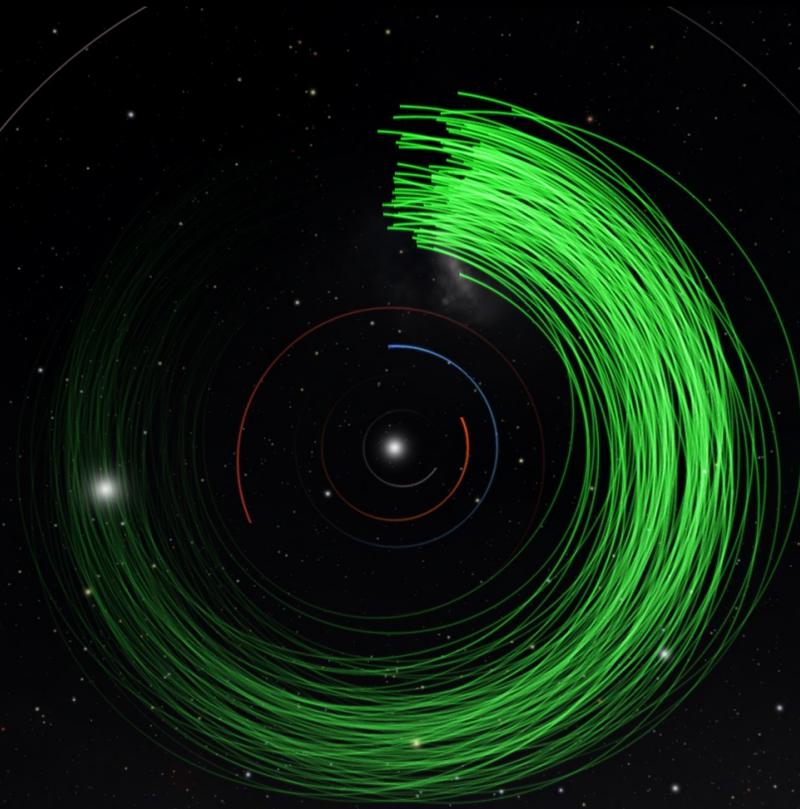
- The NOIRLab Source Catalog (NSC) is a catalog of nearly all of the public imaging data in [NOIRLab's Astro Data Archive](#)
- 68 billion individual source measurements
- Dominated by DECam + Blanco 4m measurements (3/4 of all exposures)
- Deep ($\sim 23^{\text{rd}}$ magnitude in most filters)
- ~ 1.7 billion don't appear to be static (i.e., could be asteroids).

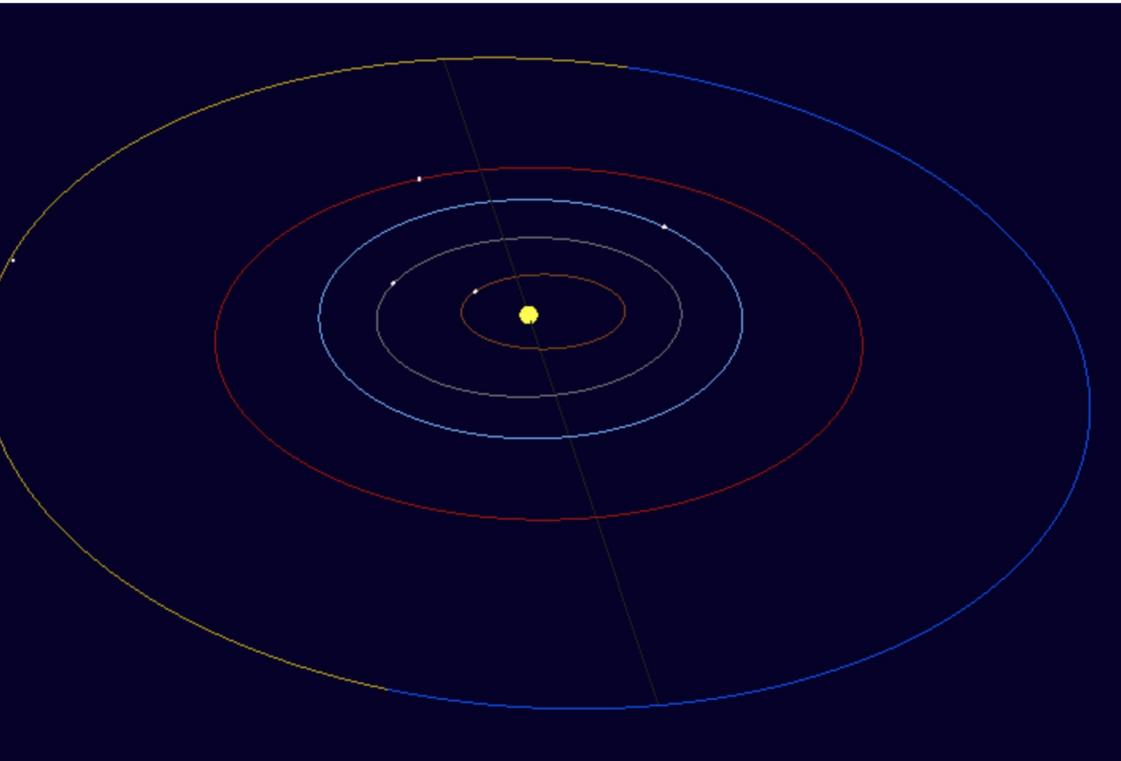


4.0 N5C DR2 Object Density Map

Initial run on 0.2% of data (15% of September 2013)

- Identified ~1200 known asteroids
- Linked 104 new objects (MPC-designated)
- Also pulled in a number of tracklets from the ITF.
- The vast majority have no 3-obsv tracklets (unidentifiable with “usual” software)
- A number of them have no tracklets (pairs of observations in a single night) at all.
- Extrapolating to the whole dataset, we’re looking at ~10-40k of new discoveries.
- The full dataset is being run right now.
- Est. cost: ~5M core-hours for the entire NSC dataset.





semimajor axis (AU)	2.5811065	<u>uncertainty</u>	1
mean anomaly (°)	25.55764	reference	E2022-L13
mean daily motion (°/day)	0.23768140	observations used	40
aphelion distance (AU)	2.805	oppositions	5
period (years)	4.15	arc length (days)	2418
P-vector [x]	0.84731874	first opposition used	2011
P-vector [y]	-0.49443833	last opposition used	2017
P-vector [z]	-0.19385996	residual rms (arc-secs)	0.26
Q-vector [x]	0.52963184	<u>perturbers coarse indicator</u>	M-v
Q-vector [y]	0.75971149	<u>perturbers precise indicator</u>	005E
Q-vector [z]	0.37726459	first observation date used	2011-01-30.0
absolute magnitude	18.61	last observation date used	2017-09-13.0
phase slope	0.15	computer name	Veres

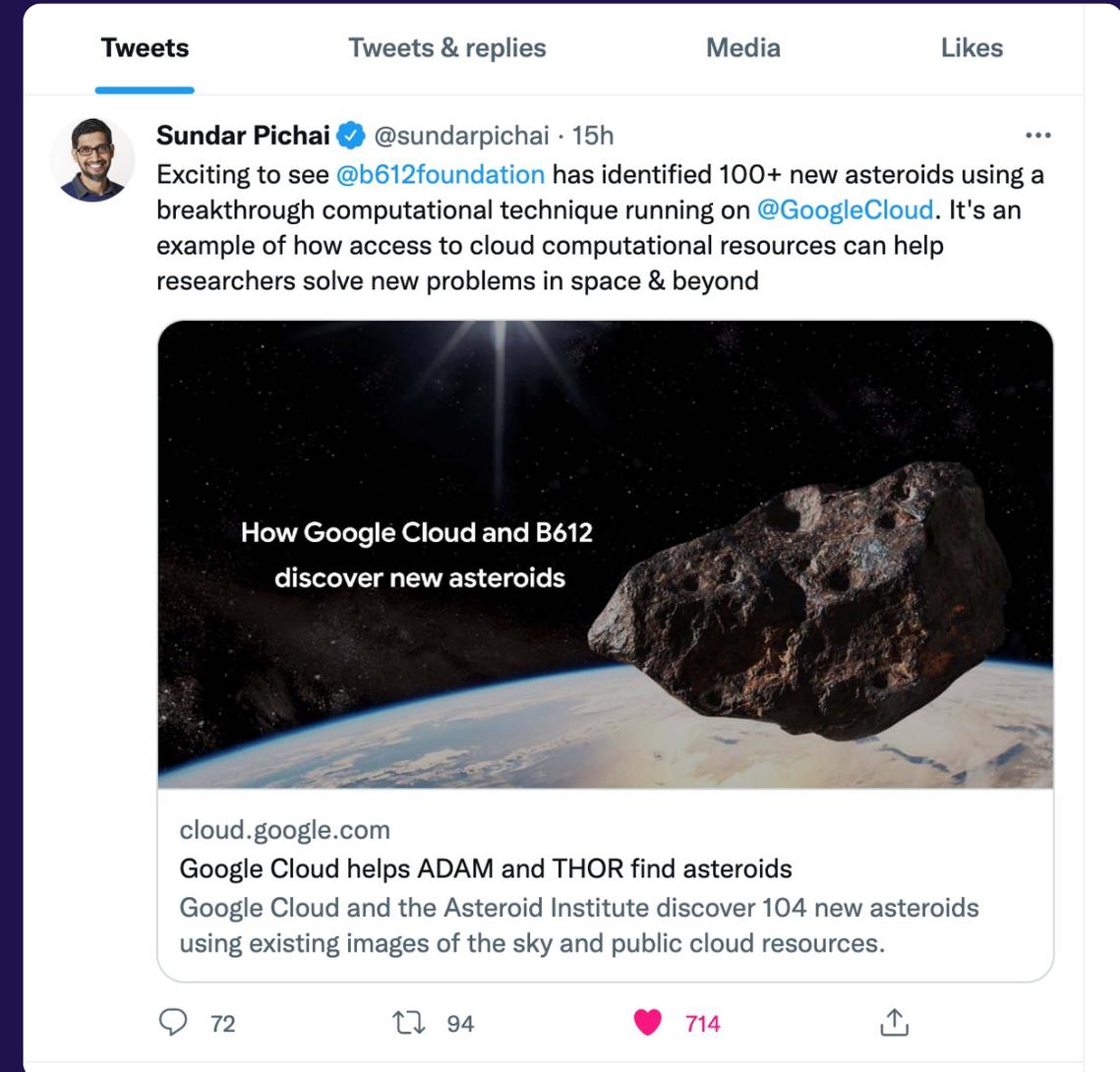
2013 RR165: An object discovered with ADAM::THOR, with previously unrecognized observations spanning over 6 years (!)

THOR: Implementation

- <https://github.com/moeyensj/thor>
- The core code is in Python + Numba JIT
- Optimizing performance, investigating GPU port.

- Parallelization on B612's ADAM ("Asteroid Discovery, Analysis and Mapping") astrodynamics platform on Google Cloud (<https://adam.b612.ai>).
- Why cloud? Available CPU power, scalability, the ease of building services.

- Hope to offer "asteroid discovery as a service": upload your detections, we'll find any asteroids that are in there.



The image shows a screenshot of a tweet from Sundar Pichai (@sundarpichai) posted 15 hours ago. The tweet text reads: "Exciting to see @b612foundation has identified 100+ new asteroids using a breakthrough computational technique running on @GoogleCloud. It's an example of how access to cloud computational resources can help researchers solve new problems in space & beyond". Below the text is a video thumbnail with the text "How Google Cloud and B612 discover new asteroids" overlaid on an image of an asteroid in space. Below the video is a link to "cloud.google.com" and a summary: "Google Cloud helps ADAM and THOR find asteroids" and "Google Cloud and the Asteroid Institute discover 104 new asteroids using existing images of the sky and public cloud resources." At the bottom of the tweet, there are icons for replies (72), retweets (94), likes (714), and a share icon.

Making any dataset an “asteroid search” dataset

- Many more datasets w/o nightly revisits now become usable for asteroid searches. And DECam archival data is a prime example of that!
- NOIRLab Source Catalog is fantastic for this purpose: good astrometry, well documented, single-epoch data available.
- Thank you for building this product (and please make a DR3 ☺)!
- Other DECam reduction efforts: please keep/pass along single-epoch data!



Things that are useful for efforts like these



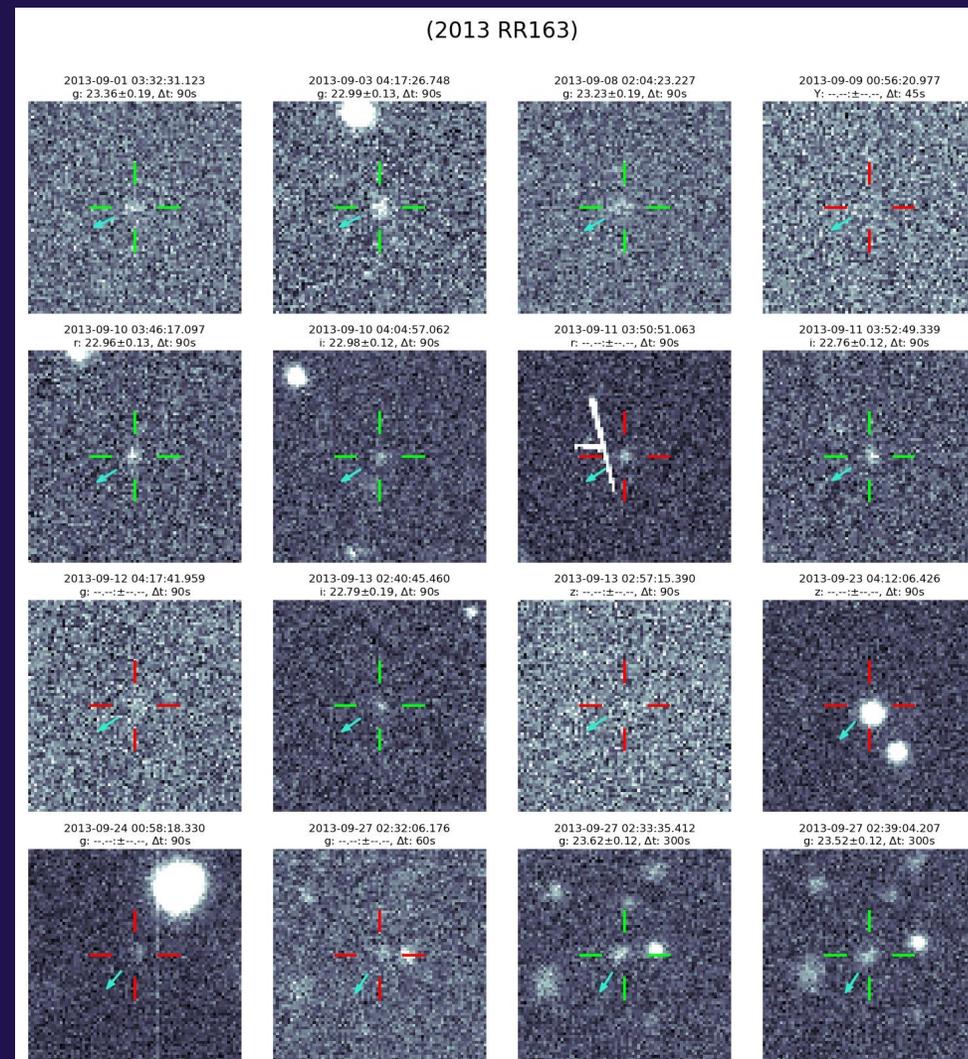
1. Bulk catalog access: we'd like to easily grab everything.

- Right now we're scraping the SQL database. Not pleasant.
- We love files™!

2. Fast image cutout service.

- ***Incredibly*** useful for visual confirmation, precise fits (note to Rubin: this will remain true even for LSST!).
- We'd like latencies of $O(1s)$ when building mosaics such as 🖱️

NSC asteroid cutout mosaic generator
<https://github.com/B612-Asteroid-Institute/cutouts>



What can we do with THOR(-like) algorithms?

- It allows searching of archival data but also, ...
- ... it allows for better yields and cadence re-optimization of future surveys.
- Example #1: with a tracklet-less algorithm, the LSST would be free to (in theory) double the area of the sky observed each night. Helps with non-Solar System science.
- Example #2: w. THOR + cadence changes, the LSST could half the number of remaining unknown PHAs (Jones et al. 2018).
- Example #3: a tracklet-less algorithm *may* enable a mission such as NASA NEO Surveyor to reach the 90% PHA discovery mandate.
- **A O(\$5M) investment in algorithms and computing could yield O(\$100M+) improvements on O(\$1Bn) missions.**



Thank You! Questions?

<https://dirac.astro.washington.edu>

A Universe Understood Through
Data-Intensive Discovery

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