

# Asteroids!

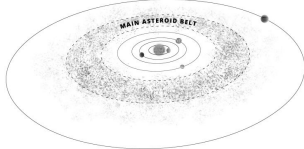
Jes Ford, Moore/Sloan & WRF Data Science Postdoc  
 Jake VanderPlas, eScience Institute (Mentor)  
 Mario Juric, Department of Astronomy (Mentor)



UNIVERSITY of WASHINGTON  
 eScience Institute

## Motivation

Be smarter than the dinosaurs! Instead of getting surprised by a killer asteroid, let's detect it with enough time to take action.



## Background

- Asteroid paths are specified by 6 Orbital Elements.
- Most asteroids live in the Asteroid Belt, so have similar orbital elements.
- Currently asteroids are discovered by taking sets of observations very closely spaced in time, so that individual asteroids can be matched up.
- The Large Synoptic Survey Telescope (LSST) will begin surveying the entire sky in 2021, and is tasked with detecting 90% of all potentially hazardous asteroids (among many other science goals).
- The main complicating factor is *noise* in images - for LSST we expect a ratio of false positives to real asteroids to be  $> 100$ .

## Goal

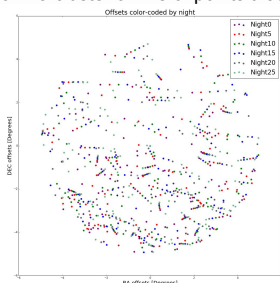
Design an algorithm for asteroid detection that does away with the requirement of repeat observations closely-spaced in time.

Then future telescopes like LSST will have freedom to spend more time pursuing diverse scientific questions about our Universe.

## Methods

### Phasing Observations

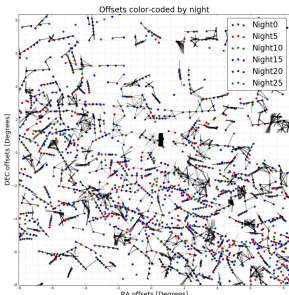
Instead of actual asteroid positions, observe positions *relative* to a reference asteroid. Then an asteroid with orbital parameters that are similar to the reference asteroid will form a cluster or line of points that map its relative position.



In the absence of noise, your eye easily picks out these phased nightly sequences of asteroids.

### Hough Transform

Algorithm for feature extraction from images. In the simplest implementation, the idea is that a line  $y = mx + b$  maps to a point  $(m, b)$ , and conversely each point  $(x, y)$  can be mapped to a line in slope-intercept space. A voting procedure results in the most popular lines being those that pass through many points.

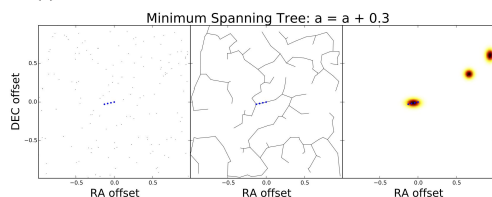


Hough Transform within gridded sections of the sky (no noise).

### Minimum Spanning Tree

Machine learning clustering algorithm, with a tunable cutoff distance between clusters:

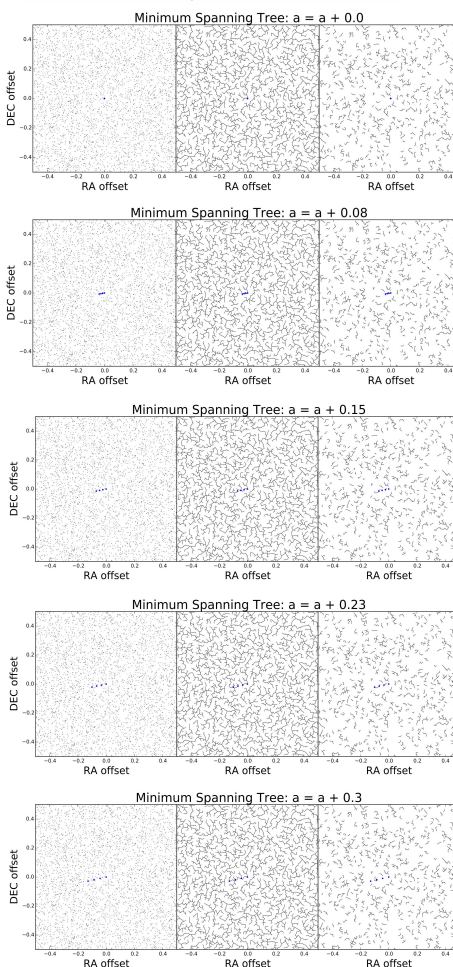
[https://github.com/jakevdp/mst\\_clustering](https://github.com/jakevdp/mst_clustering)



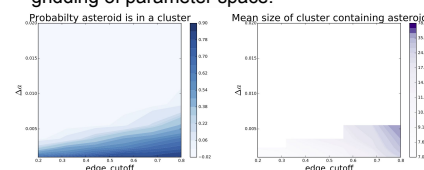
- Example with a single asteroid is shown here (in blue), observed on 4 nights, with a background of random noise.
- A minimal-length graph is constructed to connect all points.
- This graph is then trimmed to include only the edges shorter than some threshold distance (the "edge cutoff").
- The final panel reveals 3 clusters for this choice of cutoff.

## Results

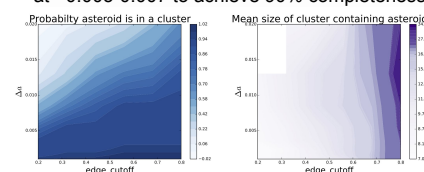
**Minimum Spanning Tree applied in a realistic noisy background:** Below we vary one of the orbital parameters, relative to the reference asteroid. In this example we "lose" the asteroid once its semi-major axis (distance from the sun) is offset from the reference asteroid by 0.15 astronomical units.



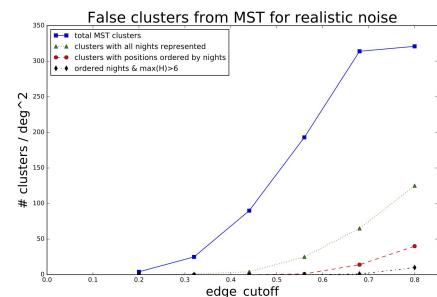
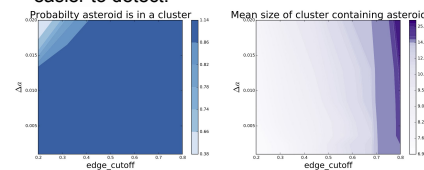
**Asteroids at Earth's radius ( $a=1$ ) are hard to detect** (they move fast), and will require a fine gridding of parameter space.



**Main Belt Asteroids ( $a=3$ ) will require a search through semi-major axis space gridded at  $\sim 0.005$ - $0.007$  to achieve 90% completeness.**



**Distant Asteroids ( $a=5$  and beyond) are much easier to detect.**



**Minimize False Detections:** a cluster representing a *real* asteroid will...

- contain a point from every night of observation,
- ordered sequentially by night, and
- lie close to a straight line (Hough Transform).

## Future Work

- Design an optimal grid through all 6 dimensions of orbital parameter space.
- Demonstrate asteroid detection at  $\geq 90\%$  completeness.
- Ensure algorithm can scale to LSST data rate of about 20 TB/night.

## Contact

email: [jesford@uw.edu](mailto:jesford@uw.edu)  
 website: <http://jesford.github.io>  
 desk: WRF Data Science Studio

<https://github.com/jesford>