Identifying Notable Objects from Spitzer Enhanced Imaging Astronomical Observations

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Goals

- Identify and locate notable objects (outliers) within the sky
- Define an area of interest around that object
- Narrow down search area for manual analysis

Approach

Excess infrared light could mean:

- Young Stellar Objects
- Active Galactic Nucleii
- Colliding Galaxies

Hypotheses: Young Stellar Objects

- Notable objects can be identified as extrema in terms of infrared light
- Some notable objects are grouped into interesting structures

Data

- NASA/IPAC Infrared Science Archive:
 - Wide-field Infrared Survey Explorer (WISE)
 - Identifies objects, and readings on the energies they emit
 - 800m objects (records), 815 GB
 - Contains:
 - Location (right ascension, declination)
 - Movement
 - Colour
 - Readings across a number of bands

Unexpected Colour Interesting Object

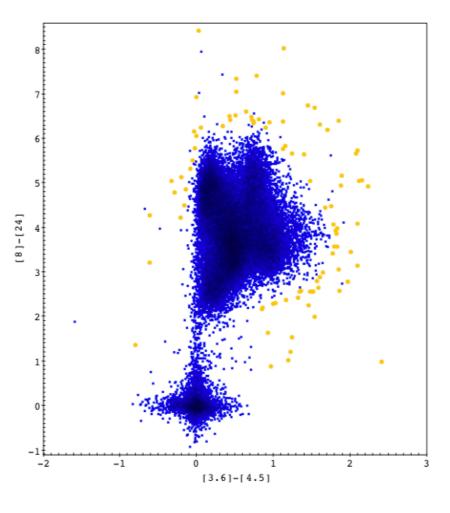


FIGURE 4: Color-color plot of SEIP sources, SNR≥10, not in the galactic plane, and not in major survey areas (blue), vetted color outliers (yellow).

Source: Gorjian et al.

Algorithm I

Algorithm I, Step 1: Preprocessing with K-Means Classification

- Canned K-Means (Dataproc Pending)
- Split points into two groups, ~ N within each

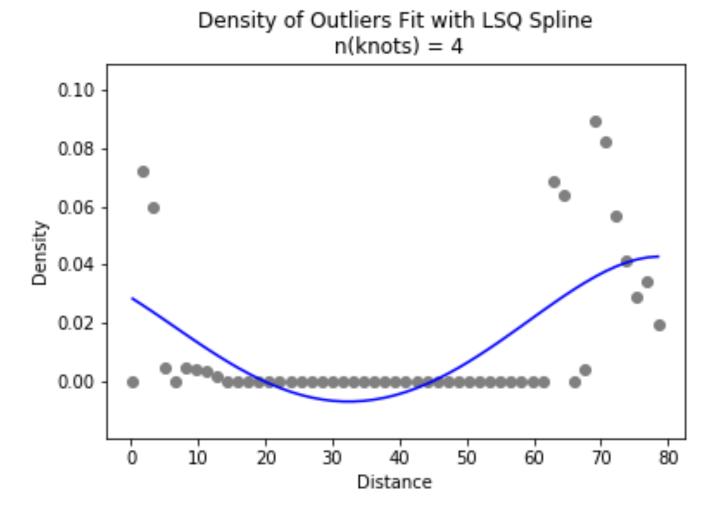
Algorithm I, Step 2: Outliers as Nodes, Distances as Edges

- MRJob
- Within each group, take outliers: x_i : $x_i \ge \mu \pm 2\sigma$
- Developed algorithm with parameters N, P, K.
- Complexity:
 - If number of cases < N(P): Does not run

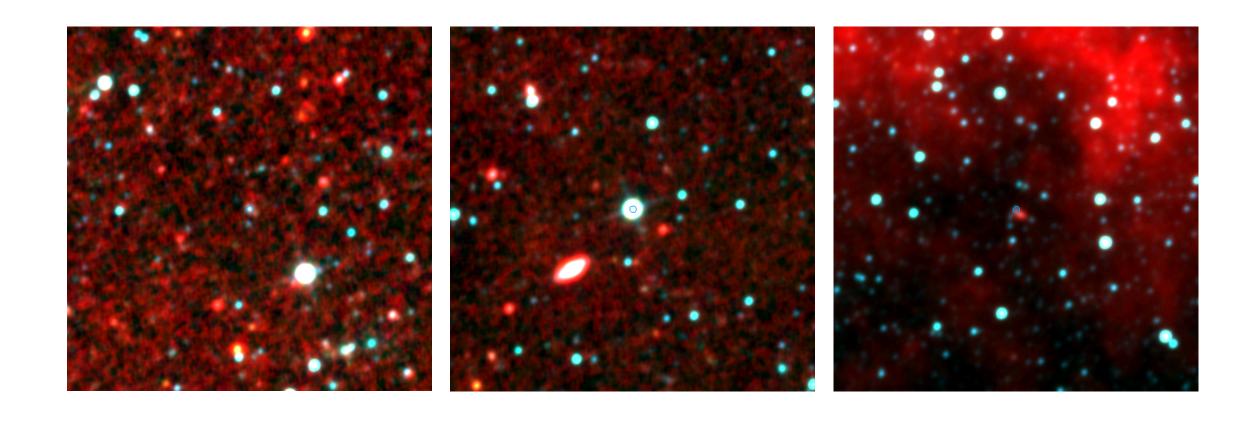
• Otherwise, approximately:
•
$$(\sum_{i=1}^{(N-N*P)} (N*P+i))*\frac{size-(N*P)}{(N*P)}$$

- Compare to:
 - $size^2$

Algorithm I: Distance Density with Fitted Spline



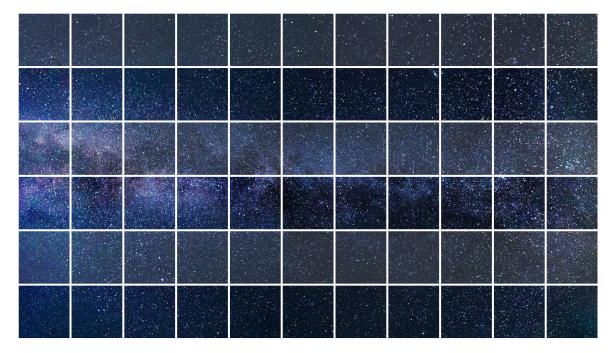
Algorithm I: Results



Algorithm II

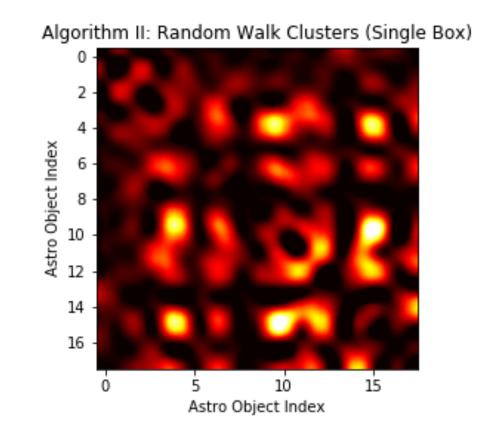
Algorithm II (MapReduce)

- Step 1: Split Sky (Dataproc Implementation in Progress)
 - Create grids of the entire sky
 - Create graphs of each grid



Algorithm II (MapReduce)

- Step 1: Split Sky (Dataproc Implementation in Progress)
 - Create grids of the entire sky
 - Create graphs of each grid
- Step 2: Random Walk
- (Dataproc in Progress)
- Step 3: Find Probabilistic Clusters (Coding In Progress)



Runtime Comparison

Algorithm I

- Local:
 - 85 min. 51 sec.
 - 365,601 rows
- Dataproc:
 - 8 min. 44 sec.
 - 25 workers with n1-standard-4 specs
 - 358,169 rows

Algorithm II (RandomWalk)

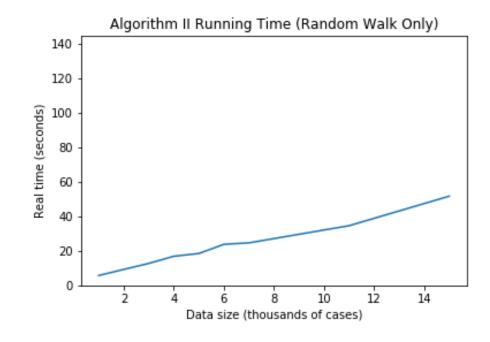
- Local:
 - 16 min. 9 sec.
 - 365,601 rows
- Dataproc:
 - In Progress

Runtime Comparison

Algorithm I



Algorithm II (RandomWalk)



Challenges

Problems:

- Graph algorithms without all data in memory (sky is too large)
- Complexity too great for a fully connected graph

Solutions:

- Running random sample (Algorithm I)
- Grids and streaming processing (Algorithm II)

Algorithm 1 Implementation

- 1. Initialize a node list (Map 1)
- 2. for each object: (Map 1)
 - add the object to our node list
 - 2. if the node list is too large: remove nodes at random until the list is size N(P):

- 3. If the node list is sufficently full (size N(P)):
 - 1. yield the object's id and distance to each node

- 4. For each object (Reduce 1)
 - 1. take K edges (the closest K)

Algorithm 1 Implementation (Cont'd)

5. For each object (Map 2)

- 1. Create a density function of distances from the object
- 2. Fit a spline function along this density curve
- 3. Find the first saddle point in this spline, usually a local minimum
- 4. If the first saddle point is less than 1 return the object and its saddle point

Algorithm 2 Implementation

1. Grids:

- 1. r x c grids -> bite-sized chunks
- 2. Output key: grid id

Algorithm 2 Implementation (Cont'd)

3. Random Walk

- 1. Fully connected graphs w/ distance\$^{-1}\$ as edge weight
- 2. Random walks on graph -> gradient of re-visit frequency

Algorithm 2 Implementation (Cont'd)

4. Watershed Algorithm:

4. Segmentation of the "image" - i.e, gradient

Code:

https://github.com/ibhojwani/seip_big_data