Construction of a catalog of galaxy clusters in collision process.

Martín de los Rios, Mariano Dominguez & Dante Paz

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Mock Catalog.

Millenium

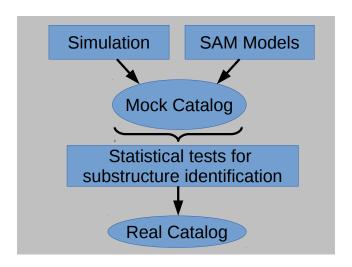
Statistical tests for identification of substructures.

The Dressler-Shectman test Iterative Dressler-Shectman test. Gaussian mixture..

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Real catalogs of galaxy clusters.

Future work.



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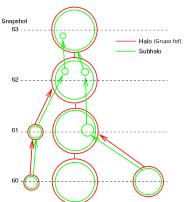
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- $\sigma_8 = 0.9 \, 0.9$
- ► Snapshots = 64

- Millenium simulation: Springel et al. 2005.
- ► WMAP results: *Spergel et al. 2003*.

Study of the merger trees.

▶ Based on the subhalos merger trees, we construct the merger tree for every fof group in the simulation.



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- ▶ Pinkney et al. 1996
 - They say that this is the best test known so far for substructure identification.
 - The test efficiency is lower when the substructures are aligned along the line of sight, and also when the cluster under consideration have less than 30 galaxy members.
 - They recommend to look for significant values of the skewness moment on radial velocity distribution.
 - This test can be used to identify fof-group substructures accreted in a major merger event up to 3

Gyr later. This time interval is equivalent to 10 snapshots in the simulation.



The Dressler-Shectman test

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- We define the p-value as

$$p = \frac{N(\Delta_{MC} > \Delta)}{N_{MC}} \tag{1}$$

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Algorithm steps

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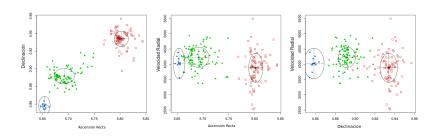
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- ► The algorithm converges if the number of galaxies between 2 steps is the same.

Gaussian mixture..

Gaussian mixture. (Mclust)

The application of the DS test give us a group of galaxies with high probability to lie in a substructure. Then, in order to define substructures we identify clumps of galaxies based on their proximity.



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Application of the DS Test in the mock catalog.

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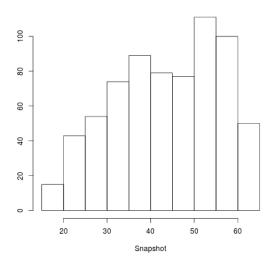
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- ▶ If we consider only the clusters whose radial velocity skewness is significantly different from zero, the sample is reduce to 715 clusters.

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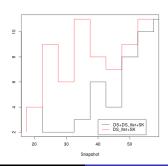
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- Over those 119, only 46 clusters have been detected for the traditional DS test complemented with the skewness test.



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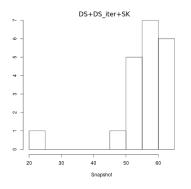
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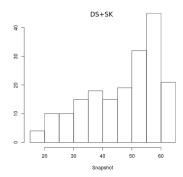
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- ▶ Over the sample of 46 clusters, 20 systems have a relative occupation higher than 0.5.





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After the substructure identification, each one was related with the subhalo in the mock that have more galaxies in the group identified by mclust.

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- To improve the identification of the substructure, we calculate the coordinates of the center taking in account the luminosity.
- After that, we estimate the velocity dispersion and the virial radius.

$$\begin{array}{lcl} R_{vir} & = & \frac{\pi}{2} \frac{ngal(ngal-1)}{\sum_{i>j}^{ngal} R_{ij}^{q-1}} \\ \\ \sigma & = & \frac{\sqrt{\pi}}{ngal(ngal-1)} \sum_{i=1}^{ngal-1} \omega_i g_i \\ \\ \omega_i & = & i(ngal-i) \\ g_i & = & v_{i+1} - v_i \end{array}$$

(2)

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- ▶ We compare the center of the identified groups with the centers of the associated subhalos, finding a good estimation.
- We compare the virial radius of the identified groups with the virial radius of the subhalos, finding that we are overestimating the real values.
- We compare the velocity dispersion of the identified groups with the velocity dispersion of the associated subhalos, finding that our values are in good concordance with the real values.

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Of the 715 clusters, we find 28 with 2 substructures with a relative occupation higher than 0.5 while each group have more than 3 galaxy members. From this 28, in 19 we find the type 0 subhalo and a type 1 subhalo (case 1), in 4 we find 2 type 1 subhalos (case 2) and in 5 we identify a false substructure (case 3).

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Application of the method of detection to real catalogs.

Catalog	Berlind et al.	Tempel et al.	Eke et al.	Mock
Galaxy survey	SDSS DR3	SDSS DR8	2dF	-
N° galaxy clusters	8148	77858	28877	-
N° clusters ($N_{gal} > 30$)	77	389	144	2854
pval > 0,15	30	246	112	1448
pval > 0.15 + SK	30	132	60	715
Ds iter	20	86	41	46
Mclust	15	80	38	44

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- Study the physics properties of the galaxies that belong to the identify substructures.
- Perform astrophysical test over our sample of colliding cluster candidates of the catalog looking for impose some constraints on dark matter particle properties.

