Project 2

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1 Introduction

In this study, I examined the morphological type of galaxies in clusters as a function of the distance from the galaxy to the cluster center (d_{center}). Cluster quantities with overdensity 200 were used for this work. I separated the GalWCat19 catalog into several mass bins and examined this trend in each of these subpopulations of massive clusters. The cumulative fraction of elliptical galaxies within an expanding bubble of radius d_{center} is displayed in the figure at the end of the report.

2 Methodology

To begin, I filter the cluster catalog to only include cluster masses greater than $1 \times 10^{28} \rm \ M_{\odot}$ and comoving distances less than 225 Mpc h⁻¹, according to the catalog completeness thresholds [?]. I then sort the catalog of clusters into five mass bins. Subsections of the catalog are taken within these mass thresholds and processed according to the procedure below.

Within each mass limit, the galaxy catalog is cross matched with every cluster ID present in this mass bin. The distance from each galaxy to its host cluster's center is calculated by

$$d_{center} = \sqrt{d_{los}^2 + d_{sky}^2} \tag{1}$$

where r_{gal} is the line of sight comoving distance from the observer to the galaxy, and

$$d_{los} = |r_{gal} - r_{cluster}| \tag{2}$$

$$d_{sky} = r_{gal} \times \cos^{-1}(\cos \Delta_{RA} \cos \Delta_{DEC})$$
(3)

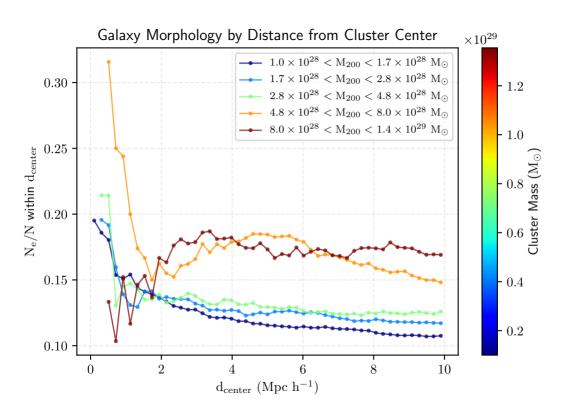
with Δ_{RA} and Δ_{DEC} equal to the differences in RA and DEC between the cluster center and the galaxy, in radians. This presumes that the angle between the galaxy and the cluster is sufficiently small that the distance between them can be calculated to reasonable precision using the Pythagorean Theorem.

The cluster center distances for all galaxies belonging to clusters in this mass bin are then tabulated along with a flag to indicate whether the galaxy is elliptical or not. Then, these galaxies are further binned by the distance to their cluster's center. The data is then plotted cumulatively to show how the fraction of elliptical galaxies changes as the threshold distance is increased outward from the cluster's center. A maximum distance is set at 10 Mpc h^{-1} , according to the GalWeight maximum projected radius [?].

3 Results

Examining the figure, it is found that in general low mass clusters have a denser relative concentration of elliptical galaxies at their centers. Further out in the low mass cluster, the fraction of elliptical galaxies decreases.

An entirely different trend emerges from the high mass clusters. Here, it appears that the very highest mass clusters have very few elliptical galaxies at their cores. This fraction steadily increases out to large center distances, meaning that the relative fraction of elliptical galaxies in high mass clusters peaks at around 4 - 6 Mpc h^{-1} , and drops slightly on either side.



```
import pdb
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.pylab as pl
import matplotlib
matplotlib.rcParams['text.usetex'] = True
def scinot(num):
    Formats numbers in scientific notation for LaTeX.
    numstrs = '{:.1E}'.format(num).split('E+')
    return r'${} \times 10^{{{{}}}}$'.format(numstrs[0], int(numstrs[1]))
# Set up figure
fig, ax = plt.subplots(1, 1)
# Collect data from external files
cluster_data = np.loadtxt('GalWCls19.txt', skiprows=41)
galaxy_data = np.loadtxt('GalWGal.txt', skiprows=9)
# Bin the clusters by mass
n bins = 5
cluster masses = cluster data[:, 18]
cluster_distances = cluster_data[:, 5]
# Only take clusters larger than 10^14 for completeness
cluster_data = cluster_data[np.logical_and(
    cluster_masses >= 1e14, cluster_distances <= 225)]</pre>
cluster_masses = cluster_masses[np.logical_and(
    cluster_masses >= 1e14, cluster_distances <= 225)]</pre>
# Make mass bins, if necessary
mass bins = np.logspace(
    np.log10(
        min(cluster_masses)), np.log10(
            max(cluster_masses)), n_bins + 1)
inds = np.digitize(cluster_masses, mass_bins)
colors = pl.cm.jet(np.linspace(0, 1, n_bins))
# For each mass bin in the sample...
for i in range(1, n_bins + 1):
    # Narrow the sample to clusters in this mass range
    clusters = cluster_data[inds == i]
    center_distances = np.array([])
    elliptical_flags = np.array([])
    if len(clusters) == 0:
        continue
    # Loop over all the clusters in this group
    for j in range(len(clusters)):
        # Collect galaxies in this cluster by matching cluster ID
        galaxies = galaxy_data[galaxy_data[:, 10] == clusters[j, 0]]
        # Calculate the distance of each galaxy to the cluster center
        r gal = galaxies[:, 3]
        r_c = clusters[j, 5]
```

```
d los = np.abs(r gal - r c) # Line-of-sight distance
        # Get sky-projected distance
        dRA = (galaxies[:, 0] - clusters[j, 1]) * np.pi / 180.
        dDEC = (galaxies[:, 1] - clusters[j, 2]) * np.pi / 180.
        c = np.arccos(np.cos(dRA) * np.cos(dDEC)) # Distance on sky in radians
        d_sky = r_gal * c # Distance on sky in same units as r_gal
        # Distance from galaxies in this sample to this cluster's center
        d = np.sqrt(d_los**2 + d_sky**2)
        is_elliptical = galaxies[:, 9] == 2
        # Add data from this cluster to the mass group's data
        center_distances = np.concatenate((center_distances, d))
        elliptical_flags = np.concatenate((elliptical_flags, is_elliptical))
    # Bin by distance to create x and y data
    n_dist_bins = 50
    try:
        #dist_bins = np.logspace(np.log10(min(center_distances)),
np.log10(max(center_distances)), n_dist_bins)
        dist_bins = np.linspace(0, 10, n_dist_bins)
    except BaseException:
        pdb.set trace()
    dist inds = np.digitize(center distances, dist bins)
    # Find fraction of ellipticals in each distance bin
    all bincenters = (dist bins[:-1] + dist bins[1:]) / 2
    frac ellip = []
    bincenters = []
    for k in range(1, len(dist_bins)):
            rule = dist_inds == k
        rule = dist inds <= k
        if len(elliptical flags[rule]) > 10:
            frac_ellip.append(
                sum(elliptical_flags[rule]) / len(elliptical_flags[rule]))
            bincenters.append(all_bincenters[k - 1])
    # Plot results
    ax.plot(bincenters,
            frac_ellip,
            alpha=0.75,
            c=colors[i - 1],
            lw=1,
            marker='o',
            markersize=2,
            label=r'{} $< \rm M_{{200}} < $ {} $\rm M_{{\odot}}
\frac{1}{2}.format(scinot(mass_bins[i - \frac{1}{1}] * \frac{1}{2}),
scinot(mass_bins[i] * 1e14)))
if n bins > 1:
    sm = plt.cm.ScalarMappable(
        cmap=pl.cm.jet,
        norm=plt.Normalize(
            vmin=min(
                cluster masses *
                le14),
            vmax=max(
```