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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>
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    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]]
        # Get distance information
        r gal = galaxies[:, 3]
        r_c = clusters[j, 5]
        # Get angle between galaxy and cluster center
        RA_gal = galaxies[:, 0] * np.pi / 180.
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RA_{cluster} = clusters[j, 1] * np.pi / 180.
         DEC gal = galaxies[:, 1] * np.pi / 180.
         DEC cluster = clusters[j, 2] * np.pi / 180.
         theta = np.arccos(np.sin(DEC gal)*np.sin(DEC cluster) +
np.cos(DEC_gal)*np.cos(DEC_cluster)*np.cos(RA_gal - RA_cluster))
        # Distance from galaxies in this sample to this cluster's center
         d = np.sqrt(r_gal^{**2} + r_c^{**2} - 2.*r_gal^*r_c^*np.cos(theta))
         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
    dist_bins = np.linspace(0, 10, n_dist_bins)
    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 # Cumulative</pre>
         if len(elliptical flags[rule]) > 10:
             frac = sum(elliptical_flags[rule]) / len(elliptical_flags[rule])
             frac ellip.append(frac)
             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'{} < \mbox{rm} \mbox{M}_{{200}} < \mbox{$ {} \mbox{mm} \mbox{M}_{{000t}}
$'.format(scinot(mass_bins[i - 1]), scinot(mass_bins[i])))
if n_bins > 1:
    sm = plt.cm.ScalarMappable(cmap=pl.cm.jet,
norm=plt.Normalize( vmin=min(cluster_masses), vmax=max(cluster_masses)))
    cbar = fig.colorbar(sm)
    cbar.ax.set_ylabel(r'Cluster Mass ($\rm M_{\odot}$)', rotation=90)
ax.set_xlabel(r'$\rm d_{center}\ (\rm Mpc\ h^{-1})$')
ax.set_ylabel(r'$\rm N_e / N$ within $\rm d_{center}$')
ax.set_title('Galaxy Morphology by Distance from Cluster Center')
ax.grid(ls='--', alpha=0.25)
ax.legend(fontsize=8)
plt.show()
```