Mass Sampling Documentation

Functions

Single Redshifts

Main function:

```
A. def mass_sampling(mass_range, redshift = 0.0, mdef = '200c', model = 'bocquet16', sample num = 100000):
```

the function to give back a sample of single-redshift cluster mass distribution based on the halo mass function

Library:

• Colossus

step:

- Initiate a NumPy array as cluster masses
- Use the halo mass function given by colossus, give back an array of number density
- Extract power from the mass array using helper 3, from 10^14 M☉ to 14
- Use helper 2 to calculate the final mass sampling

Helpers:

1. lnpo(mass, min, max, test fun):

likelihood function used by MCMC

parameters:

mass: a float or a 1d NumPy array of cluster mass in 10ⁿ M⊙ unit

2. interpolate_MCMC(mass_array_p, mfunc_n, mass_range, sample_num,
redshift):

interpolate and normalize mfunc_n, use the result as a likelihood function and perform MCMC method to get the sample.

parameters:

input:

mass_arr_p: 1d NumPy array of cluster mass power (for example, 10^14 M⊙ represented as 14 in arr)

mfunc_n: 1d NumPy array of halo number density * 10^5
mass_range: a tuple of cluster masses, lower limit, and upper limit for
sampling

sample num: an integer of the number of samples

output:

sample_chain.flatten(): an 1d NumPy array of mass sampling, same unit as
mass arr p

Library:

- scipy(interpolate & integrate)
- Emcee

step:

- Interpolate mass arr p & mfunc n, return a function object
- Calculate the integration of function between the mass range, normalize the function using integration, result in test_func
- Use emcee package to do MCMC simulation, Helpers 1 (lnpo) used as the likelihood function, return test_func and mass_chain.flatten()

3. extract power(mass arr):

```
Function to extract the power of galaxy cluster mass array, switch from 10^n to n
```

Multiple Redshifts

Main function:

for more info

```
B. def mul_redshift_mass_sampling(rs_dist = "skewnorm", rs_range = (0.0, 1.5), mass_range = (14.0, 16.0), mdef = '200c', model = 'bocquet16', sample_num = 100000, store = True):
```

the function to give back a sample of multi-redshift cluster mass distribution based on halo mass function

```
Parameters:
_______
input:
rs_dist: a string, representing the distribution of cluster redshift,
"skewnorm" by default
rs_range: a tuple of redshift range, (0.0, 1.5) by default
mass_range: a tuple of cluster masses, lower limit and upper limit for sampling, [min, max]
in 10^min M② unit
mdef: The mass definition in which the halo mass M is given; see
colossus doc for more info
(https://bdiemer.bitbucket.io/colossus/lss_mass_function.html#lss.mass_function.massFunction)
model: the halo mass function model used by colossus; see colossus doc
```

sample num: an integer of the number of samples, 100000 by default

store: a boolean, if True store mass array and redshift into a csv file
and return a string of path to file if False returns None

output:

fin_cluster: a Pandas dataframe with 2 columns of ["mass_arr",
"redshift"], NumPy array of cluster mass corresponding to redshift stored
in each row

filepath: str of file path if store=True, else None

Library:

• pandas

Step:

- Call helper 2 to obtain the redshift interval (chop) and corresponding cluster number.
- Loop through every redshift in chop, call function A(mass_sampling) to do single redshift mass sampling, store mass and redshifts in two pandas Series

Helpers:

1. skew_sample(size = 10000):

the function to give back a sample of redshift based on skew gaussian distribution imitating SPT cluster data:

https://pole.uchicago.edu/public/data/sptsz-clusters/

Parameters:

input:

size: integer, sample number

output:

mass_chain: a NumPy array of length = sample_num

rs_sample: a 1d NumPy array of clusters' redshift sample with length =
size

Library:

• scipy.stats

Step:

• Use scipy.stats to imitate redshift distribution

- Use skewnorm.rvs to draw a sample from it
- 2. single_redshift_num(rs_range, sample_num, rs_dist_model):
 """

the function to give back redshifts and sample_num per redshift for multi-redshift sampling

Parameters:

input:

rs_dist_model: a string, representing the distribution of cluster
redshift

rs_range: a tuple of redshift range, (0.0, 1.5) by default
sample num: an integer of the number of samples, 100000 by default

output:

chop: a NumPy array of redshifts

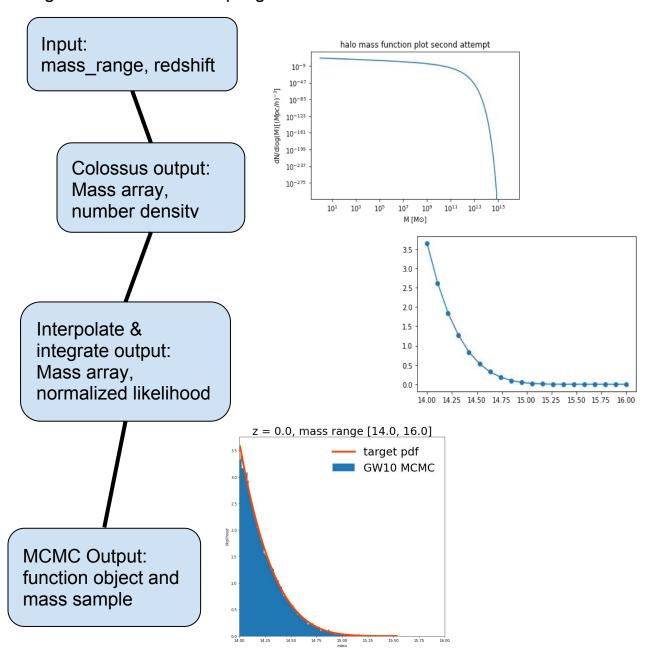
num_per_redshift: a NumPy array of cluster num within the corresponding
redshift interval of the same index number in chop
"""

Step:

- Call helper 1 to obtain redshift distribution array
- Based on redshift range, divide it to several chops of redshift, each chop used as limit
- Loop through the redshifts array given by helper 1, approximate redshifts between two chops to the lower limit of it.
- Use another NumPy array of the same size to keep track of cluster number in each interval

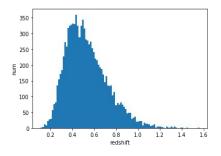
Procedure

Function A [mass_sampling] Single redshift mass sampling



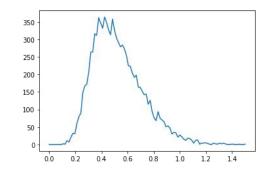
Input: mass_range, redshift range, sample size

Function B [mul_redshift_mass_sampling] Multiple redshift



[Skew_sample] output: redshift array based on skewed gaussian distribution imitating

[single_redshift_num] output: a chop array of redshift interval, a num_per_redshift array of cluster number/redshift interval



Called [mass_sampling] to sample for each redshift in chop

Output: a pandas dataframe with mass and redshift as two columns

