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**atelier**

***Release 0.1a0***

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**Version:** 0.1a0

The “atelier” is my personal compilation of custom written software designed for my astrophysical research. Please feel welcome to use the python package and its modules. However, I would like to emphasize that the software is provided ‘as is’ and no warranty is given for its functionality (BSD 3-Clause license).

If you would like to contribute to the software please contact Jan-Torge Schindler via [github](#).



## ATELIER.LUMFUN - INTRODUCTION

### 1.1 Introduction to the Atelier's luminosity function class with a strong emphasis on quasar luminosity functions

```
[1]: # General imports
import numpy as np
import matplotlib.pyplot as plt
from astropy.cosmology import FlatLambdaCDM

# Importing the luminosity function module
from atelier import lumfun
```

#### 1.1.1 1) Introduction to the luminosity function object

### 1.2 1.1) Instantiating the luminosity function object

ADS reference: <https://ui.adsabs.harvard.edu/abs/2013ApJ...768..105M/abstract> The quasar luminosity function in the paper is defined using a broken double power law with the luminosity in magnitudes. An implementation of a magnitude based broken double power law luminosity function already exists in the *lumfun* module.

```
[2]: print(lumfun.DoublePowerLawLF.__doc__)

Luminosity function, which takes the functional form of a double
power law with the luminosity in absolute magnitudes.

The luminosity function has four main parameters:

- "phi_star": the overall normalization
- "lum_star": the break luminosity/magnitude where the power law slopes
  change.
- "alpha": the first power law slope
- "beta": the second power law slope
```

The four main parameters need to be specified either as parameters or as functions. In this example only the source density normalization is a function, which depends on redshift (redsh), a  $z=6$  normalized value of the source density  $\log\_phi\_star$  and a parameter  $k$ .

```
[3]: # Define the source density normalization as in McGreer+2013
def phi_star(redsh, log_phi_star, k):
    return 10 ** (log_phi_star + k * (redsh - 6))

log_phi_star = lumfun.Parameter(-8.94, 'log_phi_star', one_sigma_unc = [0.24, 0.20])
k = lumfun.Parameter(-0.47, 'k', vary=False)
```

We begin with defining the function for *phi\_star* and its parameters. Redshift (*redsh*) and luminosity (*lum*) are always arguments of the luminosity function and do not require further definition. However, we need to define and provide *log\_phi\_star* and *k*.

For these parameters we use the **lumfun.Parameter** class which has the following attributes:

```
[4]: print(lumfun.Parameter.__doc__)

A class providing a data container for a parameter used in the
luminosity function class.

Attributes
-----
value : float
    Value of the parameter
name : string
    Name of the parameter
bounds : tupler
    Bounds of the parameter, used in fitting
vary : bool
    Boolean to indicate whether this parameter should be varied, used in
    fitting
one_sigma_unc: list (2 elements)
    1 sigma uncertainty of the parameter.
```

In the next step we define the other three main parameters and instantiate the broken double power law luminosity function (*atelier.lumfun.DoublePowerLawLF*).

```
[5]: lum_star = lumfun.Parameter(-27.21, 'lum_star', one_sigma_unc = [0.33, 0.27])
alpha = lumfun.Parameter(-2.03, 'alpha', vary=True, one_sigma_unc = [0.14, 0.15])

# The beta slope does not have an uncertainty as it set to -4 in McGreer+2013
beta = lumfun.Parameter(-4, 'beta', bounds=[-8, -3])

# Wrapping the parameters in a dictionary
parameters = {'log_phi_star': log_phi_star, 'k': k, 'lum_star': lum_star,
              'beta': beta, 'alpha': alpha}
# Wrapping the parameter functions (param_functions) in a dictionary
param_functions = {'phi_star': phi_star}

# The luminosity function type attribute is used in specific calculations based on the
# luminosity function
# only lum_type = 'M1450' has a special meaning at the moment.
lum_type = 'M1450'
```

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```
mcgreer2013 = lumfun.DoublePowerLawLF(parameters, param_functions, lum_type=lum_type)

[INFO]-----
[INFO] Performing initialization checks
[INFO]-----
[INFO]-----
[INFO] Main parameter phi_star is described by a function.
[INFO] The function parameters are: ['redsh', 'log_phi_star', 'k']
[INFO] All parameters are supplied.
[INFO] Parameters "lum" and "redsh" were ignored as they are luminosity function_
↳arguments.
[INFO]-----
[INFO] Main parameter lum_star is supplied as a normal parameter.
[INFO]-----
[INFO] Main parameter alpha is supplied as a normal parameter.
[INFO]-----
[INFO] Main parameter beta is supplied as a normal parameter.
[INFO]-----
[INFO] Initialization check passed.
[INFO]-----
```

**Important** With the instantiation of the luminosity function object a verbose check is carried out in which the code automatically checks if all *main parameters* have been either defined as parameters or as a function of parameters.

## 1.3 1.2) Basic calculations

### 1.4 1.2.1) Calculating the luminosity function

To retrieve values of the luminosity function one can either directly call it supplying a luminosity and redshift (e.g., `mcgreer2013(M1450,redsh)`) or use the `.evaluate(lum,redsh)` function. In both cases either the luminosity or the redshift can also be specified as a `numpy.ndarray` instead of a single float value.

```
[6]: # Defining a magnitude array
M1450 = np.arange(-29, -22, 0.01)
# Defining a redshift array
z= np.arange(4, 6, 0.1)

# Calling the luminosity function with a luminosity and a redshift value produces one_
↳output value
print(mcgreer2013(-27, 4.9))

# Calling the luminosity function with a np.ndarray in luminosity and a redshift value_
↳produces a np.ndarray
print(mcgreer2013(M1450, 4.9).shape)

# Calling the luminosity function with a np.ndarray in redshift and a magnitude value_
↳produces a np.ndarray
print(mcgreer2013(-27, z).shape)
```

```
2.7377613276931382e-09
(700,)
(20,)
```

Let us now display the McGreer+2013 quasar luminosity function in comparison to the binned values from the paper, basically reproducing Fig.16 of the publication.

```
[7]: # Binned QLF data from the publication

dr7 = {'M1450':np.array([-28.05, -27.55, -27.05, -26.55, -26.05]),
       'logPhi':np.array([-9.45, -9.24, -8.51, -8.20, -7.9]),
       'sigma_logPhi':np.array([0.21, 0.26, 0.58, 0.91, 1.89])*1e-9}

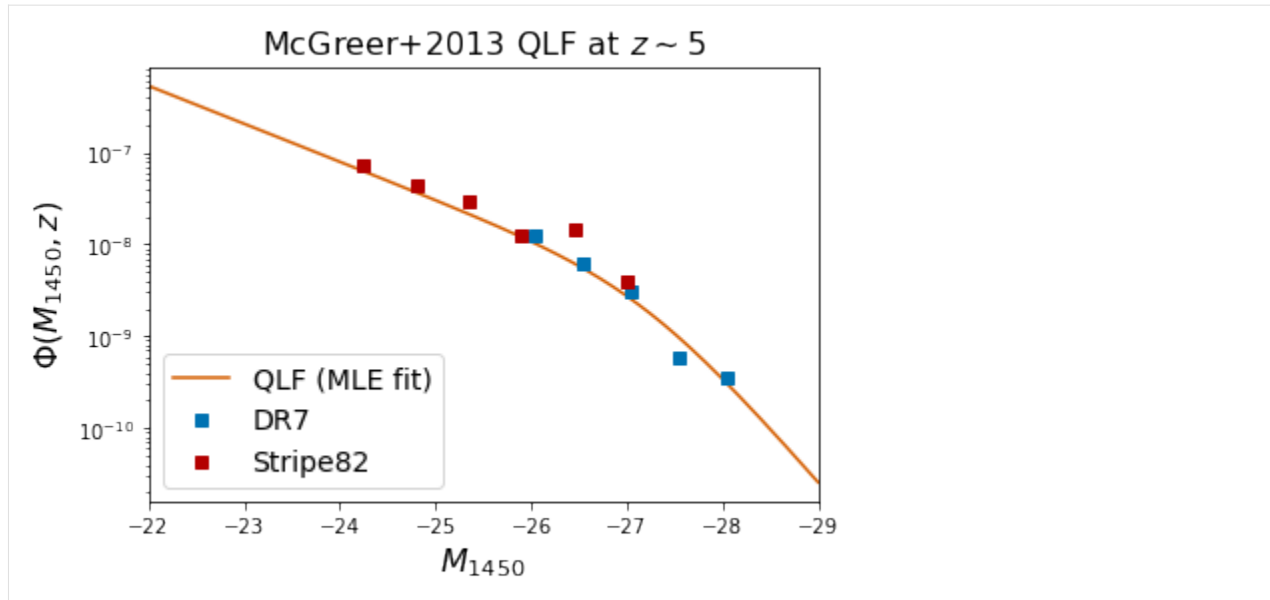
str82 = {'M1450':np.array([-27.0, -26.45, -25.9, -25.35, -24.8, -24.25]),
        'logPhi':np.array([-8.4, -7.84, -7.9, -7.53, -7.36, -7.14]),
        'sigma_logPhi':np.array([2.81, 6.97, 5.92, 10.23, 11.51, 19.9])*1e-9}

# Color definition for plotting
vermillion = (213/255., 94/255., 0)
dblue = (0, 114/255., 178/255.)
dred = (179/255., 0, 0)

qlf = mcgreer2013(M1450, 4.9)

plt.plot(M1450, qlf, lw=1.5, color=vermillion, label='QLF (MLE fit)')
plt.plot(dr7['M1450'], 10**dr7['logPhi'], 's', color=dblue, label='DR7')
plt.plot(str82['M1450'], 10**str82['logPhi'], 's', color=dred, label='Stripe82')
plt.xlim(-22, -29)

# Make the plot nice
plt.semilogy()
plt.ylabel('$\Phi(M_{1450},z)$', fontsize=16)
plt.xlabel('$M_{1450}$', fontsize=16)
plt.title('McGreer+2013 QLF at $z\sim5$', fontsize=16)
plt.legend(loc=3, fontsize=14)
plt.show()
```



## 1.5 1.2.2) Calculating the number of quasars in a redshift and luminosity range (per steradian)

The built-in class function `.integrate_over_lum_redsh` performs the double integral of the quasar luminosity function using the provided input parameters. For this calculation we need to specify a cosmology as the differential comoving solid volume element, which can be provided directly, depends on it.

```
[8]: print(mcgreer2013.integrate_over_lum_redsh.__doc__)
```

Calculate the number of sources described by the luminosity function over a luminosity and redshift interval in units of per steradian.

Either a cosmology or `dVdzd0` have to be supplied.

```
:param lum_range: Luminosity range
:type lum_range: tuple
:param redsh_range: Redshift range
:type redsh_range: tuple
:param dVdzd0: Differential comoving solid volume element (default =
    None)
:type dVdzd0: function
:param selfun: Selection function (default = None)
:type selfun: atelier.selfun.QsoSelectionFunction
:param cosmology: Cosmology (default = None)
:type cosmology: astropy.cosmology.Cosmology
:param kwargs:
:return: :math:`N = \int \int \Phi(L, z) (dV/(dz d\Omega)) dL dz`
:rtype: float
```

```
[11]: cosmology = FlatLambdaCDM(H0=70, Om0=0.272)

n_qso_per_steradian = mcgreer2013.integrate_over_lum_redsh([-29,-24],[5,5.5],
    ↪cosmology=cosmology)
print('There are {:.5f} quasars per steradian within z=5-5.5 and with M1450=-29 to -24.
    ↪(using romberg integration)'.format(n_qso_per_steradian))

n_qso_per_steradian = mcgreer2013.integrate_over_lum_redsh_simpson([-29,-24],[5,5.5],
    ↪cosmology=cosmology)
print('There are {:.5f} quasars per steradian within z=5-5.5 and with M1450=-29 to -24.
    ↪(using Simpsons rule on a grid)'.format(n_qso_per_steradian))

There are 936.08722 quasars per steradian within z=5-5.5 and with M1450=-29 to -24.
    ↪(using romberg integration)
There are 936.08637 quasars per steradian within z=5-5.5 and with M1450=-29 to -24.
    ↪(using Simpsons rule on a grid)
```

## 1.6 1.2.3) Calculating the quasar number density over a luminosity range (per Mpc<sup>3</sup>)

The built-in class function `.integrate_lum` performs this integral using the provided input parameters.

```
[10]: print(mcgreer2013.integrate_lum.__doc__)

Calculate the volumetric source density described by the luminosity
      function at a given redshift and over a luminosity interval in units of
      per Mpc^3.

      :param redsh: Redshift
      :type redsh: float
      :param lum_range: Luminosity range
      :type lum_range: tuple
      :param kwargs:
      :return: :math:\int \Phi(L,z) dL`
      :rtype: float
```

```
[11]: n_qso_per_Mpc3 = mcgreer2013.integrate_lum(4.9,[-29,-24])
print('The quasar density (in Mpc^-3) over M1450=-29 to -24 is {:.2e} at redshift z=4.9'.
    ↪format(n_qso_per_Mpc3))

The quasar density (in Mpc^-3) over M1450=-29 to -24 is 8.53e-08 at redshift z=4.9
```

## 1.7 1.2.4) Calculating the quasar number density over a luminosity range (per steradian)

The built-in class function `.redshift_density` performs this integral using the provided input parameters. This quantity is rarely used in normal calculations based on the quasar luminosity function.

```
[12]: print(mcgreer2013.redshift_density.__doc__)
```

Calculate the volumetric source density described by the luminosity function at a given redshift and over a luminosity interval in units of per steradian per redshift.

```
:param redsh: Redshift
:type redsh: float
:param lum_range: Luminosity range
:type lum_range: tuple
:param dVdzd0: Differential comoving solid volume element (default =
               None)
:type dVdzd0: function
:param kwargs:
:return: :math:\int \Phi(L,z) (dV/dz) d\Omega dL
:rtype: float
```

```
[13]: dVdzd0 = lumfun.interp_dVdzd0([4.0,6.0], cosmology)
```

```
n_qso_per_steradian_per_z = mcgreer2013.redshift_density(4.9, [-29, -24], dVdzd0)
print('The quasar density (per steradian and per redshift(!)) over M1450=-29 to -24 is {:  
↪.2e} at redshift z=4.9'.format(n_qso_per_steradian_per_z))
```

```
The quasar density (per steradian and per redshift(!)) over M1450=-29 to -24 is 2.80e+03  
↪at redshift z=4.9
```

### 1.7.1 1.3) Calculating the ionizing emissivity at 1450A (or 912A)

Another built-in class function, which is only defined for broken double power laws and single power laws (both based on magnitude) and for quasar luminosity functions with `lum_type = 'M1450'` is `.calc_ionizing_emissivity_at_1450A`. To properly test this function and compare it with published results on the ionizing emissivity we use the pre-defined quasar luminosity function (single power law parametrization) of Wang+2019.

ADS reference: <https://ui.adsabs.harvard.edu/abs/2019ApJ...884...30W/abstract>

```
[14]: wang2019spl = lumfun.WangFeige2019SPLQLF()
```

```
[INFO]-----
[INFO] Performing initialization checks
[INFO]-----
[INFO]-----
[INFO] Main parameter phi_star is supplied as a normal parameter.
[INFO]-----
[INFO] Main parameter alpha is supplied as a normal parameter.
[INFO]-----
```

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```
[INFO] Main parameter lum_ref is supplied as a normal parameter.
[INFO]-----
[INFO] Initialization check passed.
[INFO]-----
```

We use the same integration boundaries (M1450=-30 to -28) at a redshift of z=6.7

```
[15]: e1450 = wang2019spl.calc_ionizing_emissivity_at_1450A(6.7, [-30,-18])

# Adopting a power law slope of -0.6 we scale the ionizing emissivity at 1450A to 912A
↳ (Lyman break)
e912 = e1450*(1450./912)**-0.6
print('Calculated ionizing emissivity at 1450A e1450={:.2e}'.format(e1450))
print('Rescaled ionizing emissivity to 912A e912={:.2e}'.format(e912))
print('Ionizing emissivity at 912A calculated in Wang2019 for the single power law QLF
↳ e912=2.10e+23')

Calculated ionizing emissivity at 1450A e1450=2.77e+23
Rescaled ionizing emissivity to 912A e912=2.10e+23
Ionizing emissivity at 912A calculated in Wang2019 for the single power law QLF e912=2.
↳ 10e+23
```

## 1.7.2 2) Sampling from the luminosity function

A built-in class function allows to sample sources from a luminosity function. The sampling is done in a two step process. 1) The redshift range is divided into small bins and the number of expected sources per bin is calculated. 2) A cumulative distribution of the quasar luminosity function is calculated at the redshift bin middle and the number of expected sources for the redshift bin is taken. Sources for all bins are concatenated and returned.

**Note!** This sampling routine is in large part inspired by <https://github.com/imcgreer/simqso/blob/master/simqso/lumfun.py>, lines 219 and following.

```
[16]: print(mcgreer2013.sample.__doc__)

Sample the luminosity function over a given luminosity and
redshift range.

This sampling routine is in large part adopten from
https://github.com/imcgreer/simqso/blob/master/simqso/lumfun.py
, lines 219 and following.

If the integral over the luminosity function does not have an
analytical implementation, integrals are calculated using
integrate.romberg, which can take a substantial amount of time.

:param lum_range: Luminosity range
:type lum_range: tuple
:param redsh_range: Redshift range
:type redsh_range: tuple
:param cosmology: Cosmology (default = None)
:type cosmology: astropy.cosmology.Cosmology
:param sky_area: Area of the sky to be sampled in square degrees
```

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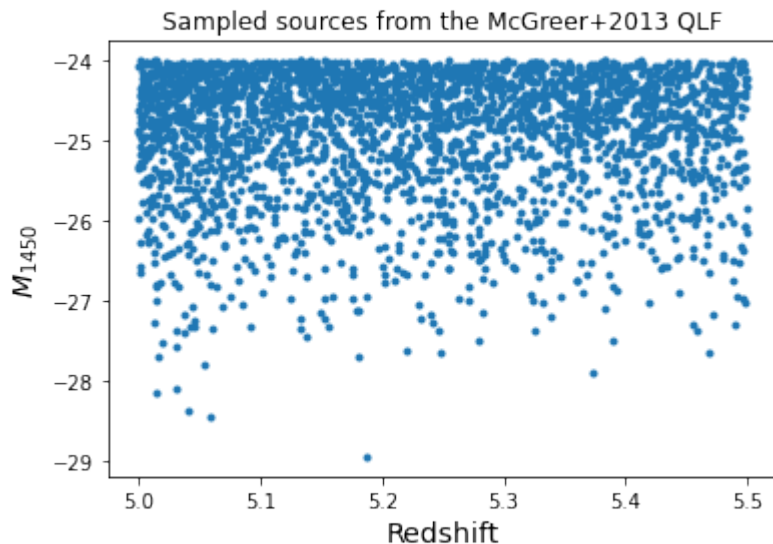
```
:type sky_area: float
:param seed: Random seed for the sampling
:type seed: int
:param lum_res: Luminosity resolution (default = 1e-2, equivalent to
    100 bins)
:type lum_res: float
:param redsh_res: Redshift resolution (default = 1e-2, equivalent to
    100 bins)
:type redsh_res: float
:param verbose: Verbosity
:type verbose: int
:return: Source sample luminosities and redshifts
:rtype: (numpy.ndarray,numpy.ndarray)
```

```
[17]: Mrange = [-29, -24]
      zrange = [5.0, 5.5]
      sky_area = 10000

      sample_M, sample_z = mcgreer2013.sample(Mrange, zrange, cosmology, sky_area)

[INFO] Integration returned 2851 sources
```

```
[18]: plt.plot(sample_z, sample_M, '.')
      plt.xlabel('Redshift', fontsize=14)
      plt.ylabel('$M_{1450}$', fontsize=14)
      plt.title('Sampled sources from the McGreer+2013 QLF')
      plt.show()
```



```
[ ]:
```





## ATELIER.LUMFUNFIT - INTRODUCTION

### 2.1 Introduction to the Atelier's luminosity function fit class with a strong emphasis on quasar luminosity functions

This notebook leads through an example of using the luminosity function fit class to determine the free parameters of a luminosity function.

The luminosity function fit uses MCMC (emcee) to sample the free parameters space minimizing the negative logarithmic likelihood defined in Marshall et al. (1983) Equation 3.

ADS reference: <https://ui.adsabs.harvard.edu/abs/1983ApJ...269...35M/abstract>

```
[1]: # General imports
import corner
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from astropy.cosmology import FlatLambdaCDM

# Importing the atelier modules
from atelier import lumfun
from atelier import lumfunfit
from atelier import survey
```

#### 2.1.1 1) Setting up the mock data

We start by generating a mock sample of quasars adopting the McGreer+2018 quasar luminosity function at  $z \sim 5$ . Using the lumfun class method “.sample” we generate quasars in 10000 square degrees over the defined luminosity and redshift range.

**Note:** The sample method relies on a large number of computational operations and has not been optimized, yet. So executing the function will take time.

```
[2]: cosmology = FlatLambdaCDM(H0=70, Om0=0.272)

lum_range = [-29, -24]
redsh_range = [5.0, 5.5]
sky_area = 10000

mcgreer2018 = lumfun.McGreer2018QLF()
```

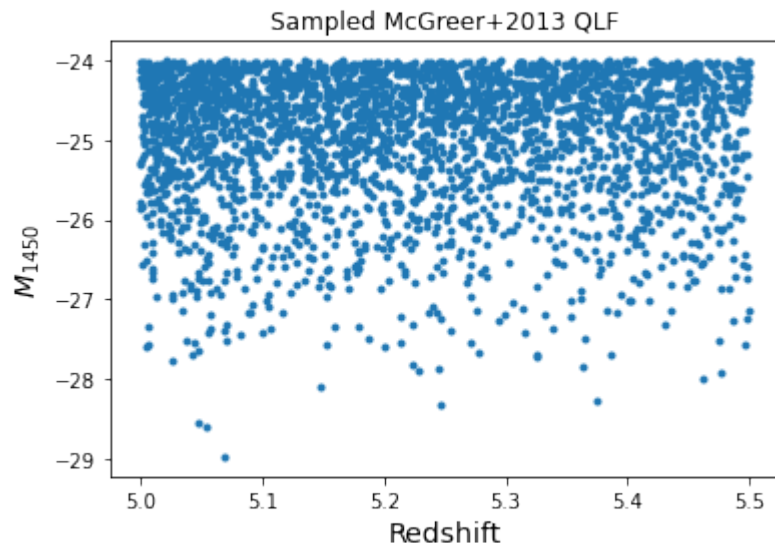
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```
sample_M, sample_z = mcgreer2018.sample(lum_range, redsh_range, cosmology, sky_area)
mock_df = pd.DataFrame(data={'absmag':sample_M, 'redsh':sample_z})
```

```
plt.plot(sample_z, sample_M, '.')
plt.xlabel('Redshift', fontsize=14)
plt.ylabel('$M_{1450}$', fontsize=14)
plt.title('Sampled McGreer+2013 QLF')
plt.show()
```

```
[INFO]-----
[INFO] Performing initialization checks
[INFO]-----
[INFO]-----
[INFO] Main parameter phi_star is described by a function.
[INFO] The function parameters are: ['redsh', 'log_phi_star_z6', 'k', 'z_ref']
[INFO] All parameters are supplied.
[INFO] Parameters "lum" and "redsh" were ignored as they are luminosity function
arguments.
[INFO]-----
[INFO] Main parameter lum_star is supplied as a normal parameter.
[INFO]-----
[INFO] Main parameter alpha is supplied as a normal parameter.
[INFO]-----
[INFO] Main parameter beta is supplied as a normal parameter.
[INFO]-----
[INFO] Initialization check passed.
[INFO]-----
[INFO] Integration returned 3013 sources
```



## 2.1.2 2) Setting up the survey class

The mock quasar sample in M1450 and redshift will now be stored in the “Survey” class of the *survey* module. The “Survey” class is a container for important information of an astronomical survey such as the luminosity and redshift of it’s sources and the overall survey area. As a keyword argument a selection function (see *selfun* module) can also be passed.

```
[3]: mock_survey = survey.Survey(mock_df, 'absmag', 'redsh', 10000)
```

## 2.1.3 3) Fitting the mock survey to a luminosity function model

## 2.2 3.1) Instantiating the luminosity function fit class

We begin by instantiating a LuminosityFunctionFit object before. The luminosity and redshift range for the fit are specified at this instance along with the cosmology and the list of surveys, whose data should be considered in the fit. For this example we only pass the mock\_survey object we generated above.

```
[4]: fit = lumfunfit.LuminosityFunctionFit([-29, -24],[5.0,5.5], cosmology, [mock_survey])
```

## 2.3 3.2) Setting up the luminosity function model for the fit

In the next step we define the quasar luminosity function model used in the fit. Our goal is to reproduce the values of the McGreer+2018 quasar luminosity function. So we begin by instantiating a new version of it.

```
[5]: lfmodel = lumfun.McGreer2018QLF()
lfmodel.print_parameters()

[INFO]-----
[INFO] Performing initialization checks
[INFO]-----
[INFO]-----
[INFO] Main parameter phi_star is described by a function.
[INFO] The function parameters are: ['redsh', 'log_phi_star_z6', 'k', 'z_ref']
[INFO] All parameters are supplied.
[INFO] Parameters "lum" and "redsh" were ignored as they are luminosity function
->arguments.
[INFO]-----
[INFO] Main parameter lum_star is supplied as a normal parameter.
[INFO]-----
[INFO] Main parameter alpha is supplied as a normal parameter.
[INFO]-----
[INFO] Main parameter beta is supplied as a normal parameter.
[INFO]-----
[INFO] Initialization check passed.
[INFO]-----
Parameter log_phi_star_z6 = -8.97, bounds=None, vary=True, unc=[0.18, 0.15]
Parameter lum_star = -27.47, bounds=None, vary=True, unc=[0.26, 0.22]
Parameter alpha = -1.97, bounds=None, vary=True, unc=[0.09, 0.09]
Parameter beta = -4.0, bounds=None, vary=True, unc=None
```

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```
Parameter k = -0.47, bounds=None, vary=True, unc=None
Parameter z_ref = 6, bounds=None, vary=True, unc=None
```

However, we need to modify the luminosity function parameters for the luminosity function fit. We begin by excluding three parameters from the fitting, which were fixed in the original quasar luminosity function determination. Technically we should also be able to reproduce these parameters, but we want to keep the example simple.

Next, we set fitting bounds to the other three parameters and set their values to an arbitrary value within the fitting bounds for the initialization of the fit. Now we are ready to carry out the fit.

```
[6]: # Set the parameters that are fixed in the model to not be varied during the fit.
lfmodel.parameters['z_ref'].vary=False
lfmodel.parameters['k'].vary=False
lfmodel.parameters['beta'].vary=False

# Set the parameter bounds for the fit parameters and set the starting value
lfmodel.parameters['log_phi_star_z6'].bounds = [-11,-7]
lfmodel.parameters['log_phi_star_z6'].value = -8
lfmodel.parameters['lum_star'].bounds = [-29.5, -25.5]
lfmodel.parameters['lum_star'].value = -28
lfmodel.parameters['alpha'].bounds = [-3, -1]
lfmodel.parameters['alpha'].value = -1.5

lfmodel.print_parameters()

Parameter log_phi_star_z6 = -8, bounds=[-11, -7], vary=True, unc=[0.18, 0.15]
Parameter lum_star = -28, bounds=[-29.5, -25.5], vary=True, unc=[0.26, 0.22]
Parameter alpha = -1.5, bounds=[-3, -1], vary=True, unc=[0.09, 0.09]
Parameter beta = -4.0, bounds=None, vary=False, unc=None
Parameter k = -0.47, bounds=None, vary=False, unc=None
Parameter z_ref = 6, bounds=None, vary=False, unc=None
```

## 2.4 3.3) Fitting the luminosity function model

To carry out the fit we call the `.run_mcmc` function supplying the luminosity function model and the number of steps for the MCMC chain. Default values for the MCMC are specified as attributes of the `LuminosityFunctionFit` class, are be set during initialization and can be modified at any time.

A bar indicates the progress of the MCMC sampling.

```
[ ]: fit.run_mcmc(lfmodel, steps=5000)
```

The `sampler` attribute of the `LuminosityFunctionFit` class holds all the information on the emcee sampler including the chain of samples. We can visualize the result of the fit by using emcee's sampler function to return a flat chain from the MCMC.

Using the `corner` package we plot the results of the MCMC for the three fitted parameters highlighting the true input values of the McGreer+2018 QLF from which the mock sample was created.

```
[ ]: # Plotting the covariance matrices of the final fit parameters

flat_samples = fit.sampler.get_chain(discard=100, thin=1, flat=True)
```

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(continued from previous page)

```
print(flat_samples)
labels=['log_phi_star', 'lum_star', 'alpha']
fig = corner.corner(flat_samples, labels=labels, truths=[-8.97, -27.47, -1.97])
print(np.nanpercentile(flat_samples[:,0],[16,50,84]))
print(np.nanpercentile(flat_samples[:,1],[16,50,84]))
print(np.nanpercentile(flat_samples[:,2],[16,50,84]))
```

The fitting results encompass the true input values within 1 sigma, while being biased slightly higher in all three parameters. At present, we suspect the luminosity function method `.sample` to be at the root of this issue as changing the sampling resolution affects is know to bias the fitting results. However, investigations are still ongoing.

### 2.4.1 4) Fitting the mock survey to a luminosity function model (multiprocessing!)

Depending on the number of object in the survey and the complexity of the luminosity function model, the MCMC fitting can take a few minutes. To speed up the progress we have implemented a special version of the MCMC fitting, making use of `emcee`'s native multiprocessing abilities.

The following cells provide a guide on how multiprocessing can be used to accelerate the fitting. The results initialized with the same default seed of the `LuminosityFunctionFit` class reproduce the result of the single core approach above.

```
[11]: from multiprocessing import cpu_count
```

```
ncpu = cpu_count()
print("{0} CPUs".format(ncpu))
```

```
12 CPUs
```

```
[12]: # Reset the initial guess
initial_guess = [-8, -28, -1.5]
# Run the MCMC chain
fit.run_mcmc_multiprocess(lfmodel, steps=5000, initial_guess=initial_guess,
                          processes=6)
```

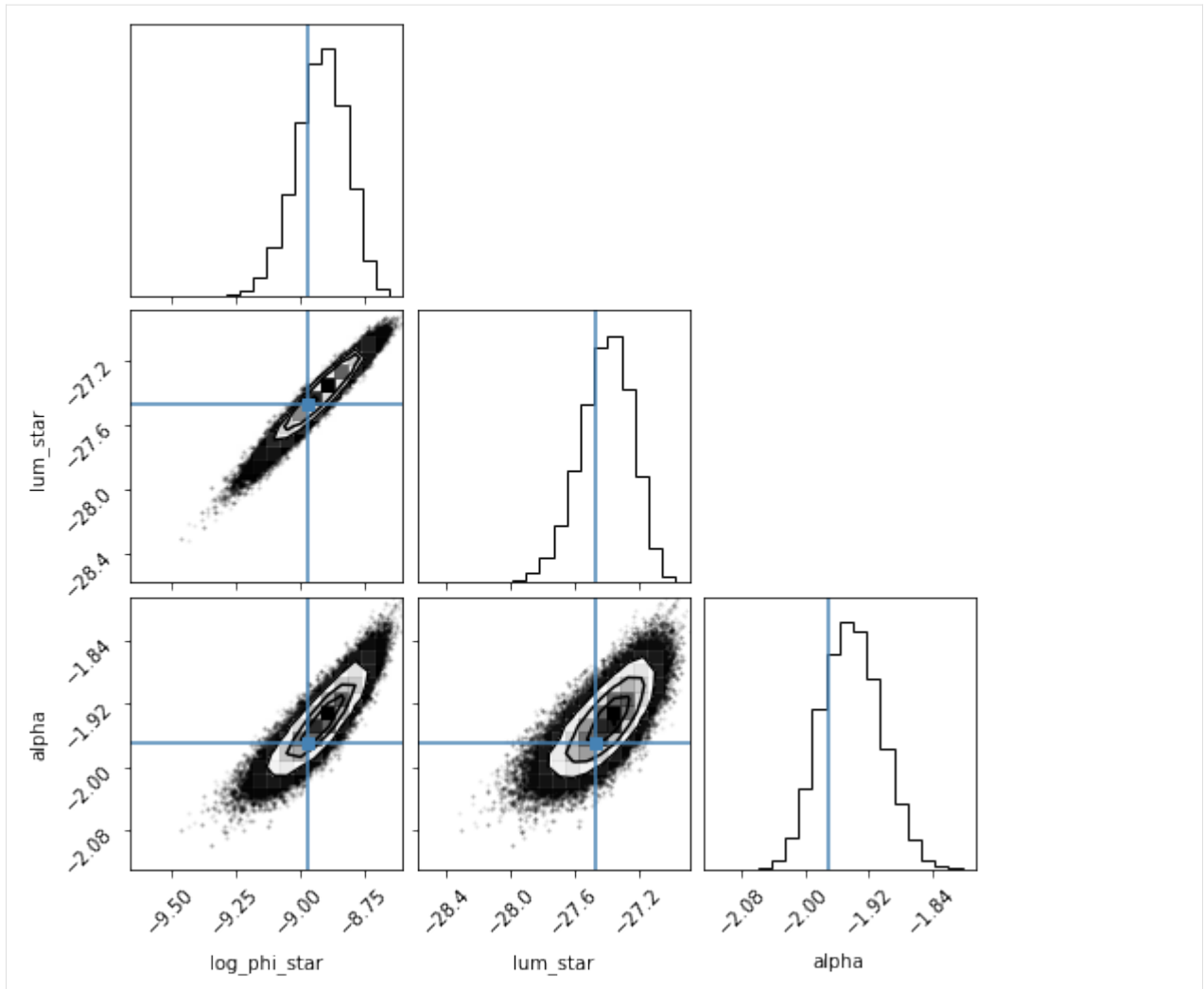
```
100%| 5000/5000 [04:13<00:00, 19.72it/s]
```

```
[INFO] Multiprocessing took 253.6 seconds
```

```
[13]: # Plotting the covariance matrices of the final fit parameters
```

```
flat_samples = fit.sampler.get_chain(discard=100, thin=1, flat=True)
labels=['log_phi_star', 'lum_star', 'alpha']
fig = corner.corner(flat_samples, labels=labels, truths=[-8.97, -27.47, -1.97])
print(np.nanpercentile(flat_samples[:,0],[16,50,84]))
print(np.nanpercentile(flat_samples[:,1],[16,50,84]))
print(np.nanpercentile(flat_samples[:,2],[16,50,84]))
```

```
[-9.01793936 -8.91553535 -8.82203563]
[-27.57611104 -27.40140614 -27.24798913]
[-1.97995996 -1.94293035 -1.90479171]
```



### 2.4.2 5) Fitting a luminosity function to multiple surveys

```
[ ]: mock_survey1 = survey.Survey(mock_df[:int(mock_df.shape[0]/2)], 'absmag', 'redsh',
                                   5000)
mock_survey2 = survey.Survey(mock_df[int(mock_df.shape[0]/2):], 'absmag', 'redsh',
                               5000)

[ ]: fit2 = lumfunfit.LuminosityFunctionFit([-29, -24],[5.0,5.5], cosmology, [mock_survey1,
                                                                              mock_survey2])

[ ]: initial_guess = [-8, -28, -1.5]
      # Run the MCMC chain
      fit2.run_mcmc(lfmodel, steps=5000, initial_guess=initial_guess)

[ ]: # Plotting the covariance matrices of the final fit parameters

      flat_samples = fit2.sampler.get_chain(discard=100, thin=1, flat=True)
```

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(continued from previous page)

```
labels=['log_phi_star', 'lum_star', 'alpha']
fig = corner.corner(flat_samples, labels=labels, truths=[-8.97, -27.47, -1.97])
print(np.nanpercentile(flat_samples[:,0],[16,50,84]))
print(np.nanpercentile(flat_samples[:,1],[16,50,84]))
print(np.nanpercentile(flat_samples[:,2],[16,50,84]))
```

### 2.4.3 6) Fitting the mock survey to a luminosity function model (integration mode: "simpson")

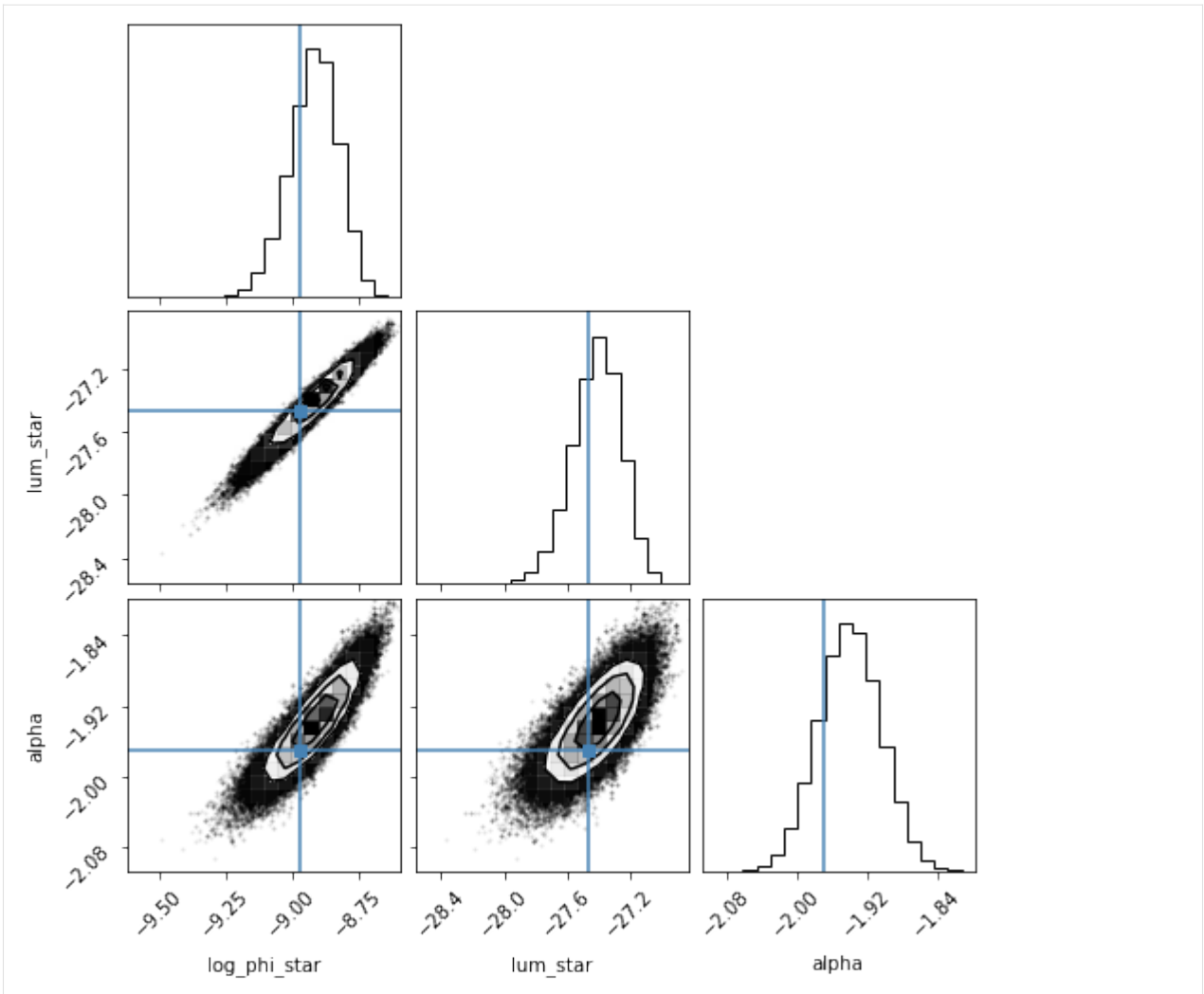
```
[7]: fit = lumfunfit.LuminosityFunctionFit([-29, -24],[5.0,5.5], cosmology, [mock_survey])
```

```
[8]: # Reset the initial guess
initial_guess = [-8, -28, -1.5]
# Run the MCMC chain
fit.run_mcmc(lfmodel, steps=5000, initial_guess=initial_guess, int_mode='simpson')
100%| 5000/5000 [02:13<00:00, 37.38it/s]
```

```
[10]: # Plotting the covariance matrices of the final fit parameters

flat_samples = fit.sampler.get_chain(discard=100, thin=1, flat=True)
# print(flat_samples)
labels=['log_phi_star', 'lum_star', 'alpha']
fig = corner.corner(flat_samples, labels=labels, truths=[-8.97, -27.47, -1.97])
print(np.nanpercentile(flat_samples[:,0],[16,50,84]))
print(np.nanpercentile(flat_samples[:,1],[16,50,84]))
print(np.nanpercentile(flat_samples[:,2],[16,50,84]))

[-9.01302699 -8.91593096 -8.82950664]
[-27.57408329 -27.40910229 -27.26364548]
[-1.97509396 -1.94016805 -1.90512569]
```



[ ]:



## THE LUMINOSITY FUNCTION MODULE

**class atelier.lumfun.Akiyama2018QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Akiyama+2018 at  $z \sim 4$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2018ApJ...869..150M/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

This implementation adopts the double power law fit presented in Table 5 (“Maximum Likelihood”).

**class atelier.lumfun.Boutsia2021\_QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Boutsia+2021 at  $z \sim 3.9$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2021ApJ...912..111B/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

The fit in Boutsia+2021 includes QLF determinations of Fontanot+2007, Glikman+2011, Boutsia+2018, and Giallongo+2019 at fainter magnitudes.

This implementation adopts the double power law fit presented in Table 4.

**class atelier.lumfun.DoublePowerLawLF**(*parameters, param\_functions, lum\_type=None, cosmology=None, ref\_cosmology=None, ref\_redsh=None, verbose=1*)

Luminosity function, which takes the functional form of a double power law with the luminosity in absolute magnitudes.

The luminosity function has four main parameters:

- “phi\_star”: the overall normalization
- “lum\_star”: the break luminosity/magnitude where the power law slopes change.
- “alpha”: the first power law slope
- “beta”: the second power law slope

**calc\_ionizing\_emissivity\_at\_1450A**(*redsh, lum\_range, \*\*kwargs*)

Calculate the ionizing emissivity at rest-frame 1450Å,  $\epsilon_{1450}$ , in units of  $\text{erg s}^{-1} \text{Hz}^{-1} \text{Mpc}^{-3}$ .

This function integrates the luminosity function at redshift “redsh” over the luminosity interval “lum\_range” to calculate the ionizing emissivity at rest-frame 1450Å.

Calling this function is only valid if the luminosity function “lum\_type” argument is “lum\_type=”M1450”.

### Parameters

- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation

- **lum\_range** (*tuple*) – Luminosity range

**Returns**

Ionizing emissivity ( $\text{erg s}^{-1} \text{Hz}^{-1} \text{Mpc}^{-3}$ )

**Return type**

float

**evaluate**(*lum, redsh, parameters=None*)

Evaluate the double power law as a function of magnitude (“lum”) and redshift (“redsh”).

Function to be evaluated: `atelier.lumfun.mag_double_power_law()`

**Parameters**

- **lum** (*float or numpy.ndarray*) – Luminosity for evaluation
- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **parameters** (*dict(atelier.lumfun.Parameters)*) – Dictionary of parameters used for this specific calculation. This does not replace the parameters with which the luminosity function was initialized. (default=None)

**Returns**

Luminosity function value

**Return type**

(*numpy.ndarray, numpy.ndarray*)

**class atelier.lumfun.Giallongo2019\_4p5\_QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Giallongo+2019 at  $z \sim 4.5$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2019ApJ...884...19G/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

The fit in Giallongo+2019 (model 1) includes QLF determinations of Fontanot+2007, Boutsia+2018, and Akiyama+2018 at brighter magnitudes.

This implementation adopts the double power law fit presented in Table 3 (model 1).

**class atelier.lumfun.Giallongo2019\_5p6\_QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Giallongo+2019 at  $z \sim 5.6$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2019ApJ...884...19G/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

The fit in Giallongo+2019 (model 4) includes data from CANDELS and SDSS.

This implementation adopts the double power law fit presented in Table 3 (model 4).

**class atelier.lumfun.Hopkins2007QLF**

Implementation of the bolometric quasar luminosity function of Hopkins+2007.

ADS reference: <https://ui.adsabs.harvard.edu/abs/2007ApJ...654..731H/abstract>

The luminosity function is described by Equations 6, 9, 10, 17, 19. The values for the best fit model adopted here are found in Table 3 in the row named “Full”.

**static alpha**(*redsh, z\_ref, gamma\_one, kg1*)

Calculate the redshift dependent luminosity function slope alpha.

Equations 10, 17

**Parameters**

- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **z\_ref** (*float*) – Reference redshift
- **gamma\_one** (*float*) – Luminosity function power law slope at *z\_ref*
- **kg1** (*float*) – Evolutionary parameter

**Returns**

**static beta**(*redsh, z\_ref, gamma\_two, kg2\_1, kg2\_2*)

Calculate the redshift dependent luminosity function slope beta.

Equations 10, 19 and text on page 744 (bottom right column)

**Parameters**

- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **z\_ref** (*float*) – Reference redshift
- **gamma\_two** (*float*) – Luminosity function power law slope at *z\_ref*
- **kg2\_1** (*float*) – Evolutionary parameter
- **kg2\_2** (*float*) – Evolutionary parameter

**Returns**

**evaluate**(*lum, redsh, parameters=None*)

Evaluate the Hopkins+2007 bolometric luminosity function.

Function to be evaluated: atelier.lumfun.lum\_double\_power\_law()

**Parameters**

- **lum** (*float or numpy.ndarray*) – Luminosity for evaluation
- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **parameters** (*dict(atelier.lumfun.Parameters)*) – Dictionary of parameters used for this specific calculation. This does not replace the parameters with which the luminosity function was initialized. (default=None)

**Returns**

Luminosity function value

**Return type**

(numpy.ndarray, numpy.ndarray)

**static lum\_star**(*redsh, z\_ref, log\_lum\_star, kl1, kl2, kl3*)

Calculate the redshift dependent break luminosity (Eq. 9, 10)

**Parameters**

- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **z\_ref** (*float*) – Reference redshift
- **log\_lum\_star** (*float*) – Logarithmic break magnitude at *z\_ref*

- **kl1** (*float*) – Function parameter kl1
- **kl2** (*float*) – Function parameter kl2
- **kl3** (*float*) – Function parameter kl3

**Returns**

Redshift dependent break luminosity

**Return type**

float

**static phi\_star**(*log\_phi\_star*)

Calculate the break luminosity number density

**Parameters**

**log\_phi\_star** – Logarithmic break luminosity number density

**Returns**

**class atelier.lumfun.JiangLinhua2016QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Jiang+2016 at  $z \sim 6$ .

ADS reference:

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450A, M1450.

This implementation adopts the double power law fit described in Section 4.5

**static phi\_star**(*redsh, phi\_star\_z6, k, z\_ref*)

Calculate the redshift dependent luminosity function normalization.

**Parameters**

- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **phi\_star\_z6** (*float*) – Logarithmic source density at  $z=6$
- **k** (*float*) – Power law exponent of density evolution
- **z\_ref** (*float*) – Reference redshift

**Returns**

**class atelier.lumfun.Kim2020\_QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Kim+2020 at  $z \sim 5$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2020ApJ...904..111K/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450A, M1450.

The fit in Kim+2020 includes data from Yang+2016 at brighter luminosities.

This implementation adopts the double power law fit presented in Table 6, Case 1.

**static phi\_star**(*redsh, log\_phi\_star*)

**Parameters**

- **redsh** –
- **log\_phi\_star** –

**Returns**

**class** atelier.lumfun.**Kulkarni2019QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Kulkarni+2019 at  $z \sim 1-6$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2019MNRAS.488.1035K/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

This implementation adopts the double power law fit presented in Table 3 (“Model 3”).

**static** **alpha**(*redsh, c2\_0, c2\_1*)

**Parameters**

- **redsh** –
- **c2\_0** –
- **c2\_1** –

**Returns**

**static** **beta**(*redsh, c3\_0, c3\_1*)

**Parameters**

- **redsh** –
- **c3\_0** –
- **c3\_1** –

**Returns**

**static** **lum\_star**(*redsh, c1\_0, c1\_1, c1\_2, c1\_3*)

**Parameters**

- **redsh** –
- **c1\_0** –
- **c1\_1** –
- **c1\_2** –
- **c1\_3** –

**Returns**

**static** **phi\_star**(*redsh, c0\_0, c0\_1, c0\_2*)

**Parameters**

- **redsh** –
- **c0\_0** –
- **c0\_1** –
- **c0\_2** –

**Returns**

```
class atelier.lumfun.LuminosityFunction(parameters, param_functions, main_parameters,
                                         lum_type=None, cosmology=None, ref_cosmology=None,
                                         ref_redsh=None, verbose=1)
```

The base luminosity function class.

In this implementation a luminosity function is defined in terms of

- luminosity (“lum”) and
- redshift (“redsh”)
- a list of main parameters (“main\_parameters”)

The number of main parameters depend on the functional form and can themselves be functions (“param\_functions”) of luminosity, redshift or additional “parameters”.

This general framework should facilitate the implementation of a wide range of luminosity functions without confining limits.

An automatic initialization will check whether the parameter functions and parameters define all necessary main parameters.

While the code does not distinguish between continuum luminosities, broad band luminosities or magnitudes, some inherited functionality is based on specific luminosity definitions. In order for these functions to work a luminosity type “lum\_type” has to be specified. The following luminosity types have special functionality:

- “M1450” : Absolute continuum magnitude measured at 1450A in the rest-frame.

#### **parameters**

Dictionary of Parameter objects, which are used either as a main parameters for the calculation of the luminosity function or as an argument for calculating a main parameter using a specified parameter function “param\_function”.

##### **Type**

dict(atelier.lumfun.Parameter)

#### **param\_functions**

Dictionary of functions with argument names for which the parameter attribute provides a Parameter or the luminosity “lum” or the redshift “redsh”.

##### **Type**

dict(functions)

#### **main\_parameters**

List of string providing names for the main parameters of the luminosity function. During the initialization the main parameters need to be either specified as a Parameter within the “parameters” attribute or as a function by the “param\_functions” attribute.

##### **Type**

list(string)

#### **lum\_type**

Luminosity type checked for specific functionality

##### **Type**

string (default=None)

#### **verbose**

Verbosity (0: no output, 1: minimal output)

##### **Type**

int

**evaluate**(*lum*, *redsh*, *parameters=None*)

Evaluate the luminosity function at the given luminosity and redshift.

:raise NotImplementedError

**Parameters**

- **lum** (*float or numpy.ndarray*) – Luminosity for evaluation
- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **parameters** (*dict(atelier.lumfun.Parameters)*) – Dictionary of parameters used for this specific calculation. This does not replace the parameters with which the luminosity function was initialized. (default=None)

**evaluate\_main\_parameters**(*lum*, *redsh*, *parameters=None*)

Evaluate the main parameters of the luminosity function

**Parameters**

- **lum** (*Luminosity for evaluation*) – float or numpy.ndarray
- **redsh** (*Redshift for evaluation*) – float or numpy.ndarray
- **parameters** (*dict(atelier.lumfun.Parameters)*) – Dictionary of parameters used for this specific calculation. This does not replace the parameters with which the luminosity function was initialized.

**Returns**

**get\_free\_parameter\_names**()

Return a list of names with all parameters for which vary == True.

**Returns**

Names of parameters with vary == True

**Return type**

list(str)

**get\_free\_parameters**()

Return a dictionary with all parameters for which vary == True.

**Returns**

All parameters with vary == True

**Return type**

dict(*atelier.lumfun.Parameter*)

**integrate\_lum**(*redsh*, *lum\_range*, *\*\*kwargs*)

Calculate the volumetric source density described by the luminosity function at a given redshift and over a luminosity interval in units of per Mpc<sup>3</sup>.

**Parameters**

- **redsh** (*float*) – Redshift
- **lum\_range** (*tuple*) – Luminosity range
- **kwargs** –

**Returns**

$\int \Phi(L, z) dL$

**Return type**

float

**integrate\_over\_lum\_redsh**(*lum\_range*, *redsh\_range*, *dVdzdO*=None, *selfun*=None, *cosmology*=None, *\*\*kwargs*)

Calculate the number of sources described by the luminosity function over a luminosity and redshift interval in units of per steradian.

Either a cosmology or dVdzdO have to be supplied.

#### Parameters

- **lum\_range** (*tuple*) – Luminosity range
- **redsh\_range** (*tuple*) – Redshift range
- **dVdzdO** (*function*) – Differential comoving solid volume element (default = None)
- **selfun** ([atelier.selfun.QsoSelectionFunction](#)) – Selection function (default = None)
- **cosmology** (*astropy.cosmology.Cosmology*) – Cosmology (default = None)
- **kwargs** –

#### Returns

$$N = \int \int \Phi(L, z) (dV/(dzd\Omega)) dL dz$$

#### Return type

float

**integrate\_over\_lum\_redsh\_appmag\_limit**(*lum\_range*, *redsh\_range*, *appmag\_limit*, *kcorrection*, *dVdzdO*=None, *selfun*=None, *cosmology*=None, *initial\_lum\_bin\_width*=0.1, *initial\_redsh\_bin\_width*=0.05, *minimum\_probability*=0.001, *\*\*kwargs*)

#### Parameters

- **lum\_range** –
- **redsh\_range** –
- **appmag\_limit** –
- **kcorrection** –
- **dVdzdO** –
- **selfun** –
- **cosmology** –
- **initial\_lum\_bin\_width** –
- **initial\_redsh\_bin\_width** –
- **minimum\_probability** –
- **kwargs** –

#### Returns

**integrate\_over\_lum\_redsh\_simpson**(*lum\_range*, *redsh\_range*, *dVdzdO*=None, *selfun*=None, *cosmology*=None, *initial\_lum\_bin\_width*=0.1, *initial\_redsh\_bin\_width*=0.05, *minimum\_probability*=0.001, *\*\*kwargs*)

Calculate the number of sources described by the luminosity function over a luminosity and redshift interval in units of per steradian.



The integration is done on a grid using the Simpson rule.

This allows the selection function to be precalculated on the grid values for speed up of the integration process.

This code is in large part adopted from <https://github.com/imcgreer/simqso/blob/master/simqso/lumfun.py> lines 591 and following.

Either a cosmology or dVdzdO have to be supplied.

#### Parameters

- **lum\_range** (*tuple*) – Luminosity range
- **redsh\_range** (*tuple*) – Redshift range
- **dVdzdO** (*function*) – Differential comoving solid volume element (default = None)
- **selfun** (`atelier.selfun.QsoSelectionFunction`) – Selection function (default = None)
- **cosmology** (`astropy.cosmology.Cosmology`) – Cosmology (default = None)
- **kwargs** –

#### Returns

$$N = \int \int \Phi(L, z) (dV / (dz d\Omega)) dL dz$$

#### Return type

float

**integrate\_to\_luminosity\_density**(*lum\_range, redsh, \*\*kwargs*)

#### Parameters

- **lum\_range** –
- **redsh** –
- **dVdzdO** –

#### Returns

$$\int \Phi(L, z) L (dV / dz) d\Omega dL$$

**print\_free\_parameters**()

Print a list of all free (vary==True) parameters.

**print\_parameters**()

Print a list of all parameters.

**redshift\_density**(*redsh, lum\_range, dVdzdO, \*\*kwargs*)

Calculate the volumetric source density described by the luminosity function at a given redshift and over a luminosity interval in units of per steradian per redshift.

#### Parameters

- **redsh** (*float*) – Redshift
- **lum\_range** (*tuple*) – Luminosity range
- **dVdzdO** (*function*) – Differential comoving solid volume element (default = None)
- **kwargs** –

#### Returns

$$\int \Phi(L, z) (dV / dz) d\Omega dL$$

**Return type**

float

**sample**(*lum\_range*, *redsh\_range*, *cosmology*, *sky\_area*, *seed*=1234, *lum\_res*=0.01, *redsh\_res*=0.01, *verbose*=1, *\*\*kwargs*)

**Sample the luminosity function over a given luminosity and**

redshift range.

This sampling routine is in large part adopted from <https://github.com/imcgreer/simqso/blob/master/simqso/lumfun.py>, lines 219 and following.

If the integral over the luminosity function does not have an analytical implementation, integrals are calculated using `integrate.romberg`, which can take a substantial amount of time.

**Parameters**

- **lum\_range** (*tuple*) – Luminosity range
- **redsh\_range** (*tuple*) – Redshift range
- **cosmology** (*astropy.cosmology.Cosmology*) – Cosmology (default = None)
- **sky\_area** (*float*) – Area of the sky to be sampled in square degrees
- **seed** (*int*) – Random seed for the sampling
- **lum\_res** (*float*) – Luminosity resolution (default = 1e-2, equivalent to 100 bins)
- **redsh\_res** (*float*) – Redshift resolution (default = 1e-2, equivalent to 100 bins)
- **verbose** (*int*) – Verbosity

**Returns**

Source sample luminosities and redshifts

**Return type**

(*numpy.ndarray*, *numpy.ndarray*)

**update()**

Update the lumfun class parameters after a manual input.

**Returns**

**update\_free\_parameter\_values**(*values*)

Update all free parameters with new values.

**Parameters**

**values** (*list(float)*) – Values in the same order as the order of free parameters.

**class atelier.lumfun.Matsuoka2018QLF**(*cosmology*=None)

Implementation of the type-I quasar UV(M1450) luminosity function of Matsuoka+2018 at  $z \sim 6$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2018ApJ...869..150M/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

This implementation adopts the double power law fit presented in Table 5 (“standard”).

**static phi\_star**(*redsh*, *phi\_star\_z6*, *k*, *z\_ref*)

Calculate the redshift dependent luminosity function normalization.

**Parameters**

- **redsh** (*float* or *numpy.ndarray*) – Redshift for evaluation

- **phi\_star\_z6** (*float*) – Logarithmic source density at  $z=6$
- **k** (*float*) – Power law exponent of density evolution
- **z\_ref** (*float*) – Reference redshift

#### Returns

**class** atelier.lumfun.**Matsuoka2023QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Matsuoka+2023 at  $z\sim 7$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2023arXiv230511225M/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

This implementation adopts the double power law fit presented in Table 7 (standard model).

**static phi\_star**(*redsh*, *phi\_star\_z7*, *k*, *z\_ref*)

Calculate the redshift dependent luminosity function normalization.

#### Parameters

- **redsh** (*float* or *numpy.ndarray*) – Redshift for evaluation
- **phi\_star\_z7** (*float*) – Source density at  $z=6$
- **k** (*float*) – Power law exponent of density evolution
- **z\_ref** (*float*) – Reference redshift

#### Returns

**class** atelier.lumfun.**McGreer2018QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of McGreer+2018 at  $z\sim 5$  ( $z=4.9$ ).

ADS reference: <https://ui.adsabs.harvard.edu/abs/2018AJ...155..131M/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

This implementation adopts the best maximum likelihood estimate fit from the second column in Table 2.

**static phi\_star**(*redsh*, *log\_phi\_star\_z6*, *k*, *z\_ref*)

Calculate the redshift dependent luminosity function normalization.

#### Parameters

- **redsh** (*float* or *numpy.ndarray*) – Redshift for evaluation
- **log\_phi\_star\_z6** (*float*) – Logarithmic source density at  $z=6$
- **k** (*float*) – Power law exponent of density evolution
- **z\_ref** (*float*) – Reference redshift

#### Returns

**class** atelier.lumfun.**Niida2020\_QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Niida+2020 at  $z\sim 5$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2020ApJ...904...89N/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

The fit in Kim+2020 includes data from SDSS at brighter luminosities.

This implementation adopts the double power law fit presented in Table 6, Case 1.

**class** atelier.lumfun.Onken2022\_Niida\_4p52\_QLF(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Onken+2022 at  $z \sim 4.52$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2021arXiv210512215O/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

The maximum likelihood fit in Onken+2022 includes constraints of Niida+2020 at fainter magnitudes.

This implementation adopts the double power law fit presented in Table 4.

**class** atelier.lumfun.Onken2022\_Niida\_4p83\_QLF(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Onken+2022 at  $z \sim 4.83$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2021arXiv210512215O/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

The maximum likelihood fit in Onken+2022 includes constraints of Niida+2020 at fainter magnitudes.

This implementation adopts the double power law fit presented in Table 4.

**class** atelier.lumfun.PanZhiwei2022\_3p8\_QLF(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Zhiwei Pan+2022 at  $z \sim 3.8$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2022ApJ...928..172P/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

This implementation adopts the double power law fit presented in Table 5 denoted as (O+S+L) “Best fit”.

**static phi\_star**(*redsh, log\_phi\_star*)

Calculate the redshift dependent luminosity function normalization.

#### Parameters

- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **log\_phi\_star** (*float*) – Logarithmic source density at  $z=6$

#### Returns

**class** atelier.lumfun.PanZhiwei2022\_4p25\_QLF(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Zhiwei Pan+2022 at  $z \sim 4.25$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2022ApJ...928..172P/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

This implementation adopts the double power law fit presented in Table 5 denoted as (O+S+L) “Best fit”.

**static phi\_star**(*redsh, log\_phi\_star*)

Calculate the redshift dependent luminosity function normalization.

#### Parameters

- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **log\_phi\_star** (*float*) – Logarithmic source density at  $z=6$

### Returns

**class atelier.lumfun.PanZhiwei2022\_4p7\_QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Zhiwei Pan+2022 at  $z \sim 4.7$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2022ApJ...928..172P/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

This implementation adopts the double power law fit presented in Table 5 denoted as (O+S+L) “Best fit”.

**static phi\_star**(*redsh, log\_phi\_star*)

Calculate the redshift dependent luminosity function normalization.

### Parameters

- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **log\_phi\_star** (*float*) – Logarithmic source density at  $z=6$

### Returns

**class atelier.lumfun.Parameter**(*value, name, bounds=None, vary=True, one\_sigma\_unc=None*)

A class providing a data container for a parameter used in the luminosity function class.

#### value

Value of the parameter

#### Type

float

#### name

Name of the parameter

#### Type

string

#### bounds

Bounds of the parameter, used in fitting

#### Type

tupler

#### vary

Boolean to indicate whether this parameter should be varied, used in fitting

#### Type

bool

#### one\_sigma\_unc

1 sigma uncertainty of the parameter.

#### Type

list (2 elements)

**class atelier.lumfun.Richards2006QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Richards+2006 ( $z=1-5$ ).

ADS reference:

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes  $M_i(z=2)$ . We convert the  $M_i(z=2)$  magnitudes to M1450 using a simple conversion factor (see Ross+2013).

$$M1450(z=0) = M_i(z=2) + 1.486$$

This implementation adopts the double power law fit presented in Table 7 (variable power law).

**evaluate**(*lum*, *redsh*, *parameters=None*)

Evaluate the single power law as a function of magnitude (“lum”) and redshift (“redsh”) for the Richards 2006 QLF.

Function to be evaluated: `atelier.lumfun.richards_single_power_law()`

**Parameters**

- **lum** (*float* or *numpy.ndarray*) – Luminosity for evaluation
- **redsh** (*float* or *numpy.ndarray*) – Redshift for evaluation
- **parameters** (*dict*(`atelier.lumfun.Parameters`)) – Dictionary of parameters used for this specific calculation. This does not replace the parameters with which the luminosity function was initialized. (default=None)

**Returns**

Luminosity function value

**Return type**

(*numpy.ndarray*, *numpy.ndarray*)

**static phi\_star**(*redsh*, *log\_phi\_star*)

**Parameters**

- **redsh** –
- **log\_phi\_star** –

**Returns**

**class** `atelier.lumfun.SchechterLF`(*parameters*, *param\_functions*, *lum\_type=None*, *ref\_cosmology=None*, *ref\_redsh=None*, *cosmology=None*, *verbose=1*)

Schechter luminosity function

**evaluate**(*lum*, *redsh*, *parameters=None*)

Evaluate the Schechter function as a function of magnitude (“lum”) and redshift (“redsh”).

Function to be evaluated: `atelier.lumfun.mag_schechter`

**Parameters**

- **lum** (*float* or *numpy.ndarray*) – Luminosity for evaluation
- **redsh** (*float* or *numpy.ndarray*) – Redshift for evaluation
- **parameters** (*dict*(`atelier.lumfun.Parameters`)) – Dictionary of parameters used for this specific calculation. This does not replace the parameters with which the luminosity function was initialized. (default=None)

**Returns**

Luminosity function value

**Return type**

(*numpy.ndarray*, *numpy.ndarray*)

**class atelier.lumfun.Schindler2019\_2p9\_QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Schindler+2019 at  $z \sim 2.9$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2019ApJ...871..258S/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450A, M1450.

The maximum likelihood fit in Schindler+2019 includes the quasar sample of Ross+2013 at fainter magnitudes.

This implementation adopts the double power law fit presented in Table 7.

**class atelier.lumfun.Schindler2019\_3p25\_QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Schindler+2019 at  $z \sim 3.25$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2019ApJ...871..258S/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450A, M1450.

The maximum likelihood fit in Schindler+2019 includes the quasar sample of Ross+2013 at fainter magnitudes.

This implementation adopts the double power law fit presented in Table 7.

**class atelier.lumfun.Schindler2019\_3p75\_QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Schindler+2019 at  $z \sim 3.75$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2019ApJ...871..258S/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450A, M1450.

The maximum likelihood fit in Schindler+2019 includes the quasar sample of Richards+2096 at fainter magnitudes.

This implementation adopts the double power law fit presented in Table 7.

**class atelier.lumfun.Schindler2019\_4p25\_QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Schindler+2019 at  $z \sim 4.25$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2019ApJ...871..258S/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450A, M1450.

The maximum likelihood fit in Schindler+2019 includes the quasar sample of Richards+2096 at fainter magnitudes.

This implementation adopts the double power law fit presented in Table 7.

**class atelier.lumfun.Schindler2019\_LEDE\_QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Schindler+2019 with LEDE evolution.

ADS reference: <https://ui.adsabs.harvard.edu/abs/2019ApJ...871..258S/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450A, M1450.

The LEDE model was fit to a range of binned QLF measurements using chi-squared minimization.

This implementation adopts the LEDE double power law fit presented in Table 8.

**static lum\_star**(*redsh*, *lum\_star\_z2p2*, *c2*)

Calculate the redshift dependent break magnitude.

**Parameters**

- **redsh** (*float* or *numpy.ndarray*) – Redshift for evaluation
- **lum\_star\_z2p2** (*float*) – Break magnitude at  $z=2.2$
- **c2** (*float*) – Redshift evolution parameter

**Returns**

**static phi\_star**(*redsh*, *log\_phi\_star\_z2p2*, *c1*)

Calculate the redshift dependent luminosity function normalization.

**Parameters**

- **redsh** (*float* or *numpy.ndarray*) – Redshift for evaluation
- **log\_phi\_star\_z2p2** (*float*) – Logarithmic source density at  $z=2.2$
- **c1** (*float*) – Redshift evolution parameter

**Returns**

**class atelier.lumfun.Schindler2023QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Schindler+2023 at  $z\sim 6$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2023ApJ...943...67S/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

This implementation adopts the double power law fit presented in Table 4 (first row).

**static phi\_star**(*redsh*, *log\_phi\_star\_z6*, *k*, *z\_ref*)

Calculate the redshift dependent luminosity function normalization.

**Parameters**

- **redsh** (*float* or *numpy.ndarray*) – Redshift for evaluation
- **phi\_star\_z6** (*float*) – Logarithmic source density at  $z=6$
- **k** (*float*) – Power law exponent of density evolution
- **z\_ref** (*float*) – Reference redshift

**Returns**

**class atelier.lumfun.ShenXuejian2020QLF**

Shen+2020 bolometric quasar luminosity function; global fit B

**static alpha**(*redsh*, *a0*, *a1*, *z\_ref*)

**Parameters**

- **redsh** –
- **a0** –
- **a1** –
- **z\_ref** –

**Returns**



**static beta**(*redsh, b0, b1, b2, z\_ref*)

**Parameters**

- **redsh** –
- **b0** –
- **b1** –
- **b2** –
- **z\_ref** –

**Returns**

**evaluate**(*lum, redsh, parameters=None*)

Evaluate the Shen+2020 bolometric luminosity function.

Function to be evaluated: atelier.lumfun.lum\_double\_power\_law()

**Parameters**

- **lum** (*float or numpy.ndarray*) – Luminosity for evaluation
- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **parameters** (*dict(atelier.lumfun.Parameters)*) – Dictionary of parameters used for this specific calculation. This does not replace the parameters with which the luminosity function was initialized. (default=None)

**Returns**

Luminosity function value

**Return type**

(numpy.ndarray, numpy.ndarray)

**static lum\_star**(*redsh, c0, c1, c2, z\_ref*)

**Parameters**

- **redsh** –
- **c0** –
- **c1** –
- **c2** –
- **z\_ref** –

**Returns**

**class atelier.lumfun.SinglePowerLawLF**(*parameters, param\_functions, lum\_type=None, ref\_cosmology=None, ref\_redsh=None, cosmology=None, verbose=1*)

Luminosity function, which takes the functional form of a single power law with the luminosity in absolute magnitudes.

The luminosity function has three main parameters:

- “phi\_star”: the overall normalization
- “alpha”: the first power law slope
- “lum\_ref”: the break luminosity/magnitude where the power law slopes change.

**calc\_ionizing\_emissivity\_at\_1450A**(*redsh*, *lum\_range*, *\*\*kwargs*)

Calculate the ionizing emissivity at rest-frame 1450A,  $\epsilon_{1450}$ , in units of  $\text{erg s}^{-1} \text{Hz}^{-1} \text{Mpc}^{-3}$ .

This function integrates the luminosity function at redshift “redsh” over the luminosity interval “lum\_range” to calculate the ionizing emissivity at rest-frame 1450A.

Calling this function is only valid if the luminosity function “lum\_type” argument is “lum\_type”=“M1450”.

**Parameters**

- **redsh** (*float* or *numpy.ndarray*) – Redshift for evaluation
- **lum\_range** (*tuple*) – Luminosity range

**Returns**

Ionizing emissivity ( $\text{erg s}^{-1} \text{Hz}^{-1} \text{Mpc}^{-3}$ )

**Return type**

float

**evaluate**(*lum*, *redsh*, *parameters=None*)

Evaluate the single power law as a function of magnitude (“lum”) and redshift (“redsh”).

Function to be evaluated: atelier.lumfun.mag\_single\_power\_law()

**Parameters**

- **lum** (*float* or *numpy.ndarray*) – Luminosity for evaluation
- **redsh** (*float* or *numpy.ndarray*) – Redshift for evaluation
- **parameters** (*dict*(*atelier.lumfun.Parameters*)) – Dictionary of parameters used for this specific calculation. This does not replace the parameters with which the luminosity function was initialized. (default=None)

**Returns**

Luminosity function value

**Return type**

(*numpy.ndarray*, *numpy.ndarray*)

**class atelier.lumfun.SmoothDoublePowerLawLF**(*parameters*, *param\_functions*, *lum\_type=None*, *cosmology=None*, *ref\_cosmology=None*, *ref\_redsh=None*, *verbose=1*)

Luminosity function, which takes the functional form of a double power law with the luminosity in absolute magnitudes.

The luminosity function has four main parameters:

- “phi\_star”: the overall normalization
- “lum\_star”: the break luminosity/magnitude where the power law slopes change.
- “alpha”: the first power law slope
- “beta”: the second power law slope
- “delta”: the smoothing parameter

**calc\_ionizing\_emissivity\_at\_1450A**(*redsh*, *lum\_range*, *\*\*kwargs*)

Calculate the ionizing emissivity at rest-frame 1450A,  $\epsilon_{1450}$ , in units of  $\text{erg s}^{-1} \text{Hz}^{-1} \text{Mpc}^{-3}$ .

This function integrates the luminosity function at redshift “redsh” over the luminosity interval “lum\_range” to calculate the ionizing emissivity at rest-frame 1450A.

Calling this function is only valid if the luminosity function “lum\_type” argument is “lum\_type”=“M1450”.

#### Parameters

- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **lum\_range** (*tuple*) – Luminosity range

#### Returns

Ionizing emissivity ( $\text{erg s}^{-1} \text{Hz}^{-1} \text{Mpc}^{-3}$ )

#### Return type

float

**evaluate**(*lum, redsh, parameters=None*)

Evaluate the double power law as a function of magnitude (“lum”) and redshift (“redsh”).

Function to be evaluated: `atelier.lumfun.mag_double_power_law()`

#### Parameters

- **lum** (*float or numpy.ndarray*) – Luminosity for evaluation
- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **parameters** (*dict(atelier.lumfun.Parameters)*) – Dictionary of parameters used for this specific calculation. This does not replace the parameters with which the luminosity function was initialized. (default=None)

#### Returns

Luminosity function value

#### Return type

(*numpy.ndarray, numpy.ndarray*)

**class** `atelier.lumfun.WangFeige2019DPLQLF`(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Wang+2019 at  $z \sim 6.7$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2019ApJ...884...30W/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

This implementation adopts the double power law fit described in Section 5.5

**static** **phi\_star**(*redsh, phi\_star\_z6p7, k, z\_ref*)

Calculate the redshift dependent luminosity function normalization.

#### Parameters

- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **phi\_star\_z6p7** (*float*) – Logarithmic source density at  $z=6$
- **k** (*float*) – Power law exponent of density evolution
- **z\_ref** (*float*) – Reference redshift

#### Returns

**class** `atelier.lumfun.WangFeige2019SPLQLF`(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Wang+2019 at  $z \sim 6.7$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2019ApJ...884...30W/abstract>

The luminosity function is parameterized as a single power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

This implementation adopts the single power law fit described in Section 5.5

**class** atelier.lumfun.**Willott2010QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Willott+2010 at  $z \sim 6$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2010AJ...139..906W/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

This implementation adopts the double power law fit presented in Section 5.2.

**static phi\_star**(*redsh, phi\_star\_z6, k, z\_ref*)

Calculate the redshift dependent luminosity function normalization.

#### Parameters

- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **phi\_star\_z6** (*float*) – Logarithmic source density at  $z=6$
- **k** (*float*) – Power law exponent of density evolution
- **z\_ref** (*float*) – Reference redshift

#### Returns

**class** atelier.lumfun.**YangJinyi2016QLF**(*cosmology=None*)

Implementation of the type-I quasar UV(M1450) luminosity function of Yang+2016 at  $z \sim 5$ .

ADS reference: <https://ui.adsabs.harvard.edu/abs/2016ApJ...829...33Y/abstract>

The luminosity function is parameterized as a double power law with the luminosity variable in absolute magnitudes at 1450Å, M1450.

The maximum likelihood fit in Yang+2016 includes the quasar samples of McGreer+2013 at fainter magnitudes (i.e., the SDSS DR7 and Stripe 82 samples).

This implementation adopts the double power law fit presented in Section 5.2.

**static phi\_star**(*redsh, log\_phi\_star\_z6, k, z\_ref*)

Calculate the redshift dependent luminosity function normalization.

#### Parameters

- **redsh** (*float or numpy.ndarray*) – Redshift for evaluation
- **phi\_star\_z6** (*float*) – Logarithmic source density at  $z=6$
- **k** (*float*) – Power law exponent of density evolution
- **z\_ref** (*float*) – Reference redshift

#### Returns

atelier.lumfun.**interp\_dVdzdO**(*redsh\_range, cosmo*)

Interpolate the differential comoving solid volume element  $(dV/dz)d\Omega$  over the specified redshift range  $z_{\text{range}} = (z_1, z_2)$ .

This interpolation speeds up volume (redshift, solid angle) integrations for the luminosity function without significant loss in accuracy.

The resolution of the redshift array, which will be interpolated is  $\Delta z = 0.025$ .

#### Parameters

- **redsh\_range** (*tuple*) – Redshift range for interpolation
- **cosmo** (*astropy.cosmology.Cosmology*) – Cosmology

### Returns

1D interpolation function

`atelier.lumfun.lum_double_power_law(lum, phi_star, lum_star, alpha, beta)`

Evaluate a broken double power law luminosity function as a function of luminosity.

### Parameters

- **lum** (*float or np.ndarray*) – Luminosity
- **phi\_star** (*float*) – Normalization of the broken power law at a value of lum\_star
- **lum\_star** (*float*) – Break luminosity of the power law
- **alpha** (*float*) – First slope of the broken power law
- **beta** (*float*) – Second slope of the broken power law

### Returns

Value of the broken double power law at a magnitude of M

### Return type

float or np.ndarray

`atelier.lumfun.mag_double_power_law(mag, phi_star, mag_star, alpha, beta)`

Evaluate a broken double power law luminosity function as a function of magnitude.

### Parameters

- **mag** (*float or np.ndarray*) – Magnitude
- **phi\_star** (*float*) – Normalization of the broken power law at a value of mag\_star
- **mag\_star** (*float*) – Break magnitude of the power law
- **alpha** (*float*) – First slope of the broken power law
- **beta** (*float*) – Second slope of the broken power law

### Returns

Value of the broken double power law at a magnitude of M

### Return type

float or np.ndarray

`atelier.lumfun.mag_schechter_function(mag, phi_star, mag_star, alpha)`

Evaluate a Schechter luminosity function as a function of magnitude.

### Parameters

- **mag** –
- **phi\_star** –
- **mag\_star** –
- **alpha** –

### Returns

`atelier.lumfun.mag_single_power_law(mag, phi_star, mag_ref, alpha)`

Evaluate a power law luminosity function as a function as a function of magnitude

### Parameters

- **mag** (*float or np.ndarray*) – Magnitude

- **phi\_star** (*float*) – Normalization of the power law at a value of mag\_ref
- **mag\_ref** (*float*) – Reference magnitude of the power law
- **alpha** (*float*) – Slope of the power law

**Returns**

Value of the broken double power law at a magnitude of M

**Return type**

float or np.ndarray

`atelier.lumfun.mag_smooth_double_power_law(mag, phi_star, mag_star, alpha, beta, log_delta)`

Evaluate a smooth broken double power law luminosity function as a function of magnitude.

**param mag**

Magnitude

**type mag**

float or np.ndarray

**param phi\_star**

Normalization of the broken power law at a value of mag\_star

**type phi\_star**

float

**param mag\_star**

Break magnitude of the power law

**type mag\_star**

float

**param alpha**

First slope of the broken power law

**type alpha**

float

**param beta**

Second slope of the broken power law

**type beta**

float

**param delta**

Smoothness parameter

**type delta**

float

**return**

Value of the broken double power law at a magnitude of M

**rtype**

float or np.ndarray

`atelier.lumfun.richards_single_power_law(mag, phi_star, mag_ref, alpha)`

**Parameters**

- **mag** –
- **phi\_star** –

- `mag_ref` –
- `alpha` –

**Returns**





## THE SELECTION FUNCTION MODULE

**class** atelier.selfun.ClippedFunction(*fun, minval, maxval*)

A helper class to clip output between two values.

**class** atelier.selfun.CompositeQsoSelectionFunction(*selfun\_list, mag\_range=None, redsh\_range=None*)

Composite quasar selection function class

This class provides the capabilities for composite quasar selection functions and updates the base class evaluate and call methods accordingly.

**selfun\_list**

List of quasar selection functions

**Type**

list(*atelier.selfun.QsoSelectionFunction*)

**mag\_range**

Magnitude range over which the quasar selection function is evaluated.

**Type**

(tuple)

**redsh\_range**

Redshift range over which the quasar selection function is evaluated.

**Type**

(tuple)

**evaluate**(*mag, redsh*)

Evaluate the composite selection function at magnitude and redshift.

**Parameters**

- **mag** (*float* or *np.ndarray*) – Magnitude at which the quasar selection function is evaluated.
- **redsh** (*float* or *np.ndarray*) – Redshift at which the quasar selection function is evaluated.

**Returns**

Value of selection function at given redshift and magnitude.

**Return type**

float or np.ndarray

**class atelier.selfun.QsoSelectionFunction**(*mag\_range=None, redsh\_range=None*)

Quasar selection function class.

This class provides the basic capabilities for quasar selection functions.

**mag\_range**

Magnitude range over which the quasar selection function is evaluated.

**Type**

(tuple)

**redsh\_range**

Redshift range over which the quasar selection function is evaluated.

**Type**

(tuple)

**evaluate**(*mag, redsh*)

Evaluate selection function at magnitude and redshift.

**Parameters**

- **mag** (*float or np.ndarray*) – Magnitude at which the quasar selection function is evaluated.
- **redsh** (*float or np.ndarray*) – Redshift at which the quasar selection function is evaluated.

**Returns**

Value of selection function at given redshift and magnitude.

**Return type**

float or np.ndarray

**plot\_selfun**(*mag\_res=0.01, redsh\_res=0.01, mag\_range=None, redsh\_range=None, title=None, levels=[0.2, 0.5, 0.7], level\_color='k', cmap=<matplotlib.colors.ListedColormap object>, vmin=0, vmax=1, sample\_mag=None, sample\_z=None, sample\_color='red', sample\_mec='k', sample\_marker='D', save\_name=None*)

Plot the selection function on a grid of redshift and magnitude.

To calculate the map of the selection function the selection function is evaluated using the .evaluate method.

If no magnitude and redshift range is provided the class attributes are used to determine the ranges.

**Parameters**

- **mag\_res** (*float*) – Magnitude resolution (default = 0.01)
- **redsh\_res** (*float*) – Redshift resolution (default = 0.01)
- **mag\_range** (*tuple*) – Magnitude range over which the quasar selection function is evaluated. (default = None)
- **redsh\_range** (*tuple*) – Redshift range over which the quasar selection function is evaluated. (default = None)
- **title** (*string*) – Title of the plot (default = None)
- **levels** (*list(float)*) – Values of contour levels to plot
- **level\_color** (*string*) – Color for the level contours
- **cmap** (*Matplotlib colormap*) – Color map for the selection function
- **save\_name** (*string*) – Populate with name to save the plot. Default='None'

### Returns

**class** atelier.selfun.QsoSelectionFunctionConst(*value*)

Composite quasar selection function class

This class provides the capabilities for the constant quasar selection function and updates the base class evaluate and call methods accordingly.

### value

Value of the constant quasar selection function

### Type

float

**evaluate**(*mag*, *redsh*)

Evaluate the constant selection function at magnitude and redshift.

### Parameters

**redsh** (*float* or *np.ndarray*) – Redshift at which the quasar selection function is evaluated.

### Returns

Value of selection function at given redshift and magnitude.

### Returns

Value of selection function at given redshift and magnitude.

### Return type

float or np.ndarray

**class** atelier.selfun.QsoSelectionFunctionGrid(*mag\_range=None*, *redsh\_range=None*,  
*mag\_bins=None*, *redsh\_bins=None*, *selfungrid=None*,  
*clip=True*, *filename=None*, *format='csv'*)

Gridded quasar selection function class.

This class provides the capabilities for a quasar selection function calculated on a magnitude redshift grid of sources defined by a query of the source properties. It updates the base class evaluate and call methods accordingly.

### mag\_range

Magnitude range over which the quasar selection function is evaluated.

### Type

(tuple)

### redsh\_range

Redshift range over which the quasar selection function is evaluated.

### Type

(tuple)

### mag\_bins

Number of bins in magnitude

### Type

(int)

### redsh\_bins

Number of bins in redshift

### Type

(int)

### **selffungrid**

Grid with selection function values

#### **Type**

(np.ndarray)

### **clip**

Boolean to indicate whether to clip the output value of the selection function between 0 and 1.

#### **Type**

(bool)

**calc\_selffungrid\_from\_df**(*df*, *mag\_col\_name*, *redshift\_col\_name*, *query=None*, *sel\_idx=None*, *n\_per\_bin=None*, *verbose=1*)

Calculate the selection function grid from a pandas DataFrame

#### **Parameters**

- **df** (*pandas.DataFrame*) – Data frame to calculate the selection function from
- **query** (*string*) – A query describing the subsample that is compared to the full data set in order to calculate the selection function.
- **mag\_col\_name** (*string*) – Name of the magnitude column in the data frame.
- **redshift\_col\_name** (*string*) – Name of the redshift column in the data frame.
- **n\_per\_bin** (*int*) – Number of sources per bin (default = None). This keyword argument is used to check whether the redshift and magnitude ranges and the number of bins per dimension produce a uniform distribution of sources per bin.
- **verbose** (*int*) – Verbosity

#### **Returns**

None

**evaluate**(*mag*, *redsh*)

Evaluate the interpolation of the selection function grid at magnitude and redshift.

#### **Parameters**

**redsh** (*float or np.ndarray*) – Redshift at which the quasar selection function is evaluated.

#### **Returns**

Value of selection function at given redshift and magnitude.

#### **Returns**

Value of selection function at given redshift and magnitude.

#### **Return type**

float or np.ndarray

**get\_selffun\_from\_grid**()

Calculate an interpolation (RectBivariateSpline) of the selection function grid to allow evaluations of the selection function at any point within the magnitude and redshift range.

This methods need to be called before the selection function can be evaluated.

#### **Returns**

None

**load**(*filename*, *format*='csv')

Load a QsoSelectionFunctionGrid from a file.

**Parameters**

- **filename** (*string*) – Filename to load a saved QsoSelectionFunctionGrid from.
- **format** – File format of saved QsoSelectionFunctionGrid.

Currently implemented formats are: “.csv” :type format: string :return:

**plot\_grid**(*title*=None)

Plot the selection function grid as a map of magnitude and redshift.

**Parameters**

**title** (*string*) – Title of the plot

**Returns**

None

**save**(*filename*, *format*='csv')

Save a QsoSelectionFunctionGrid object to a file

**Parameters**

- **filename** (*string*) – Filename to save the QsoSelectionFunctionGrid to.
- **format** – File format of saved QsoSelectionFunctionGrid.

Currently implemented formats are: “.csv” :type format: string :return:



## THE SURVEY MODULE

```
class atelier.survey.Survey(obj_df, lum_colname, redsh_colname, sky_area, selection_function=None,
                             lum_range=None, redsh_range=None, conf_interval='poisson')
```

Survey class is a container to hold information on the sources of a particular astronomical survey.

It is used primarily to forward observational data in the calculation of luminosity function fits.

**obj\_df**

Data frame with information on astronomical sources in the survey.

**Type**

pandas.DataFrame

**lum\_colname**

The name of the data frame column holding the luminosity (absolute magnitude) information of the sources in the survey.

**Type**

string

**redsh\_colname**

The name of the data frame column holding the redshift information of the sources in the survey.

**Type**

string

**sky\_area**

Sky area the survey covers in square degrees

**Type**

float

**selection\_function**

Selection function of the survey (default = None).

**Type**

lumfun.selfun.QsoSelectionFunction

```
calc_binned_lumfun_PC2000(lum_edges, redsh_edges, cosmology, kcorrection=None,
                             app_mag_lim=None, **kwargs)
```

Calculation of the binned luminosity function based on the method outlined in Page & Carrera 2000

ADS Link: <https://ui.adsabs.harvard.edu/abs/2000MNRAS.311..433P/abstract>

This function is very similar to the other method for calculating the binned luminosity function below.

**Parameters**

- `lum_edges` –
- `redsh_edges` –
- `cosmology` –
- `kcorrection` –
- `app_mag_lim` –

#### Returns

`atelier.survey.return_poisson_confidence(n, bound_low=0.15865, bound_upp=0.84135)`

Return the Poisson confidence interval boundaries for the lower and upper bound given a number of events `n`.

The default values for ther lower and upper bounds are equivalent to the 1 sigma confidence interval of a normal distribution.

#### Parameters

- `n` –
- `n_sigma` –

#### Returns



---

## THE LUMINOSITY FUNCTION FIT MODULE

```
class atelier.lumfunfit.LuminosityFunctionFit(lum_range, redsh_range, cosmology, surveys,  
                                              emcee_nwalkers=50, emcee_steps=1000)
```

Class that carries out maximum likelihood fits to a luminosity function model using MCMC (emcee).

**lum\_range**

Luminosity range for the luminosity function fit (2-element list).

**Type**

list(float, float)

**redsh\_range**

Redshift range for the luminosity function fit (2-element list).

**Type**

list(float, float)

**cosmology**

Cosmology object

**Type**

astropy.cosmology.Cosmology

**surveys**

List of survey objects used in to fit the luminosity function model to.

**Type**

list(*survey.Survey*)

**emcee\_sampler**

Emcee Ensemble sampler object. This attribute cannot be initialized but is internally populated during the .run\_mcmc or .run\_mcmc\_multiprocess methods.

**Type**

emcee.EnsembleSampler

**emcee\_nwalker**

Number of walkers for the emcee sampler.

**Type**

int (default = 50)

**emcee\_steps**

Number of steps for the emcee sampler.

**Type**

int (default = 1000)

**log\_prior**(*theta*, *parameters*)

Logarithmic prior function for luminosity function maximum likelihood estimation (MLE).

**Parameters**

- **theta** (*list of parameter values*) – parameter vector
- **parameters** (*dict(atelier.lumfun.Parameter)*) – Dictionary of parameters used in MLE. The Parameter.bounds attribute is used to keep the fit within the pre-defined parameter boundaries (bounds).

**Returns**

Logarithmic prior

**Return type**

float

**log\_probability**(*theta*, *lumfun=None*, *use\_prior=True*, *int\_mode='romberg'*)

Logarithmic probability for luminosity function maximum likelihood estimation (MLE).

The logarithmic probability implemented here goes back to the definition of Marshall+1983 Equation 3.

ADS reference: <https://ui.adsabs.harvard.edu/abs/1983ApJ...269...35M/abstract>

**Parameters**

- **theta** (*list of parameter values*) – parameter vector
- **lumfun** (*atelier.lumfun.LuminosityFunction (or children classes)*) – Luminosity function model

**Returns**

Logarithmic probability

**Return type**

float

**negative\_log\_likelihood**(*theta*, *lumfun*)

Return the negative logarithmic likelihood for a set of parameters ‘theta’ and a luminosity function model ‘lumfun’ using the default logarithmic likelihood definition.

**Parameters**

- **theta** (*list of parameter values*) – parameter vector
- **lumfun** (*atelier.lumfun.LuminosityFunction (or children classes)*) – Luminosity function model

**Returns**

Negative logarithmic likelihood

**Return type**

float

**run\_mcmc**(*lumfun*, *initial\_guess=None*, *nwalkers=None*, *steps=None*, *int\_mode='romberg'*)

Run the emcee MCMC EnsembleSampler to fit the luminosity function model to the survey data using maximum likelihood estimation.

**Parameters**

- **lumfun** (*atelier.lumfun.LuminosityFunction (or children classes)*) – Luminosity function model
- **initial\_guess** (*list(float) (default = None)*) – Initial guess of the free parameters (ordered as in the luminosity function model).

- **nwalkers** (*int* (*default* = *None*)) – Number of walkers for the emcee sampler overriding ( not overwriting) the class attribute `emcee_nwalkers`.
- **steps** (*int* (*default* = *None*)) – Number of steps for the emcee sampler overriding ( not overwriting) the class attribute `emcee_nwalkers`.

#### Returns

None

**run\_mcmc\_multiprocess**(*lumfun*, *initial\_guess*=*None*, *nwalkers*=*None*, *steps*=*None*, *processes*=*None*, *int\_mode*='romberg')

Run the emcee MCMC EnsembleSampler to fit the luminosity function model to the survey data using maximum likelihood estimation in multiprocessing mode.

#### Parameters

- **lumfun** (`atelier.lumfun.LuminosityFunction` (or *children classes*)) – Luminosity function model
- **initial\_guess** (*list(float)* (*default* = *None*)) – Initial guess of the free parameters (ordered as in the luminosity function model).
- **nwalkers** (*int* (*default* = *None*)) – Number of walkers for the emcee sampler overriding ( not overwriting) the class attribute `emcee_nwalkers`.
- **steps** (*int* (*default* = *None*)) – Number of steps for the emcee sampler overriding ( not overwriting) the class attribute `emcee_nwalkers`.
- **processes** (*int* (*default* = *None*)) – Number of processes for multiprocessing. The default value of 'None' will use the maximum number determined by the multiprocessing package.

#### Returns

`atelier.lumfunfit.log_prior(theta, parameters)`

**Logarithmic prior function for luminosity function maximum likelihood estimation (MLE) using emcee in multiprocessing mode.**

#### Parameters

- **theta** (*list of parameter values*) – parameter vector
- **parameters** (*dict* (`atelier.lumfun.Parameter`)) – Dictionary of parameters used in MLE. The `Parameter.bounds` attribute is used to keep the fit within the pre-defined parameter boundaries (bounds).

#### Returns

Logarithmic prior

#### Return type

float

**atelier.lumfunfit.log\_probability**(*theta*, *lumfun*=*None*, *lum\_range*=*None*, *redsh\_range*=*None*, *surveys*=*None*, *dVdzdO*=*None*, *log\_prior*=*None*, *use\_prior*=*True*, *int\_mode*='romberg')

Logarithmic probability for luminosity function maximum likelihood estimation (MLE) using emcee in multiprocessing mode.

The logarithmic probability implemented here goes back to the definition of Marshall+1983 Equation 3.

ADS reference: <https://ui.adsabs.harvard.edu/abs/1983ApJ...269...35M/abstract>

### Parameters

- **theta** (*list of parameter values*) – parameter vector
- **lumfun** (`atelier.lumfun.LuminosityFunction` (or *children classes*)) – Luminosity function model
- **lum\_range** (*tuple*) – Luminosity range
- **redsh\_range** (*tuple*) – Redshift range
- **surveys** (*list*(`atelier.survey.Survey`)) – List of surveys
- **dVdzd0** (*function*) – Differential comoving solid volume element
- **log\_prior** – Function describing the logarithmic prior
- **use\_prior** (*bool*) – Boolean to indicate whether the logarithmic prior function will be used or strictly flat priors for all parameters will be adopted.

### Returns

Logarithmic probability

### Return type

float

## THE K-CORRECTION MODULE

**class** atelier.kcorrection.**KCorrection**(*cosmology*)

Base class for an astronomical K-correction.

**cosmology**

Cosmology

**Type**

astropy.cosmology.Cosmology

**M2m**(*mag*, *redsh*)

Calculate the apparent (observed) magnitude based on the rest-frame absolute magnitude and redshift using the K-correction and specified cosmology.

**Parameters**

- **mag** (*float*) – Absolute magnitude
- **redsh** (*float*) – Redshift

**Returns**

Apparent magnitude

**evaluate**(*mag*, *redsh*)

Evaluate the K-correction for an object of given apparent magnitude and redshift.

This method is not implemented and will be overwritten by children methods.

**Parameters**

- **mag** (*float*) – Magnitude
- **redsh** (*float*) – Redshift

**Returns**

None

**m2M**(*mag*, *redsh*)

Calculate the rest-frame absolute magnitude based on the apparent (observed) source magnitude and redshift using the K-correction and specified cosmology.

**Parameters**

- **mag** (*float*) – Magnitude
- **redsh** (*float*) – Redshift

**Returns**

Absolute magnitude

**class atelier.kcorrection.KCorrectionGrid**(*mag\_range, redsh\_range, mag\_bins, redsh\_bins, cosmology, kcorrgrid=None, min\_n\_per\_bin=11*)

K-correction between the observed filter band and the rest-frame filter band determined on a grid of data.

Note that an astronomical K-correction is specific to the filter bands for which it was determined. The current implementation relies on the user's understanding of the K-correction in order to apply it properly.

The K-correction is calculated in redshift and rest-frame magnitude bins on a grid of data for which redshift, rest-frame magnitudes (either from the continuum or through a filter band) and observed frame apparent magnitudes through a filter band are provided. The bins are then interpolated to provide a smooth K-correction as a function of apparent magnitude and redshift.

**mag\_range**

Magnitude range over which the quasar selection function is evaluated.

**Type**  
(tuple)

**redsh\_range**

Redshift range over which the quasar selection function is evaluated.

**Type**  
(tuple)

**mag\_bins**

Number of bins in magnitude

**Type**  
(int)

**redsh\_bins**

Number of bins in redshift

**Type**  
(int)

**cosmology**

Cosmology

**Type**  
astropy.cosmology.Cosmology

**slope**

Slope of the power law (in frequency)

**Type**  
float

**kcorrgrid**

Grid of K-correction values

**Type**  
np.ndarray

**calc\_kcorrection\_from\_df**(*df, redshift\_col\_name, mag\_col\_name, appmag\_col\_name, n\_per\_bin=None, statistic='median', inverse=True, verbose=1*)

Calculate the K-correction grid from a data frame

**Parameters**

- **df** (*pandas.DataFrame*) – Data frame with redshift, absolute reference (filter band) magnitude, and apparent (filter band) magnitude
- **redshift\_col\_name** (*string*) – Name of the redshift column in the data frame.
- **mag\_col\_name** (*string*) – Name of the absolute reference (filter band) magnitude column in the data frame.
- **appmag\_col\_name** (*string*) – Name of the apparent (filter band) magnitude column in the data frame
- **n\_per\_bin** (*int*) – Number of sources per bin (default = None). This keyword argument is used to check whether the redshift and magnitude ranges and the number of bins per dimension produce a uniform distribution of sources per bin.
- **statistic** (*string* (default = 'median')) – A string indicating whether to use the 'median' or 'mean' of the K-correction value in the bin when calculating the grid.
- **inverse** – Boolean to indicate whether inverse kcorrection will

be calculated (default=True). :type inverse: bool :param verbose: Verbosity :type verbose: int :return: None

**evaluate**(*mag, redsh, inverse=False*)

Evaluate the K-correction for an object of given apparent magnitude and redshift.

#### Parameters

- **mag** (*float*) – Magnitude
- **redsh** (*float*) – Redshift

#### Returns

K-correction

#### Return type

float

**get\_inv\_kcorrection\_from\_grid**()

Interpolate the inverse K-correction grid (apparent magnitude, redshift) using a linear interpolation over a rectangular mesh

#### Returns

None

**get\_kcorrection\_from\_grid**()

Interpolate the K-correction grid (absolute magnitude, redshift) using a linear interpolation over a rectangular mesh

#### Returns

None

**plot\_grid**(*title=None, ylabel=None, show\_mad=False*)

Plot the K-correction grid values using matplotlib.

#### Parameters

- **title** (*string*) – Title of the plot
- **ylabel** (*string*) – String for the y-label of the plot.
- **show\_mad** (*bool* (default = False)) – Boolean to indicate whether to show the median absolute deviation of the K-correction instead of the mean/median values.

### Returns

None

**plot\_kcorrection**(*redsh\_arr=None, mag\_arr=None, ylabel=None, save=False, save\_filename=None, mag\_label='\$M\_{1450}={}\$'*)

Plot the K-correction (interpolation) for a range of magnitudes over the given redshifts.

### Parameters

- **redsh\_arr** (*np.ndarray*) – Array of redshifts for which the K-correction is evaluated.
- **mag\_arr** – Array of rest-frame (filter band) magnitudes for which the K-correction is evaluated over the given redshift array.
- **ylabel** (*string*) – String for the y-label of the plot.
- **save** (*bool*) – Boolean to indicate whether to save the plot.
- **save\_filename** (*string*) – Path and name to save the plot.
- **mag\_label** (*string*) – Label for the magnitude

### Returns

None

**class atelier.kcorrection.KCorrectionPL**(*slope, cosmology*)

K-correction for a power law spectrum source with a specified spectral slope per unit frequency.

Note that an astronomical K-correction is specific to the filter bands for which it was determined. The current implementation relies on the user's understanding of the K-correction in order to apply it properly.

Only for the special case of a power law spectrum source (this case) this is independent of the astronomical filter bands.

### cosmology

Cosmology

### Type

`astropy.cosmology.Cosmology`

### slope

Slope of the power law (in frequency)

### Type

`float`

**evaluate**(*mag, redsh, inverse=False*)

Evaluate the K-correction for an object of given apparent magnitude and redshift.

The power law K-correction does not depend on magnitude. Hence the argument 'mag' and the keyword argument 'inverse' play no role.

### Parameters

- **mag** (*float*) – Magnitude
- **redsh** (*float*) – Redshift

### Returns

K-correction

### Return type

`float`



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