

# 1.1\_SSDF

March 8, 2018

## 1 SSDF master catalogue

### 1.1 Preparation of SSDF data

This catalogue comes from `dmu0_SSDF`.

The SSDF data consists in two catalogue of IRAC Ch1 and Ch2 fluxes: one for Ch1 detected sources and the other for Ch2 detected sources. For now, we are only using the Ch1 detected sources. **TODO** : We may find a way to merge the two catalogues and select the best flux for each source.

This notebook was run with `herschelhelp_internal` version:  
04829ed (Thu Nov 2 16:57:19 2017 +0000)

### 1.2 I - Column selection

WARNING: UnitsWarning: 'vega' did not parse as fits unit: At col 0, Unit 'vega' not supported by

Out[6]: <IPython.core.display.HTML object>

### 1.3 II - Removal of duplicated sources

We remove duplicated objects from the input catalogues.

```
/opt/anaconda3/envs/herschelhelp_internal/lib/python3.6/site-packages/astropy/table/column.py:10  
Check the NumPy 1.11 release notes for more information.  
ma.MaskedArray.__setitem__(self, index, value)
```

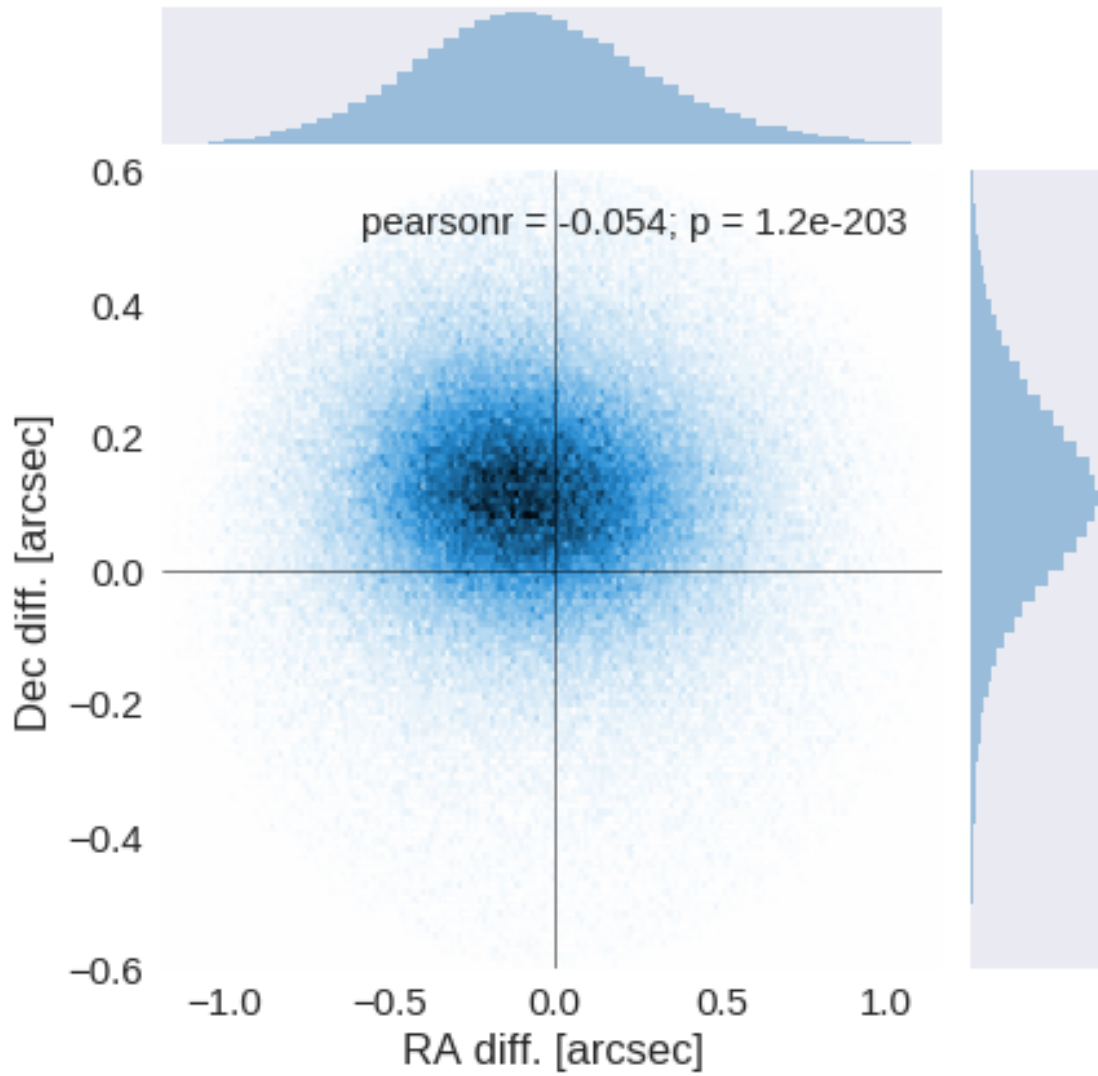
The initial catalogue had 5488166 sources.

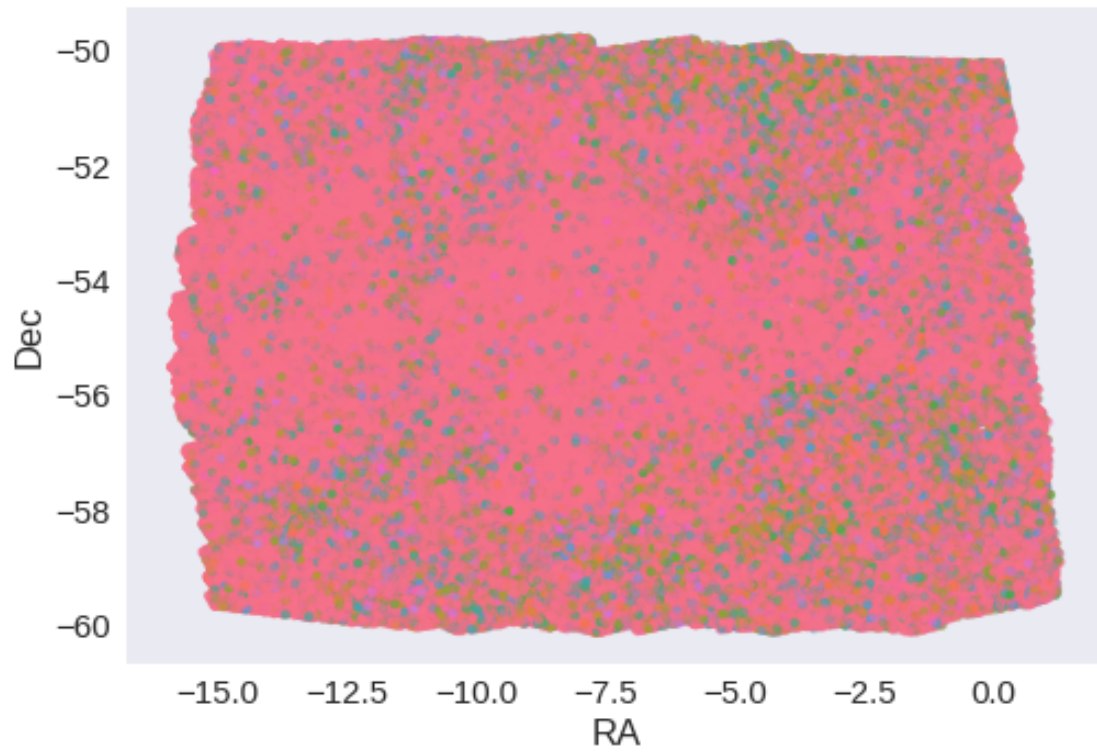
The cleaned catalogue has 5488141 sources (25 removed).

The cleaned catalogue has 25 sources flagged as having been cleaned

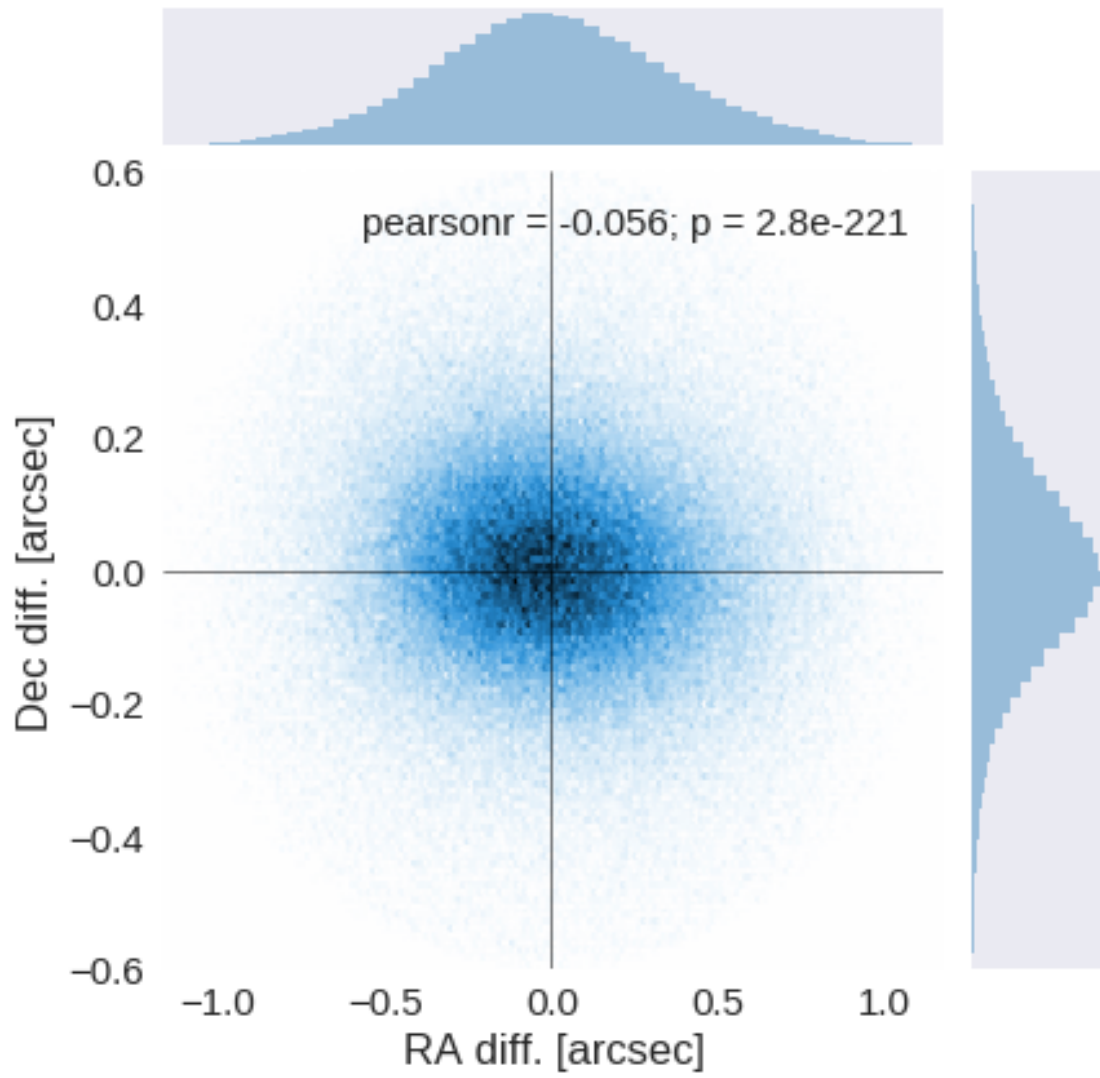
### 1.4 III - Astrometry correction

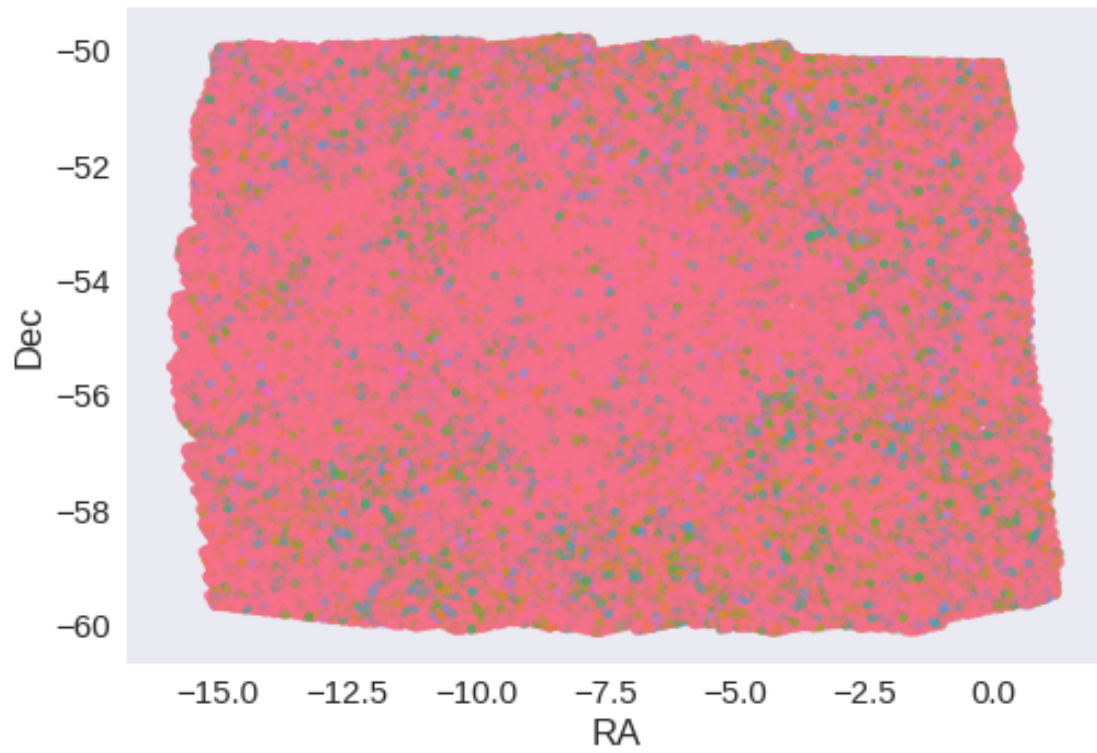
We match the astrometry to the Gaia one. We limit the Gaia catalogue to sources with a g band flux between the 30th and the 70th percentile. Some quick tests show that this give the lower dispersion in the results.





RA correction: 0.08155033374350751 arcsec  
Dec correction: -0.11236725684966586 arcsec





### 1.5 IV - Flagging Gaia objects

352272 sources flagged.

### 1.6 V - Flagging objects near bright stars

## 2 VI - Saving to disk

# 1.2\_VISTA-VHS

March 8, 2018

## 1 SSDF master catalogue

### 1.1 Preparation of VHS data

VISTA telescope/VHS catalogue: the catalogue comes from `dmu0_VHS`.

In the catalogue, we keep:

- The identifier (it's unique in the catalogue);
- The position;
- The stellarity;
- The magnitude for each band.
- The kron magnitude to be used as total magnitude (no "auto" magnitude is provided).

We don't know when the maps have been observed. We will use the year of the reference paper.

- Note: on SSDF, the VHS catalogue does not contain Y data.\*

This notebook was run with `herschelhelp_internal` version:  
0246c5d (Thu Jan 25 17:01:47 2018 +0000) [with local modifications]

### 1.2 I - Column selection

```
/opt/anaconda3/envs/herschelhelp_internal/lib/python3.6/site-packages/astropy/table/column.py:10
```

Check the NumPy 1.11 release notes for more information.

```
ma.MaskedArray.__setitem__(self, index, value)
```

Out[8]: <IPython.core.display.HTML object>

### 1.3 II - Removal of duplicated sources

We remove duplicated objects from the input catalogues.

```
/opt/anaconda3/envs/herschelhelp_internal/lib/python3.6/site-packages/astropy/table/column.py:10
```

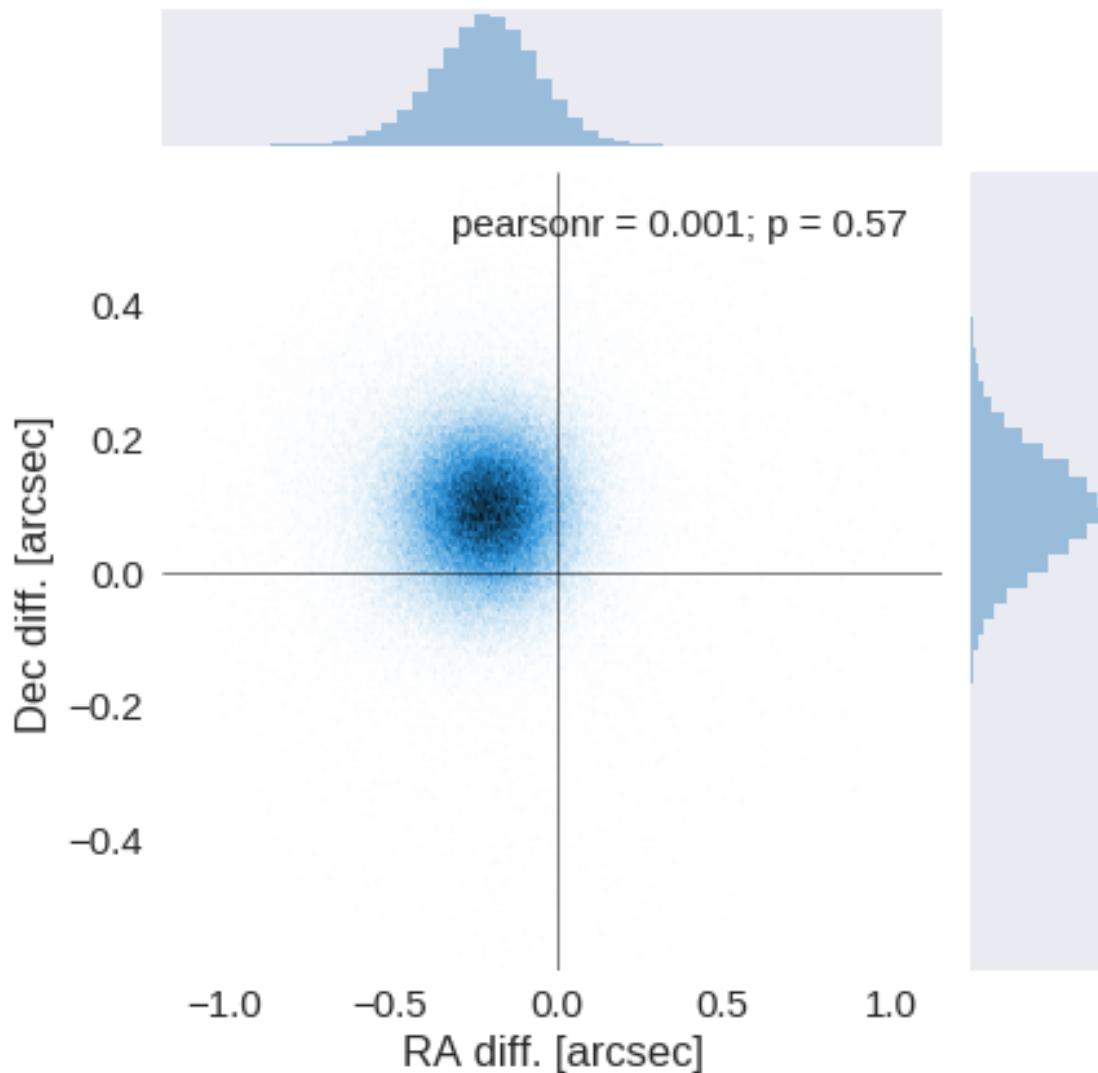
Check the NumPy 1.11 release notes for more information.

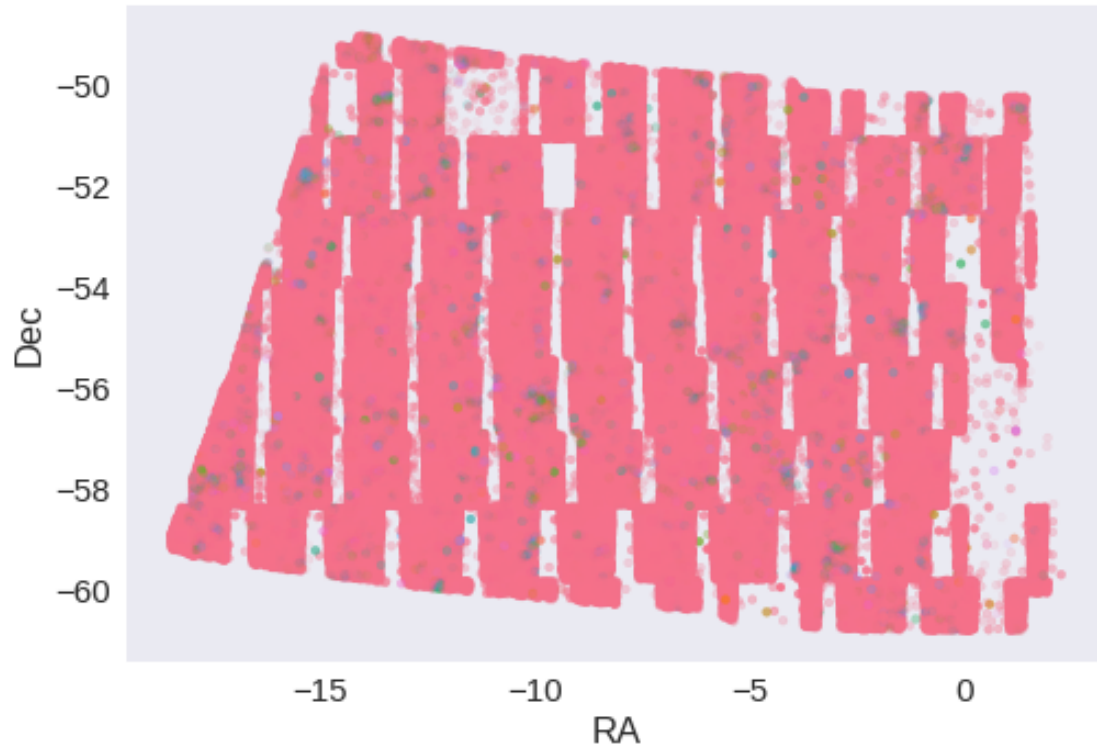
```
ma.MaskedArray.__setitem__(self, index, value)
```

The initial catalogue had 2406318 sources.  
The cleaned catalogue has 2406063 sources (255 removed).  
The cleaned catalogue has 253 sources flagged as having been cleaned

### 1.4 III - Astrometry correction

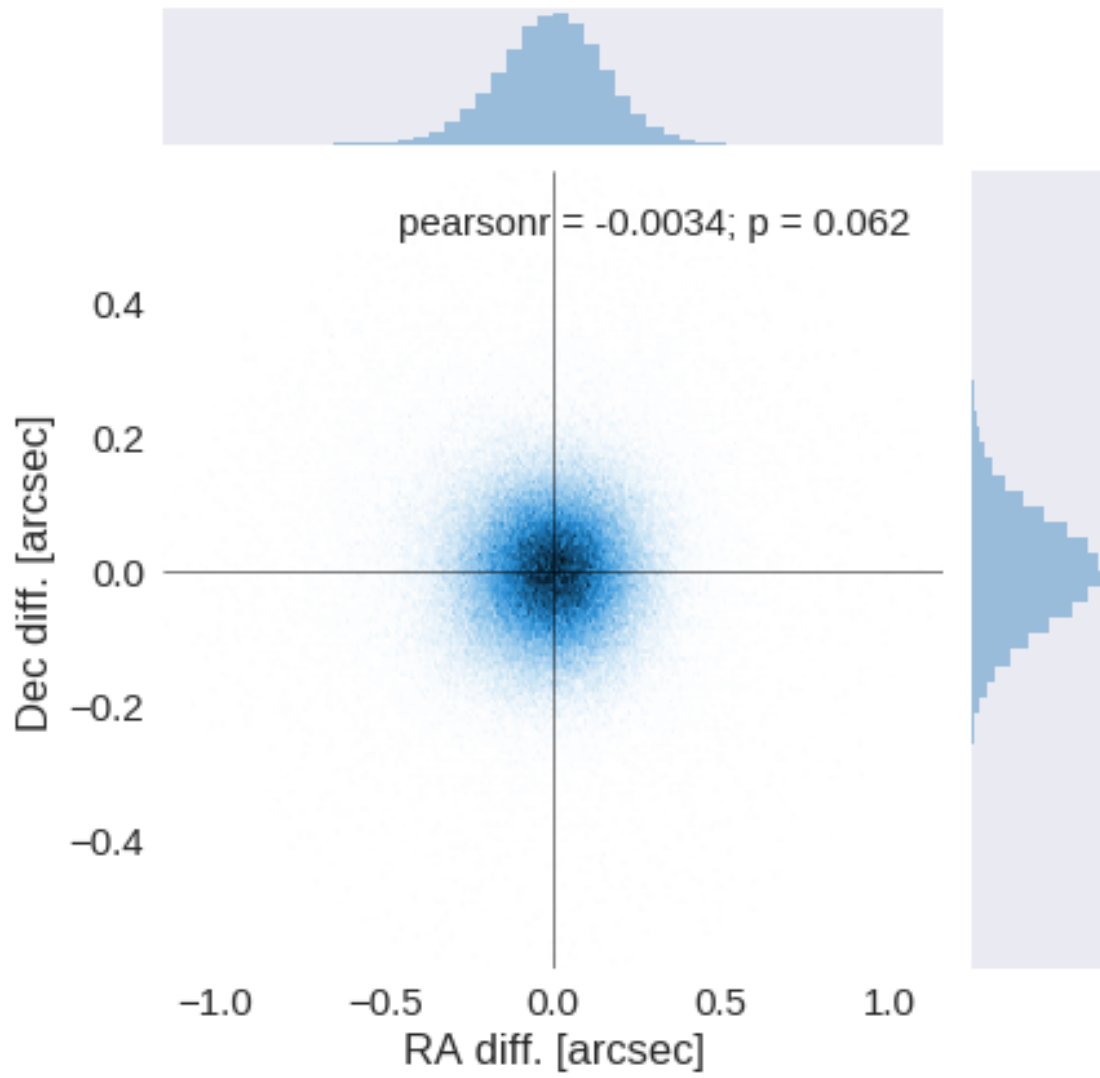
We match the astrometry to the Gaia one. We limit the Gaia catalogue to sources with a g band flux between the 30th and the 70th percentile. Some quick tests show that this give the lower dispersion in the results.

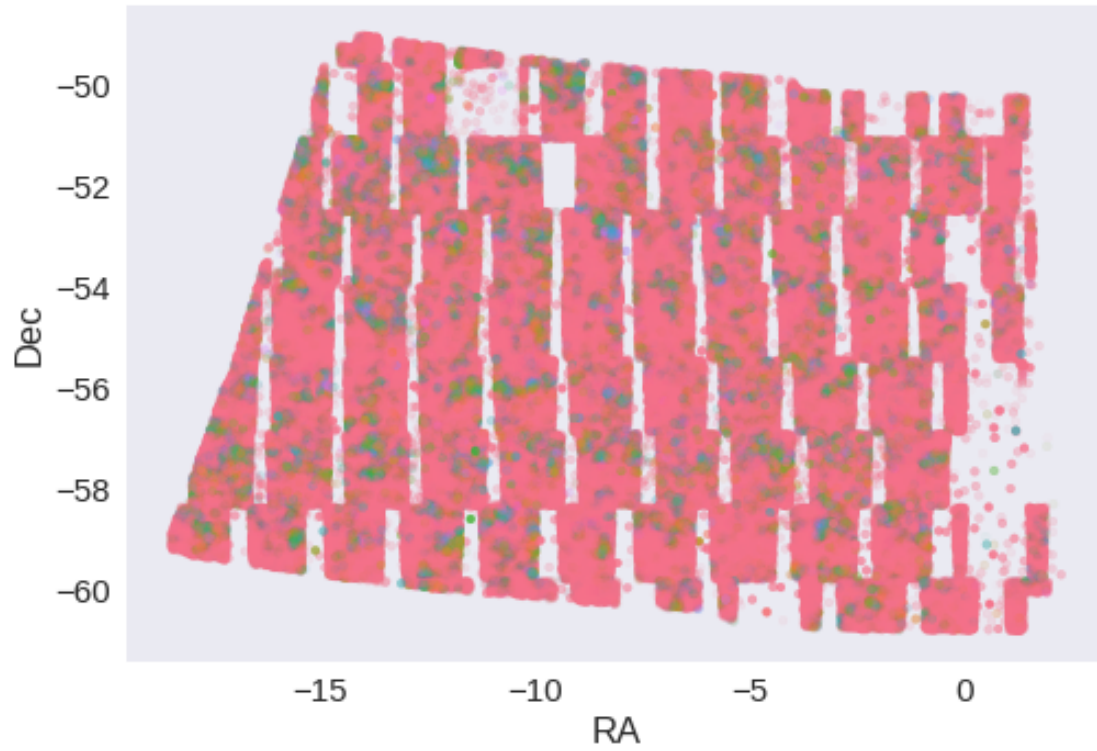




RA correction: 0.21025412717108338 arcsec  
Dec correction: -0.09461352436233028 arcsec







## 1.5 IV - Flagging Gaia objects

301682 sources flagged.

## 1.6 V - Flagging objects near bright stars

## 2 VI - Saving to disk

# 1.3\_DES

March 8, 2018

## 1 SSDF master catalogue

### 1.1 Preparation of DES data

Blanco DES catalogue: the catalogue comes from `dmu0_DES`.

In the catalogue, we keep:

- The identifier (it's unique in the catalogue);
- The position;
- The G band stellarity;
- The magnitude for each band.
- The auto/kron magnitudes/fluxes to be used as total magnitude.
- The aperture magnitudes, which are used to compute a corrected 2 arcsec aperture magnitude.

We don't know when the maps have been observed. We will take the final observation date as 2017.

```
This notebook was run with herschelhelp_internal version:  
0246c5d (Thu Jan 25 17:01:47 2018 +0000) [with local modifications]  
This notebook was executed on:  
2018-02-21 16:18:33.821790
```

### 1.2 1 - Aperture correction

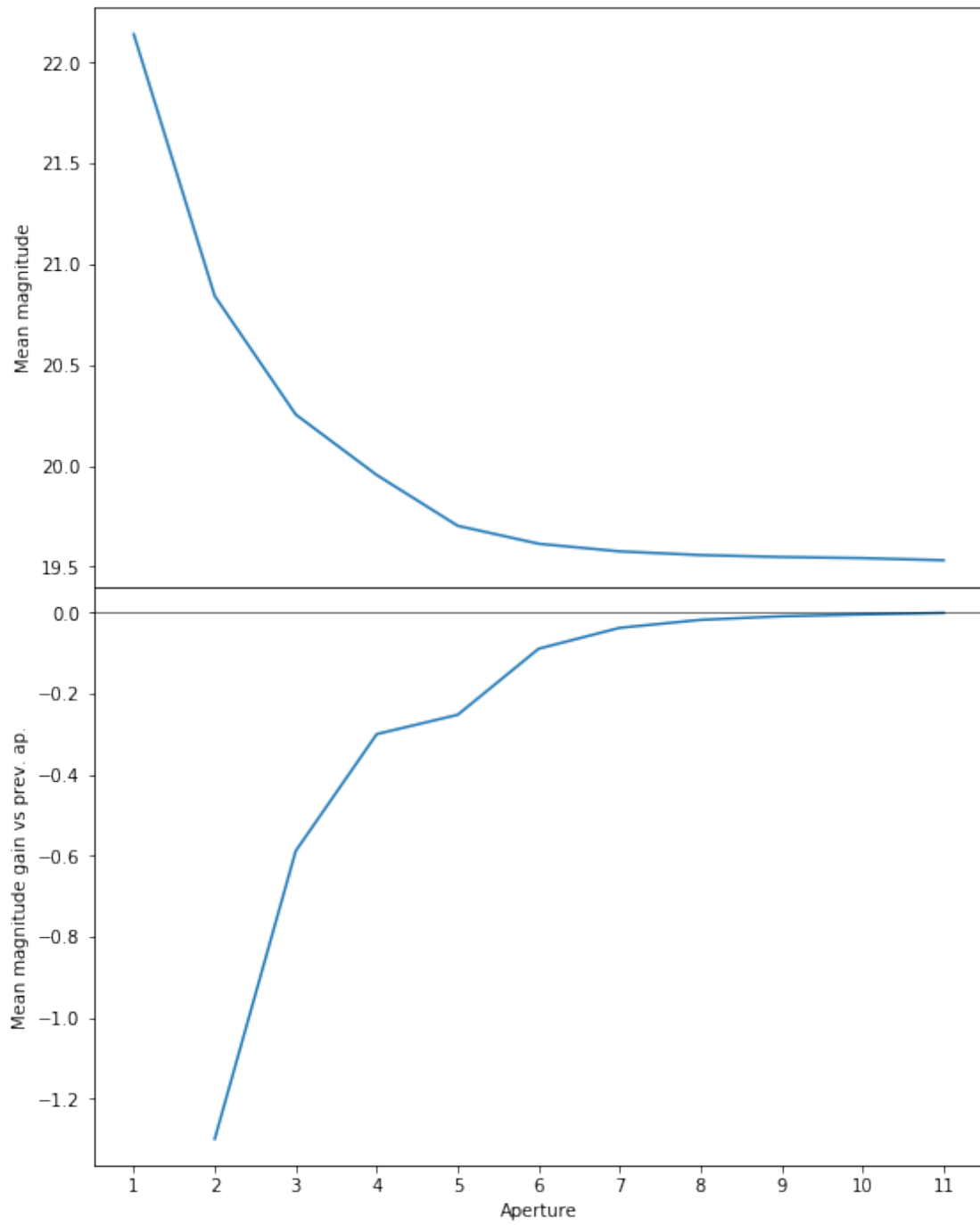
To compute aperture correction we need to determine two parameters: the target aperture and the range of magnitudes for the stars that will be used to compute the correction.

Target aperture: To determine the target aperture, we simulate a curve of growth using the provided apertures and draw two figures:

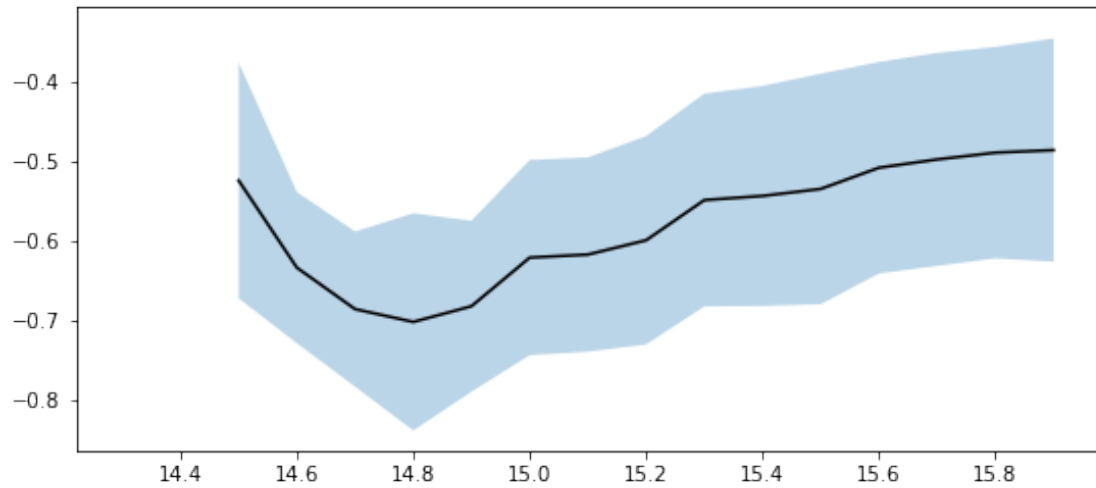
The evolution of the magnitudes of the objects by plotting on the same plot aperture number vs the mean magnitude. The mean gain (loss when negative) of magnitude is each aperture compared to the previous (except for the first of course). As target aperture, we should use the smallest (i.e. less noisy) aperture for which most of the flux is captured.

Magnitude range: To know what limits in aperture to use when doing the aperture correction, we plot for each magnitude bin the correction that is computed and its RMS. We should then use the wide limits (to use more stars) where the correction is stable and with few dispersion.

### 1.2.1 I.a - g band



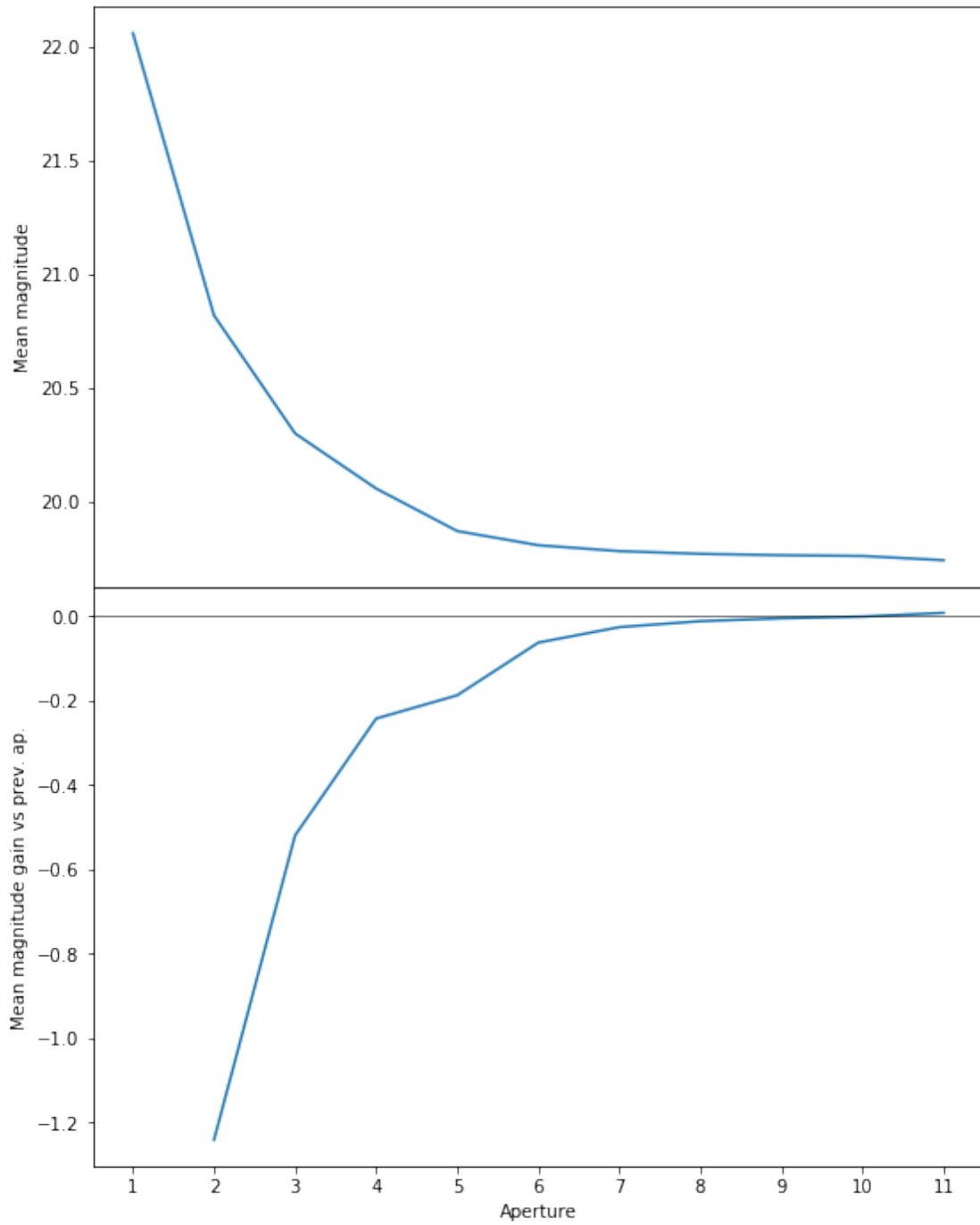
We will use aperture 10 as target.



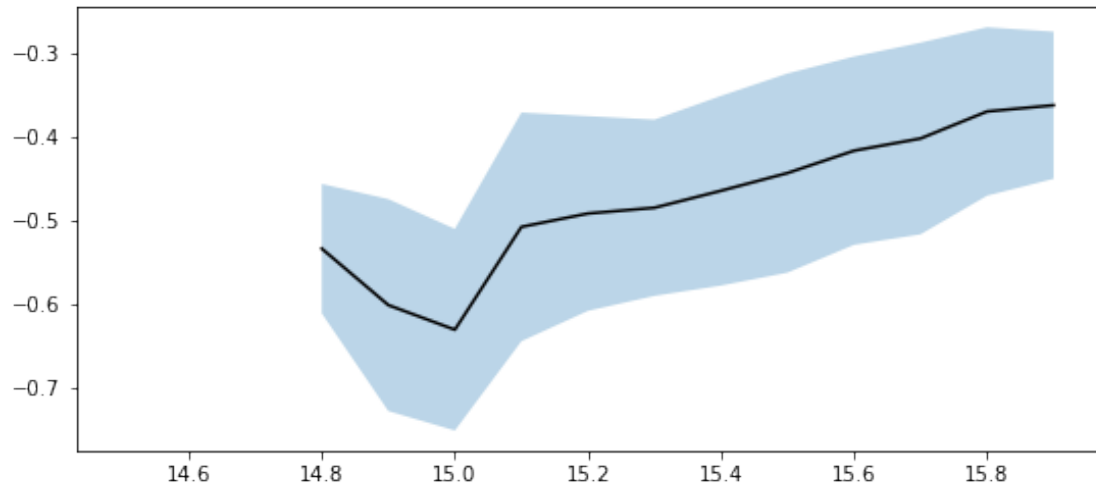
We will use magnitudes between 15.0 and 16.0

Aperture correction for g band:  
Correction: -0.5084390640258789  
Number of source used: 11527  
RMS: 0.13857766793996795

### 1.2.2 I.b - r band



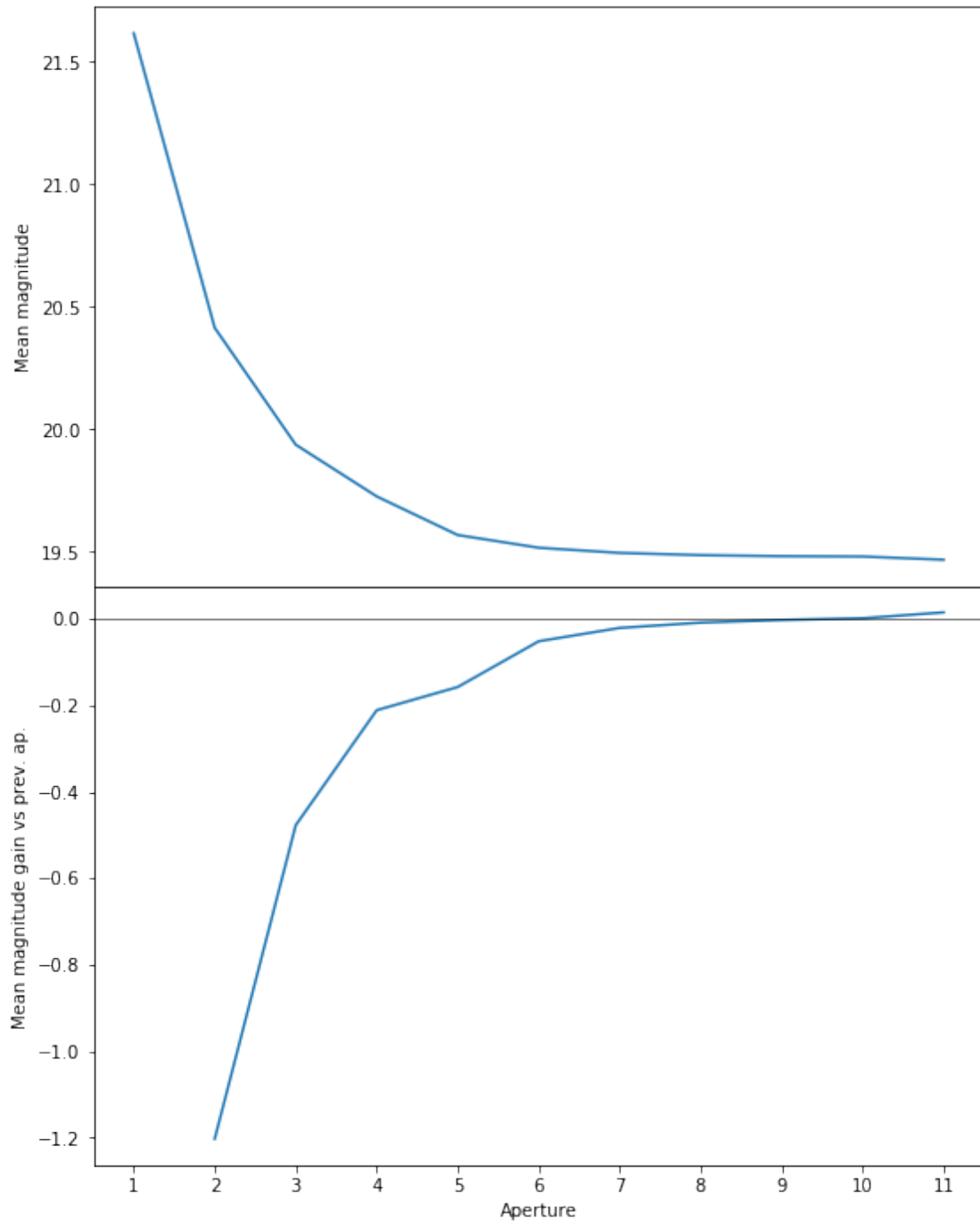
We will use aperture 10 as target.



We use magnitudes between 15.0 and 16.0.

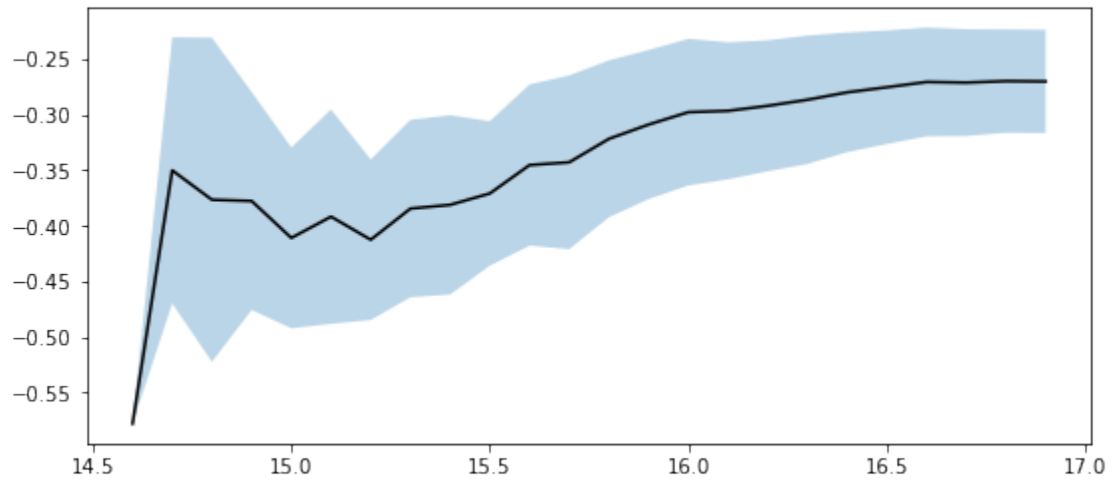
Aperture correction for r band:  
Correction: -0.3870668411254883  
Number of source used: 6262  
RMS: 0.10930435440532538

### 1.2.3 I.b - i band



We will use aperture 10 as target.

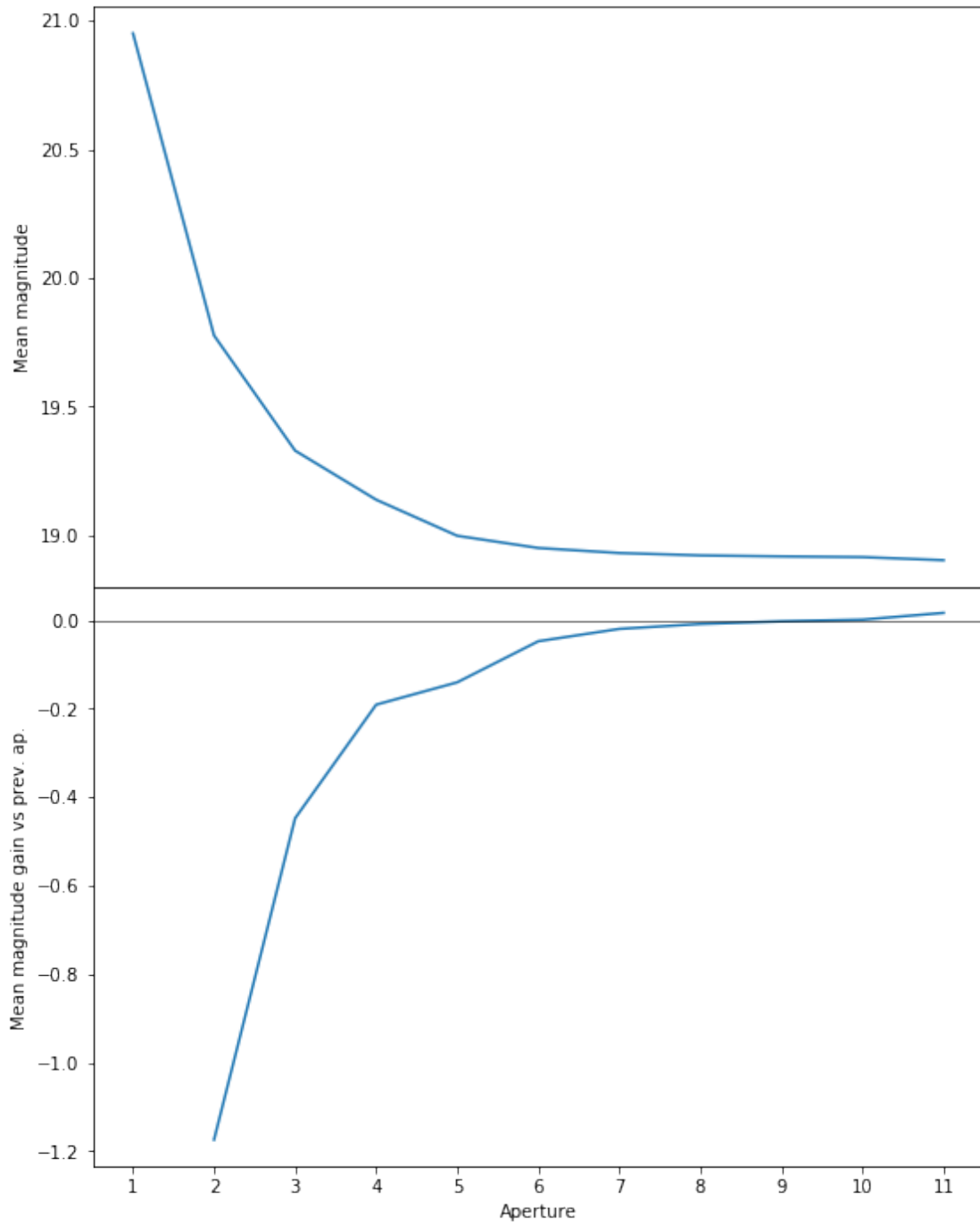




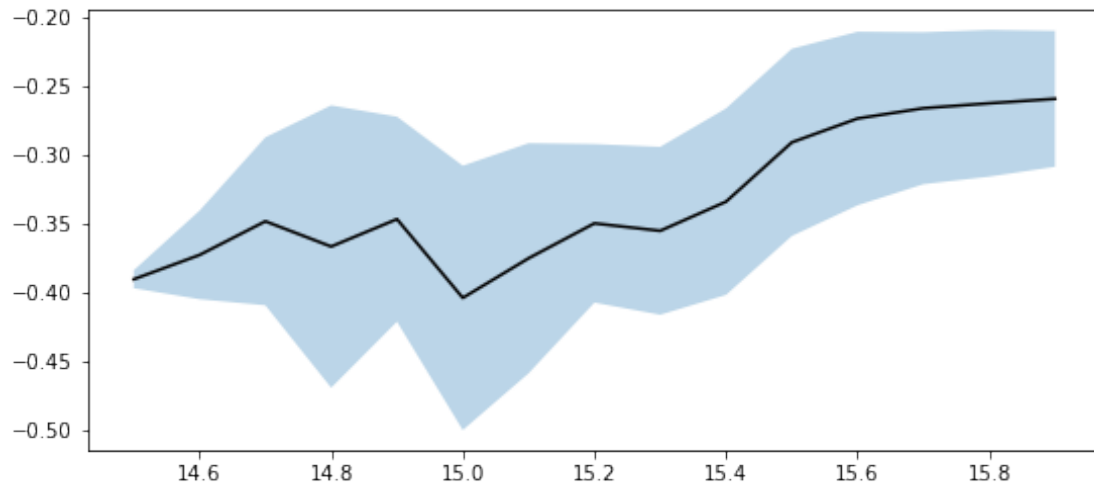
We use magnitudes between 15.0 and 16.0.

Aperture correction for i band:  
Correction: -0.3250293731689453  
Number of source used: 4829  
RMS: 0.07326213149972925

### 1.2.4 I.b - z band



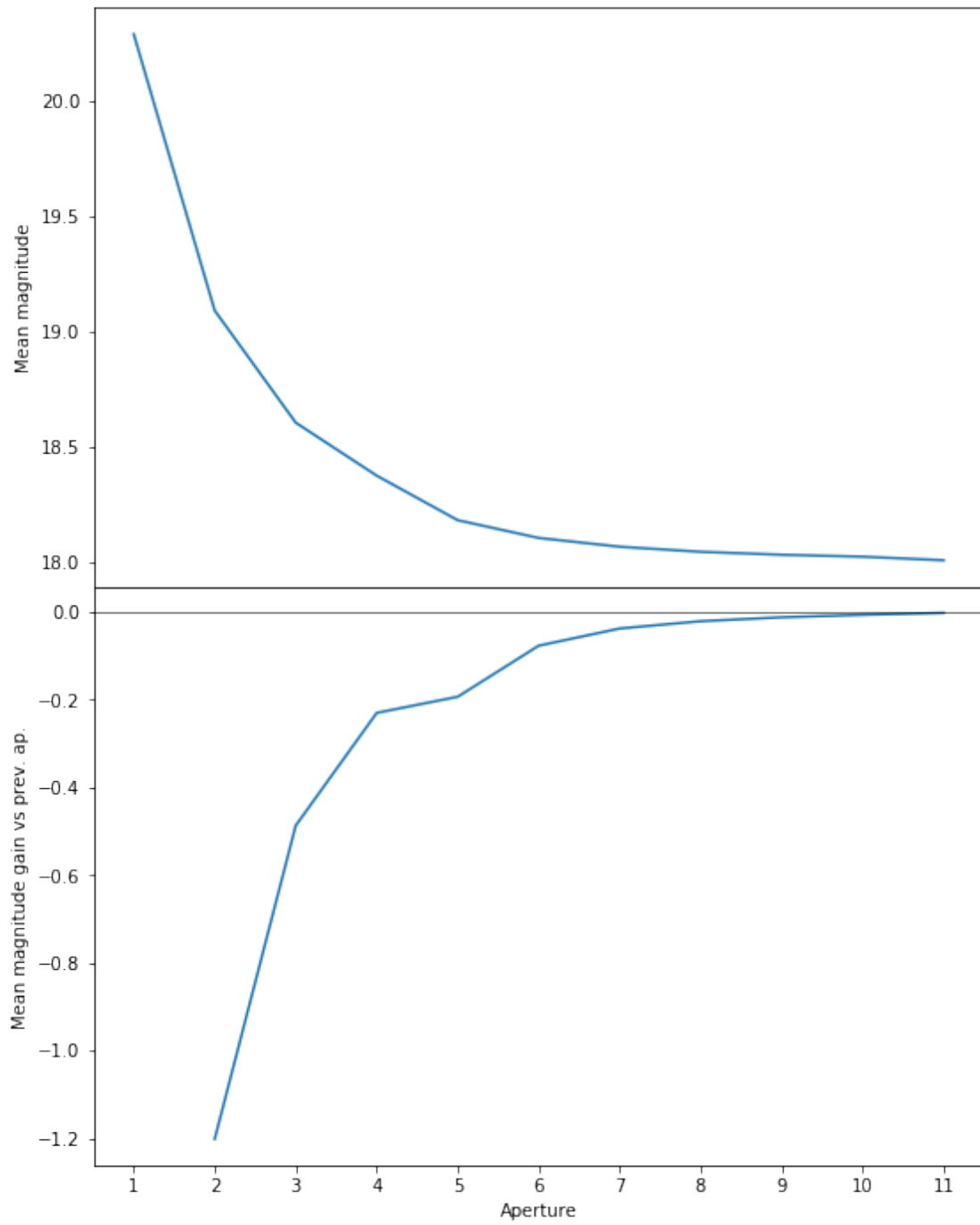
We will use aperture 57 as target.



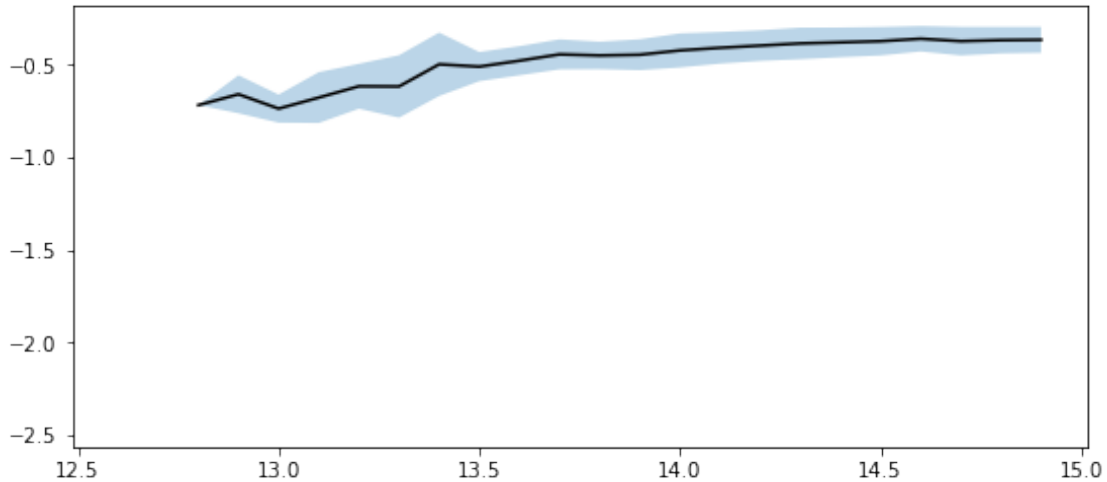
We use magnitudes between 15.0 and 16.0.

Aperture correction for z band:  
Correction: -0.26555299758911133  
Number of source used: 9874  
RMS: 0.055136808300606646

### 1.2.5 I.b - y band



We will use aperture 10 as target.



We use magnitudes between 15.0 and 16.0.

```
Aperture correction for y band:
Correction: -0.35900163650512695
Number of source used: 7472
RMS: 0.06888716175884298
```

### 1.3 2 - Column selection

```
/opt/anaconda3/envs/herschelhelp_internal/lib/python3.6/site-packages/astropy/table/column.py:10
Check the NumPy 1.11 release notes for more information.
ma.MaskedArray.__setitem__(self, index, value)
```

Out[24]: <IPython.core.display.HTML object>

### 1.4 II - Removal of duplicated sources

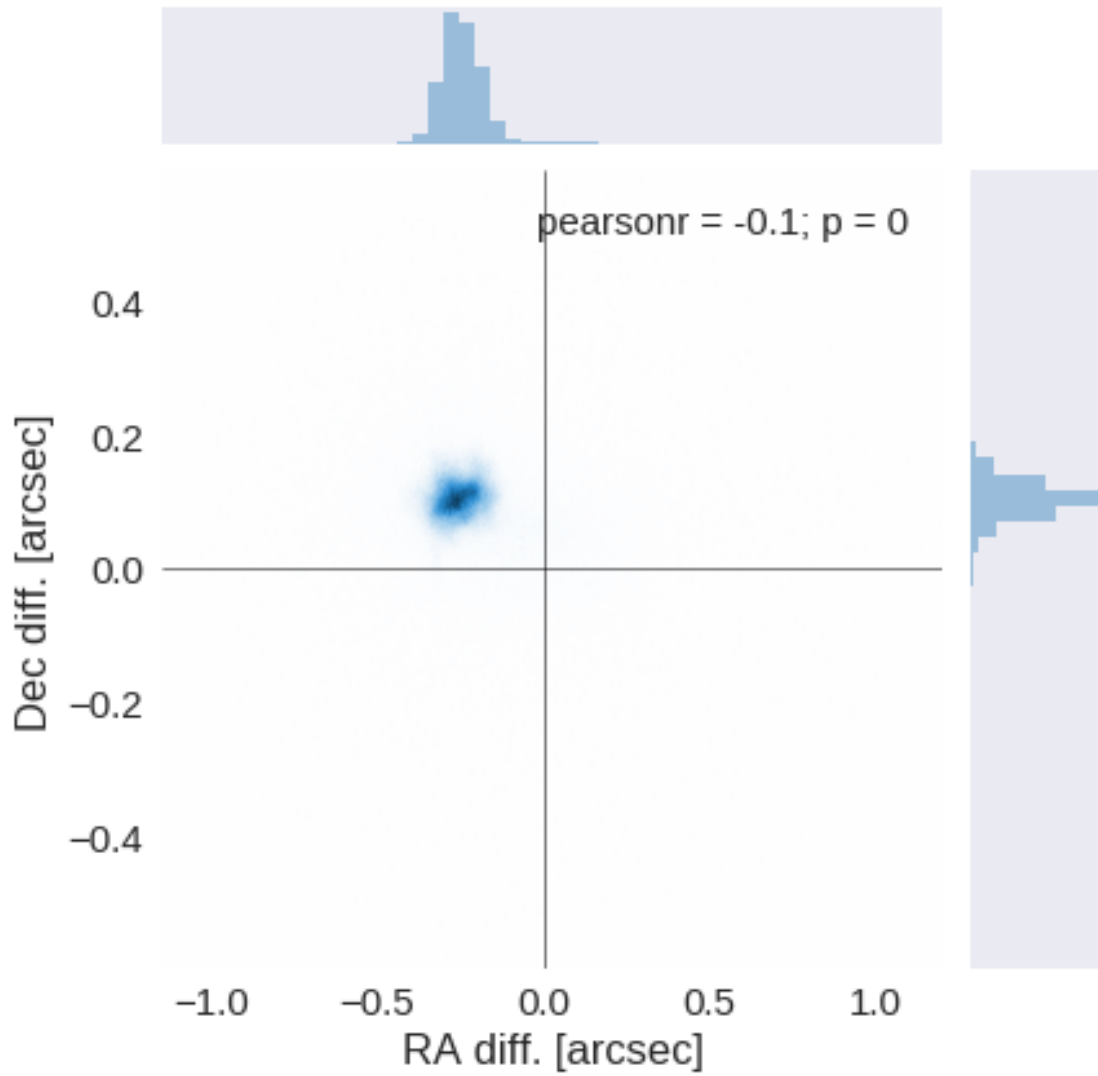
We remove duplicated objects from the input catalogues.

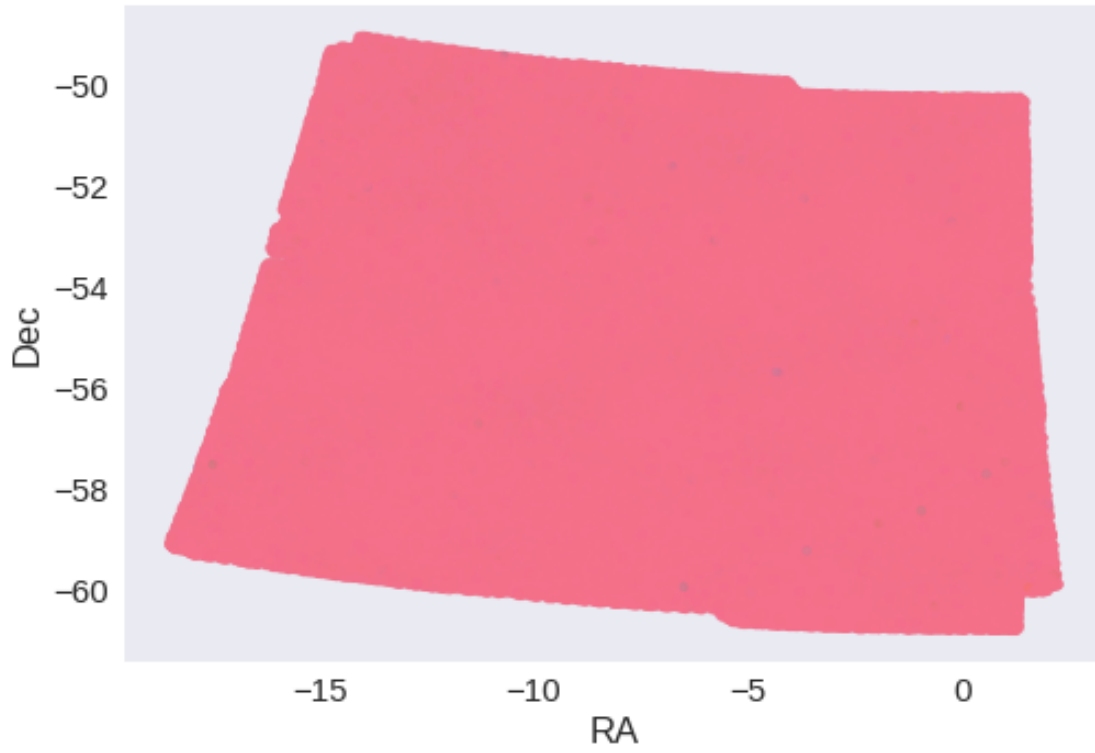
```
/opt/anaconda3/envs/herschelhelp_internal/lib/python3.6/site-packages/astropy/table/column.py:10
Check the NumPy 1.11 release notes for more information.
ma.MaskedArray.__setitem__(self, index, value)
```

```
The initial catalogue had 9307507 sources.
The cleaned catalogue has 9307386 sources (121 removed).
The cleaned catalogue has 121 sources flagged as having been cleaned
```

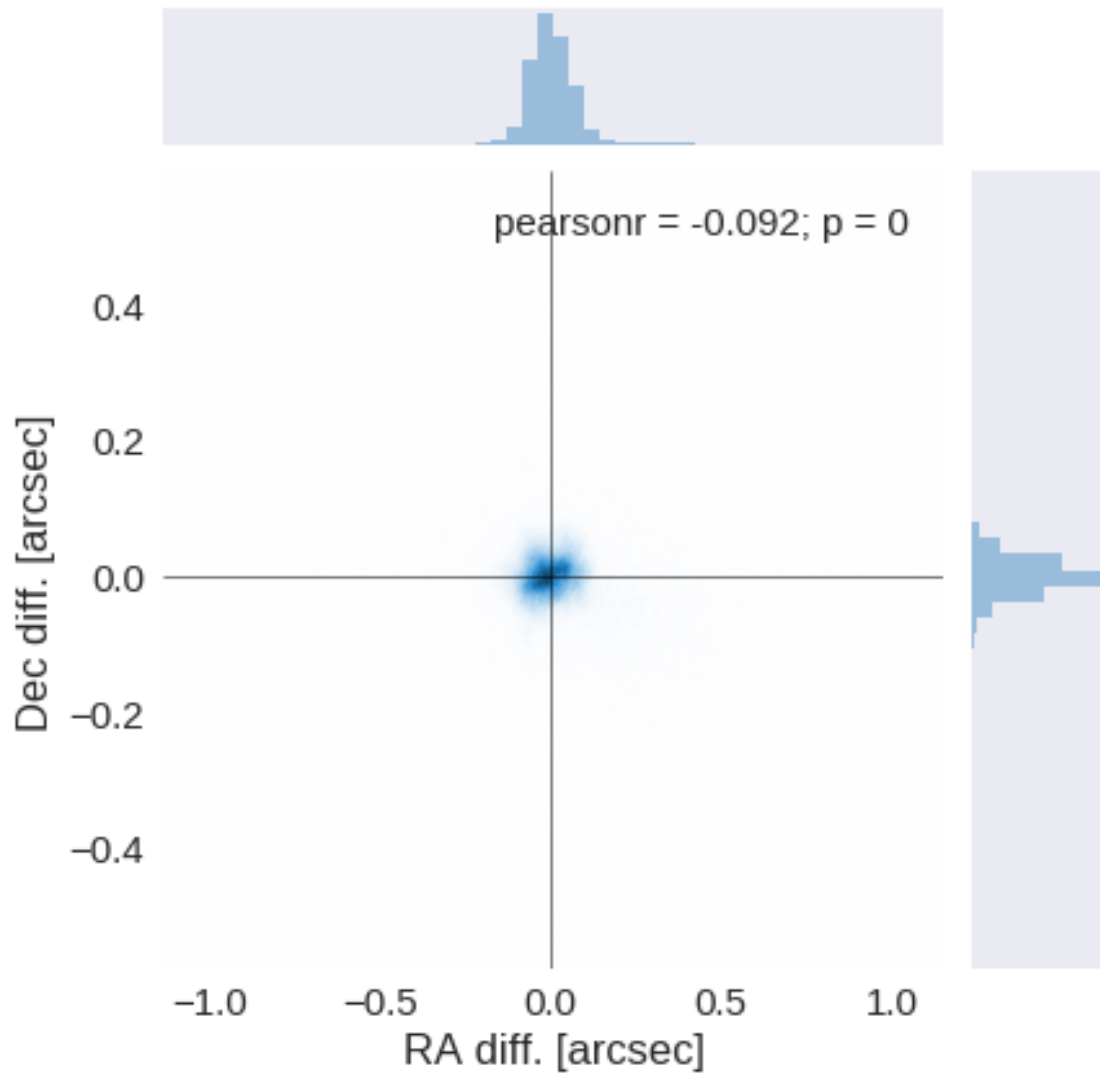
### 1.5 III - Astrometry correction

We match the astrometry to the Gaia one. We limit the Gaia catalogue to sources with a g band flux between the 30th and the 70th percentile. Some quick tests show that this give the lower dispersion in the results.

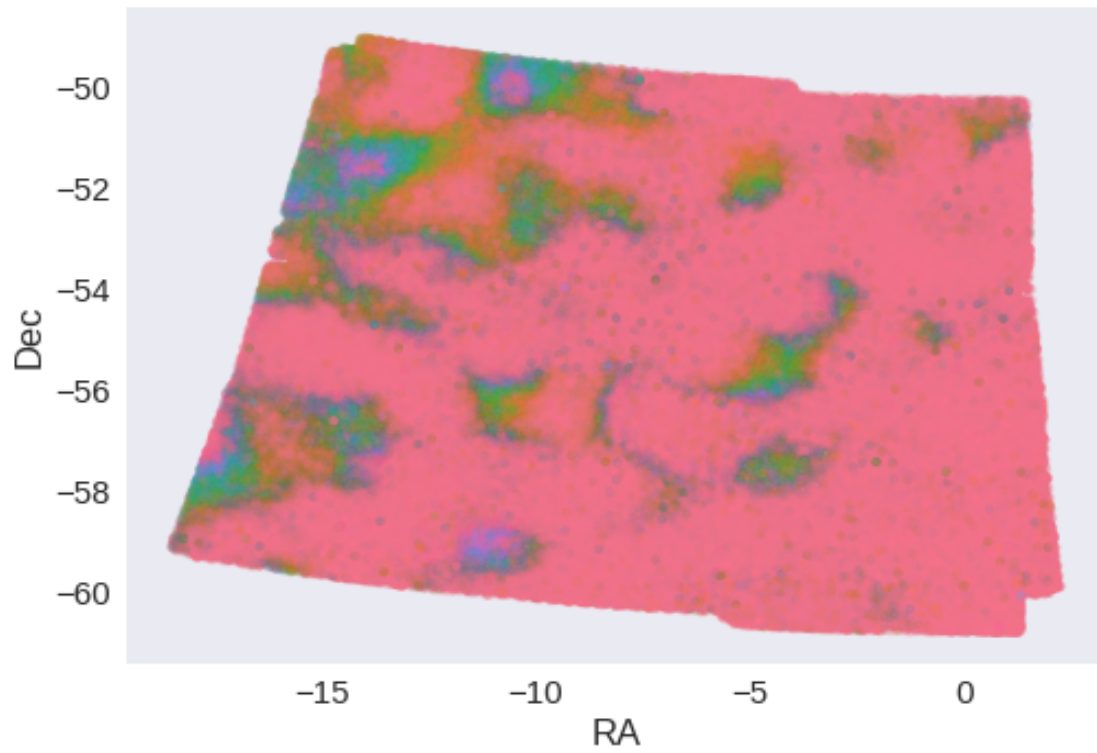




RA correction: 0.25125262735627985 arcsec  
Dec correction: -0.10440259524102657 arcsec







## 1.6 IV - Flagging Gaia objects

454986 sources flagged.

## 1.7 V - Flagging objects near bright stars

## 2 VI - Saving to disk

# 2\_Merging

March 8, 2018

## 1 SSDF master catalogue

This notebook presents the merge of the various pristine catalogues to produce HELP mater catalogue on SSDF.

This notebook was run with `herschelhelp_internal` version:  
0246c5d (Thu Jan 25 17:01:47 2018 +0000) [with local modifications]

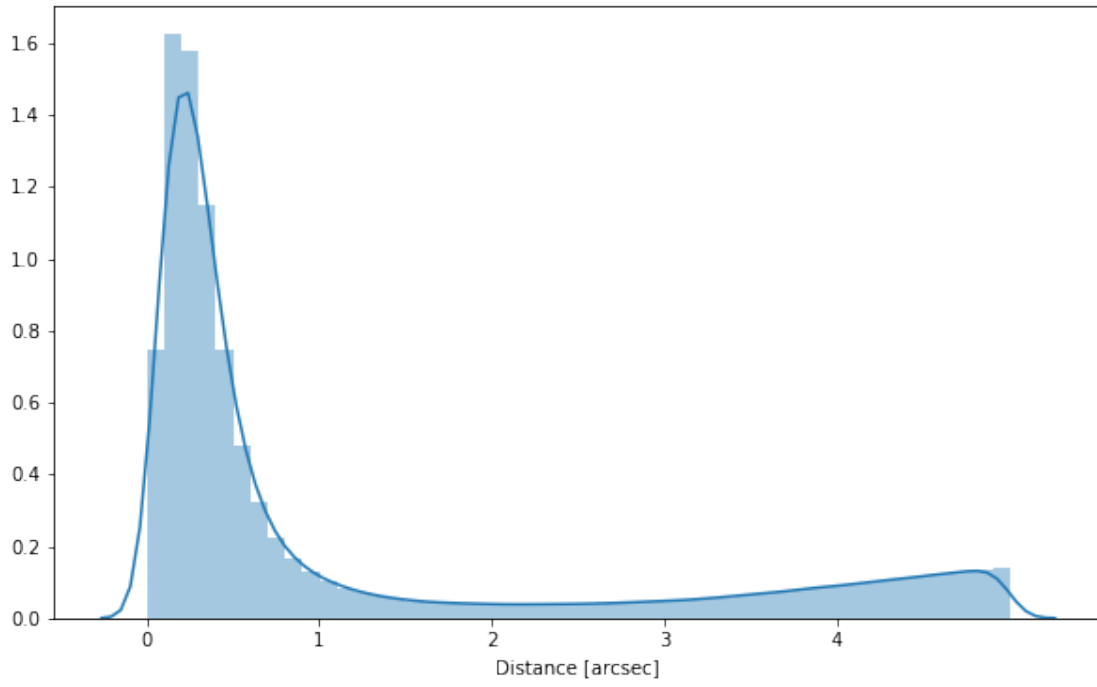
### 1.1 I - Reading the prepared pristine catalogues

### 1.2 II - Merging tables

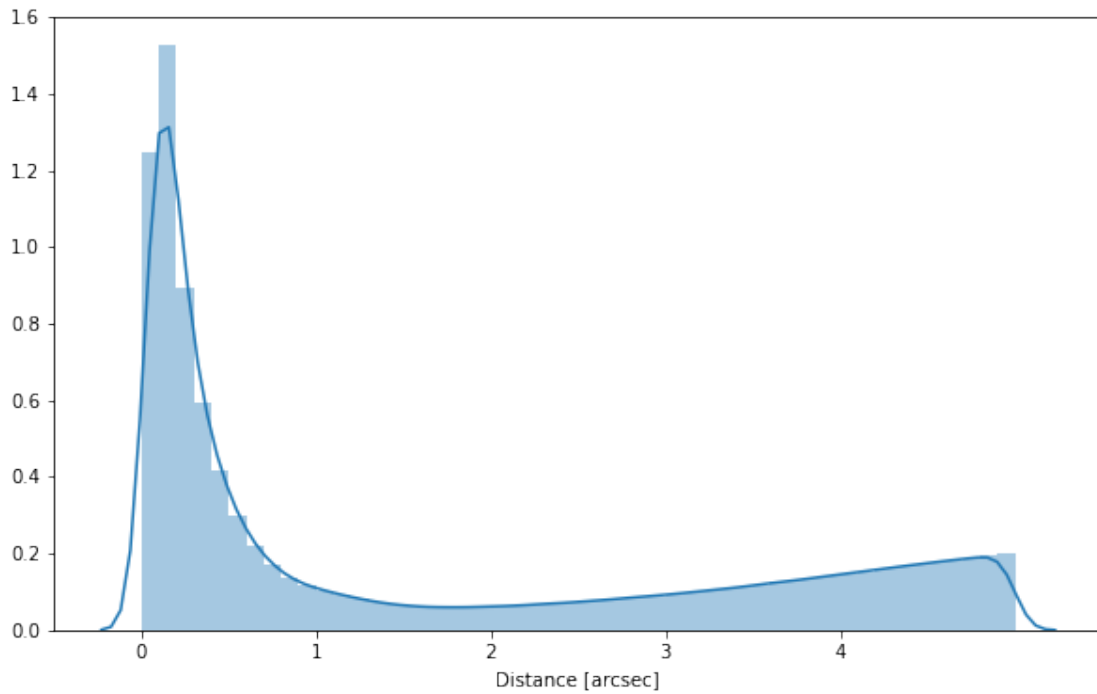
At every step, we look at the distribution of the distances to the nearest source in the merged catalogue to determine the best crossmatching radius.

### 1.2.1 VISTA-VHS

### 1.2.2 Add SSDF



### 1.2.3 Add DES



## 1.2.4 Cleaning

When we merge the catalogues, astropy masks the non-existent values (e.g. when a row comes only from a catalogue and has no counterparts in the other, the columns from the latest are masked for that row). We indicate to use NaN for masked values for floats columns, False for flag columns and -1 for ID columns.

```
Out[11]: <IPython.core.display.HTML object>
```

## 1.3 III - Merging flags and stellerity

Each pristine catalogue contains a flag indicating if the source was associated to a another nearby source that was removed during the cleaning process. We merge these flags in a single one.

Each pristine catalogue contains a flag indicating the probability of a source being a Gaia object (0: not a Gaia object, 1: possibly, 2: probably, 3: definitely). We merge these flags taking the highest value.

Each prisitine catalogue may contain one or several stellerity columns indicating the probability (0 to 1) of each source being a star. We merge these columns taking the highest value.

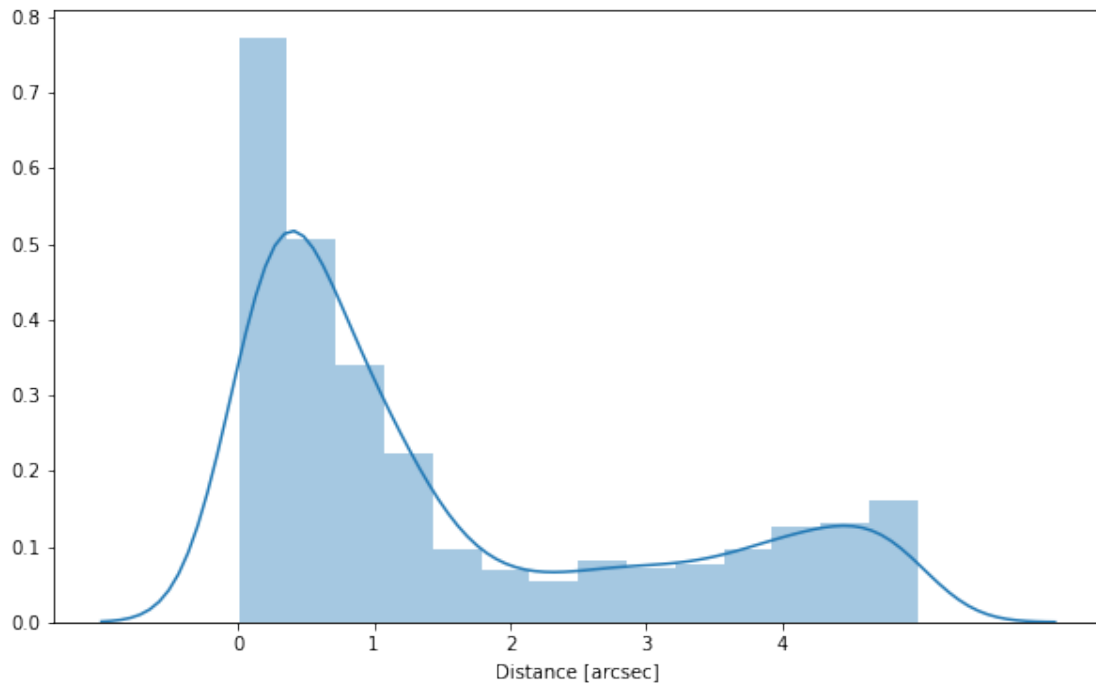
```
vhs_stellariry, ssdf_stellariry, des_stellariry
```

## 1.4 IV - Adding E(B-V) column

## 1.5 V - Adding HELP unique identifiers and field columns

OK!

## 1.6 VI - Cross-matching with the spec-z catalogue



## 1.7 VII - Choosing between multiple values for the same filter

There are no duplicate bands.

## 1.8 VIII.a Wavelength domain coverage

We add a binary flag `flag_optnir_obs` indicating that a source was observed in a given wavelength domain:

- 1 for observation in optical;
- 2 for observation in near-infrared;
- 4 for observation in mid-infrared (IRAC).

It's an integer binary flag, so a source observed both in optical and near-infrared but not in mid-infrared would have this flag at  $1 + 2 = 3$ .

*Note 1: The observation flag is based on the creation of multi-order coverage maps from the catalogues, this may not be accurate, especially on the edges of the coverage.*

*Note 2: Being on the observation coverage does not mean having fluxes in that wavelength domain. For sources observed in one domain but having no flux in it, one must take into consideration the different depths in the catalogue we are using.*

## 1.9 VIII.b Wavelength domain detection

We add a binary flag `flag_optnir_det` indicating that a source was detected in a given wavelength domain:

- 1 for detection in optical;
- 2 for detection in near-infrared;
- 4 for detection in mid-infrared (IRAC).

It's an integer binary flag, so a source detected both in optical and near-infrared but not in mid-infrared would have this flag at  $1 + 2 = 3$ .

*Note 1: We use the total flux columns to know if the source has flux, in some catalogues, we may have aperture flux and no total flux.*

To get rid of artefacts (chip edges, star flares, etc.) we consider that a source is detected in one wavelength domain when it has a flux value in **at least two bands**. That means that good sources will be excluded from this flag when they are on the coverage of only one band.

## 1.10 IX - Cross-identification table

We are producing a table associating to each HELP identifier, the identifiers of the sources in the pristine catalogue. This can be used to easily get additional information from them.

```
['vhs_id', 'ssdf_id', 'des_id', 'help_id', 'specz_id']
```

## 1.11 X - Adding HEALPix index

We are adding a column with a HEALPix index at order 13 associated with each source.

## 1.12 XI - Saving the catalogue

```
Missing columns: set()
```

# 3\_Checks\_and\_diagnostics

March 8, 2018

## 1 SSDF master catalogue

### 1.1 Checks and diagnostics

This notebook was run with herschelhelp\_internal version:  
0246c5d (Thu Jan 25 17:01:47 2018 +0000) [with local modifications]

Diagnostics done using: master\_catalogue\_ssdf\_20180221.fits

### 1.2 0 - Quick checks

```
/opt/anaconda3/envs/herschelhelp_internal/lib/python3.6/site-packages/numpy/core/numeric.py:301:
  format(shape, fill_value, array(fill_value).dtype), FutureWarning)
/opt/anaconda3/envs/herschelhelp_internal/lib/python3.6/site-packages/numpy/core/numeric.py:301:
  format(shape, fill_value, array(fill_value).dtype), FutureWarning)
```

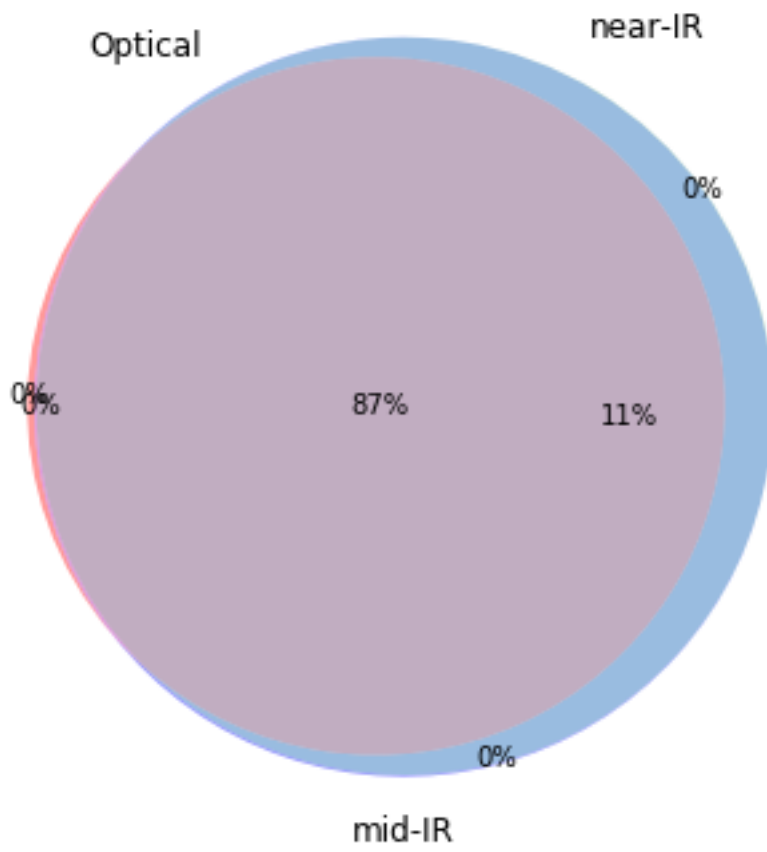
Table shows only problematic columns.

Out[4]: <IPython.core.display.HTML object>

### 1.3 I - Summary of wavelength domains

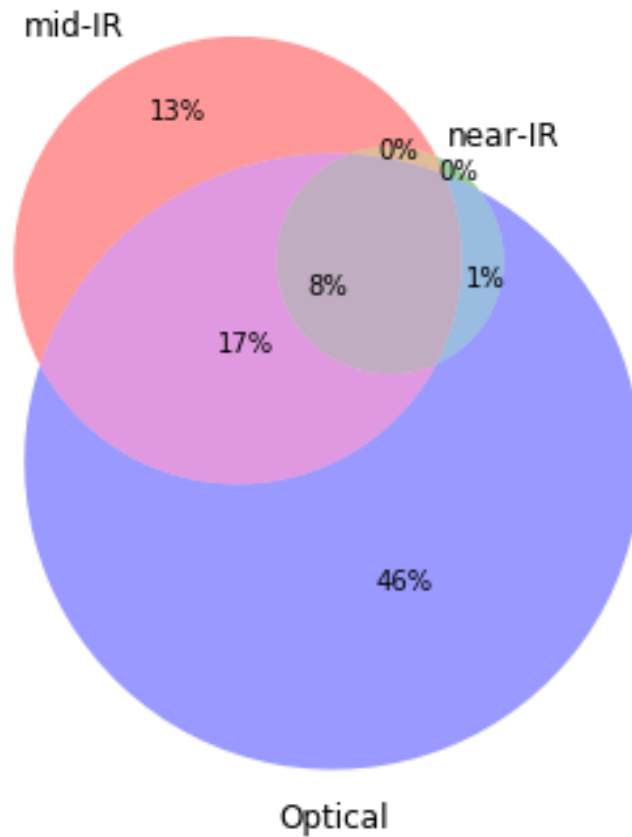
```
/opt/anaconda3/envs/herschelhelp_internal/lib/python3.6/site-packages/matplotlib_venn/_venn3.py:
  warnings.warn("Bad circle positioning")
```

# Wavelength domain observations





Detection of the 10,820,051 sources detected in any wavelength domains (among 12,661,903 sources)

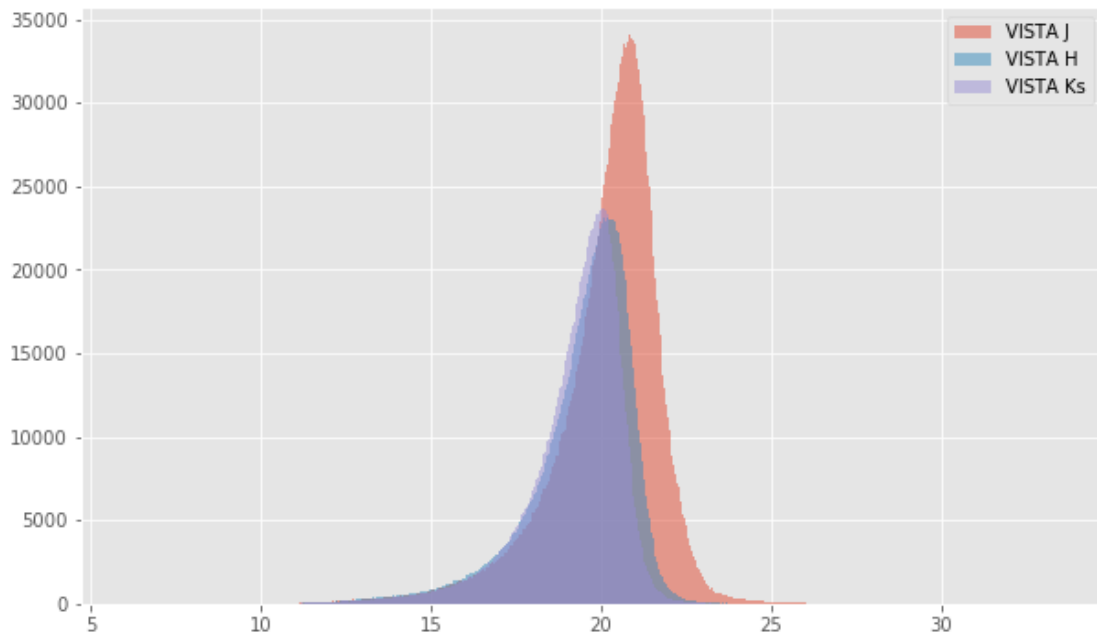
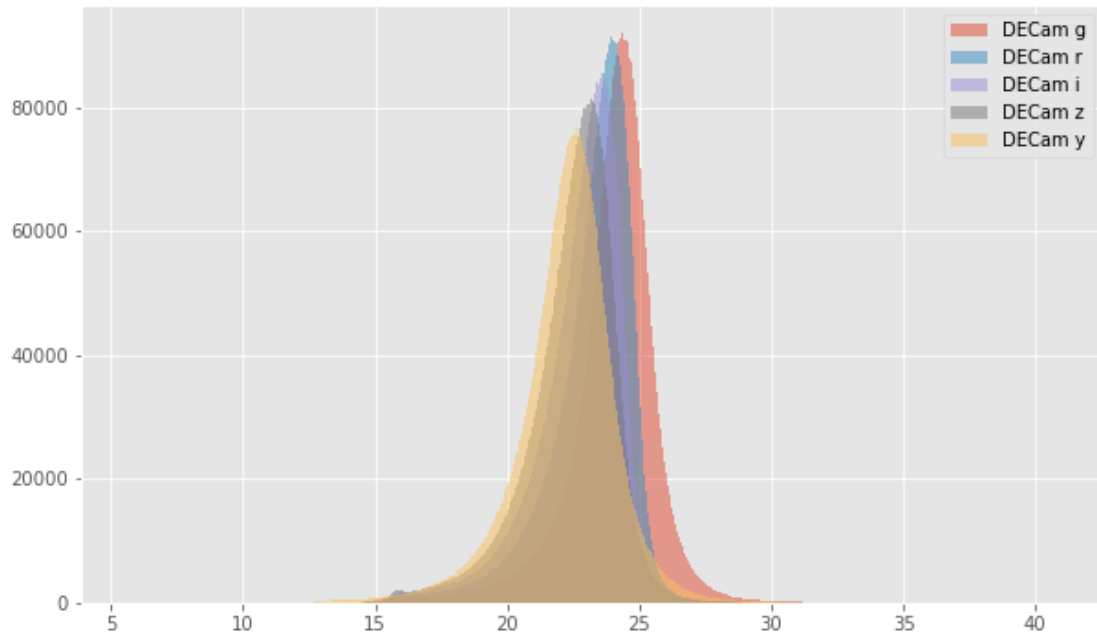


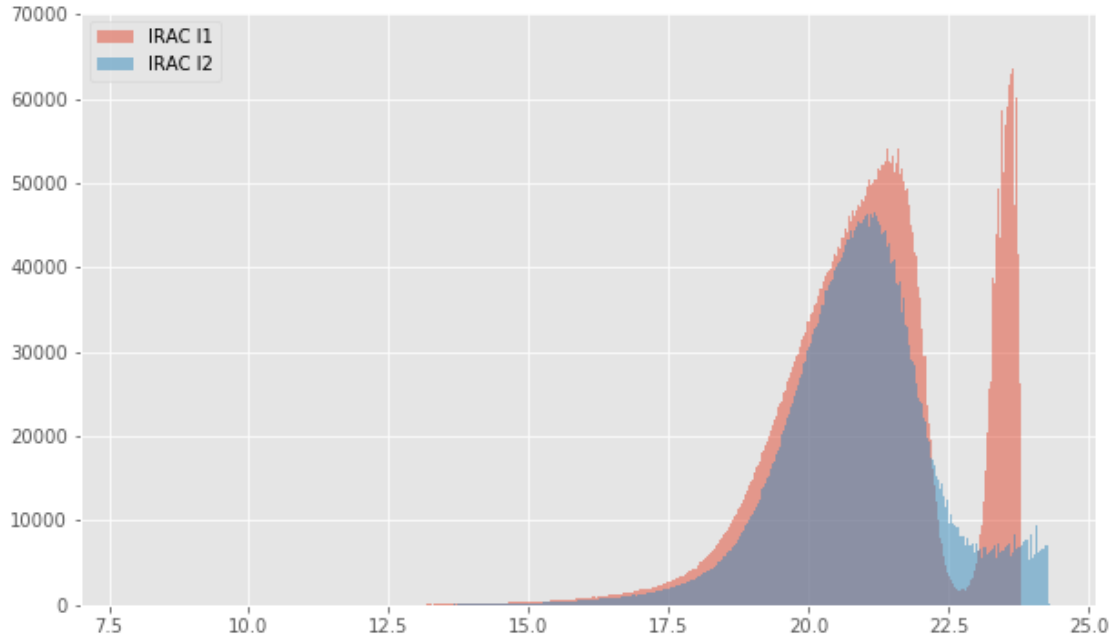
## 1.4 II - Comparing magnitudes in similar filters

The master list is composed of several catalogues containing magnitudes in similar filters on different instruments. We are comparing the magnitudes in these corresponding filters.

### 1.4.1 II.a - Comparing depths

We compare the histograms of the total aperture magnitudes of similar bands.





### 1.4.2 II.b - Comparing magnitudes

There are no similar bands from different instruments.

## 1.5 III - Comparing magnitudes to reference bands

Cross-match the master list to SDSS.

### 1.5.1 III.b - Comparing J and K bands to 2MASS

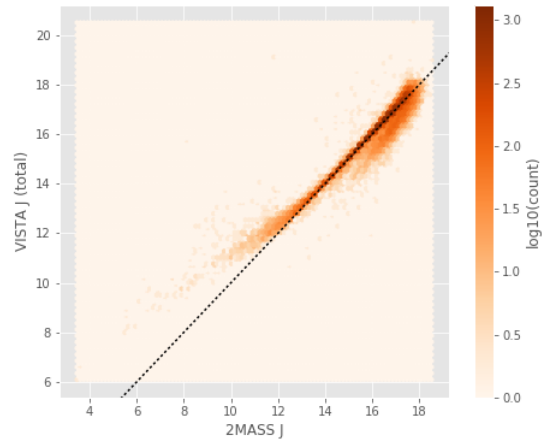
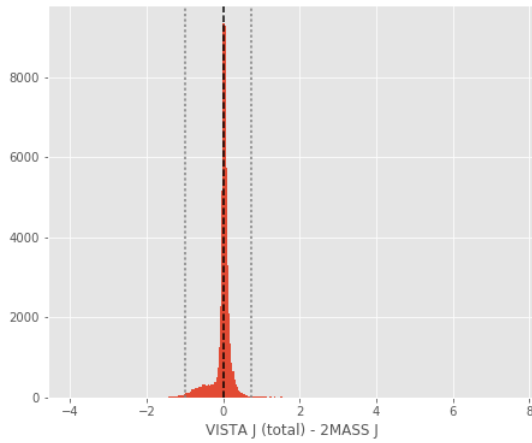
The catalogue is cross-matched to 2MASS-PSC withing 0.2 arcsecond. We compare the UKIDSS total J and K magnitudes to those from 2MASS.

The 2MASS magnitudes are “Vega-like” and we have to convert them to AB magnitudes using the zero points provided on [this page](#):

Band	F - 0 mag (Jy)
J	1594
H	1024
Ks	666.7

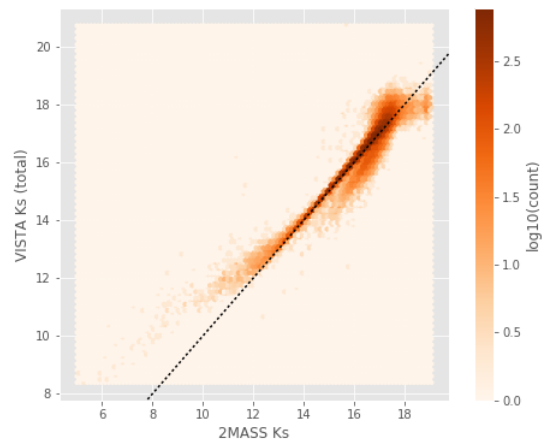
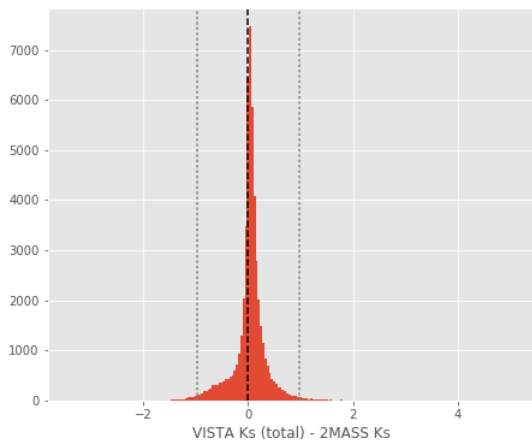
VISTA J (total) - 2MASS J:

- Median: 0.03
- Median Absolute Deviation: 0.06
- 1% percentile: -0.9993877432157341
- 99% percentile: 0.7214355603120987



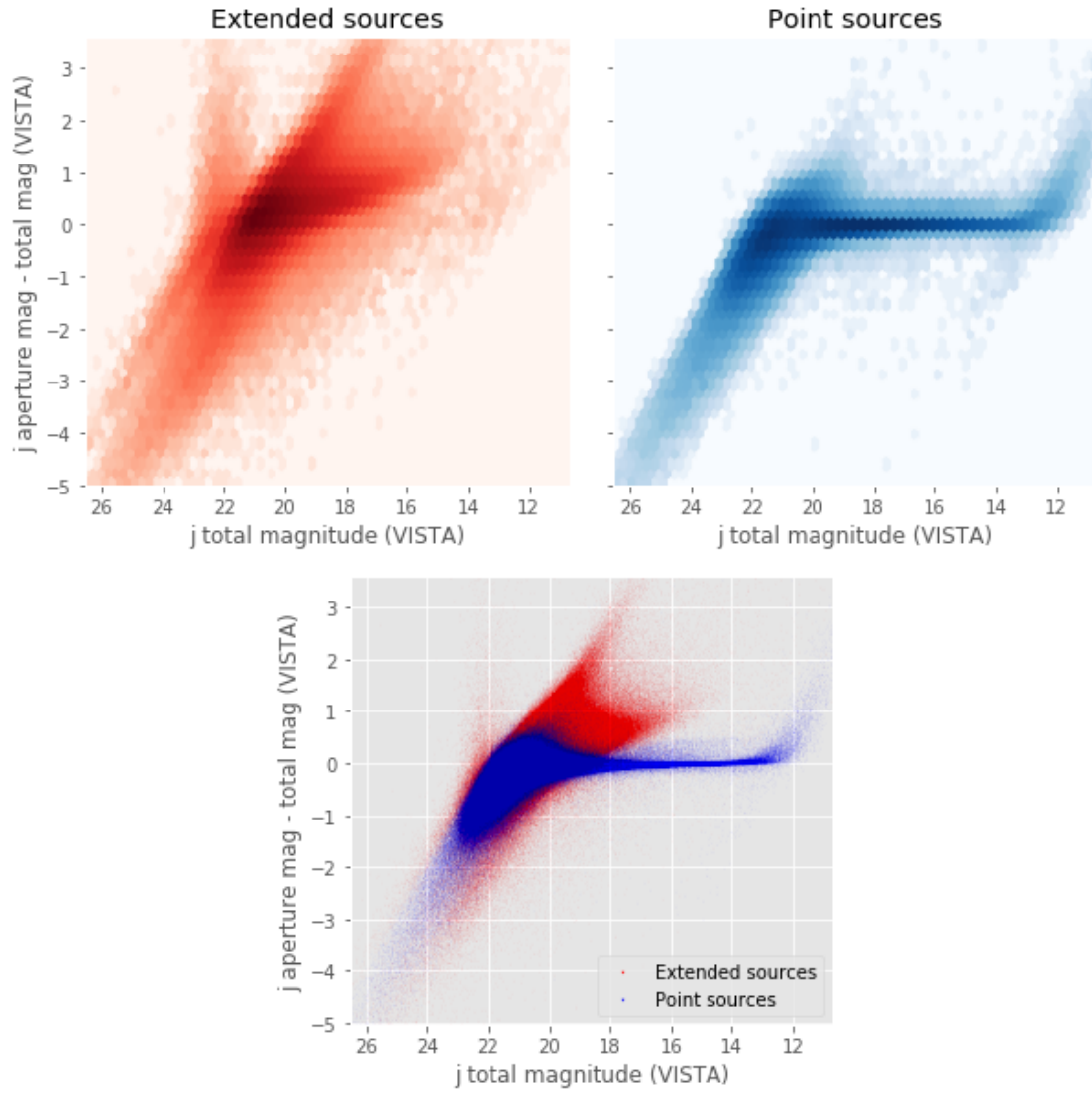
VISTA Ks (total) - 2MASS Ks:

- Median: 0.04
- Median Absolute Deviation: 0.09
- 1% percentile: -0.9690106682451767
- 99% percentile: 0.9662497443524783

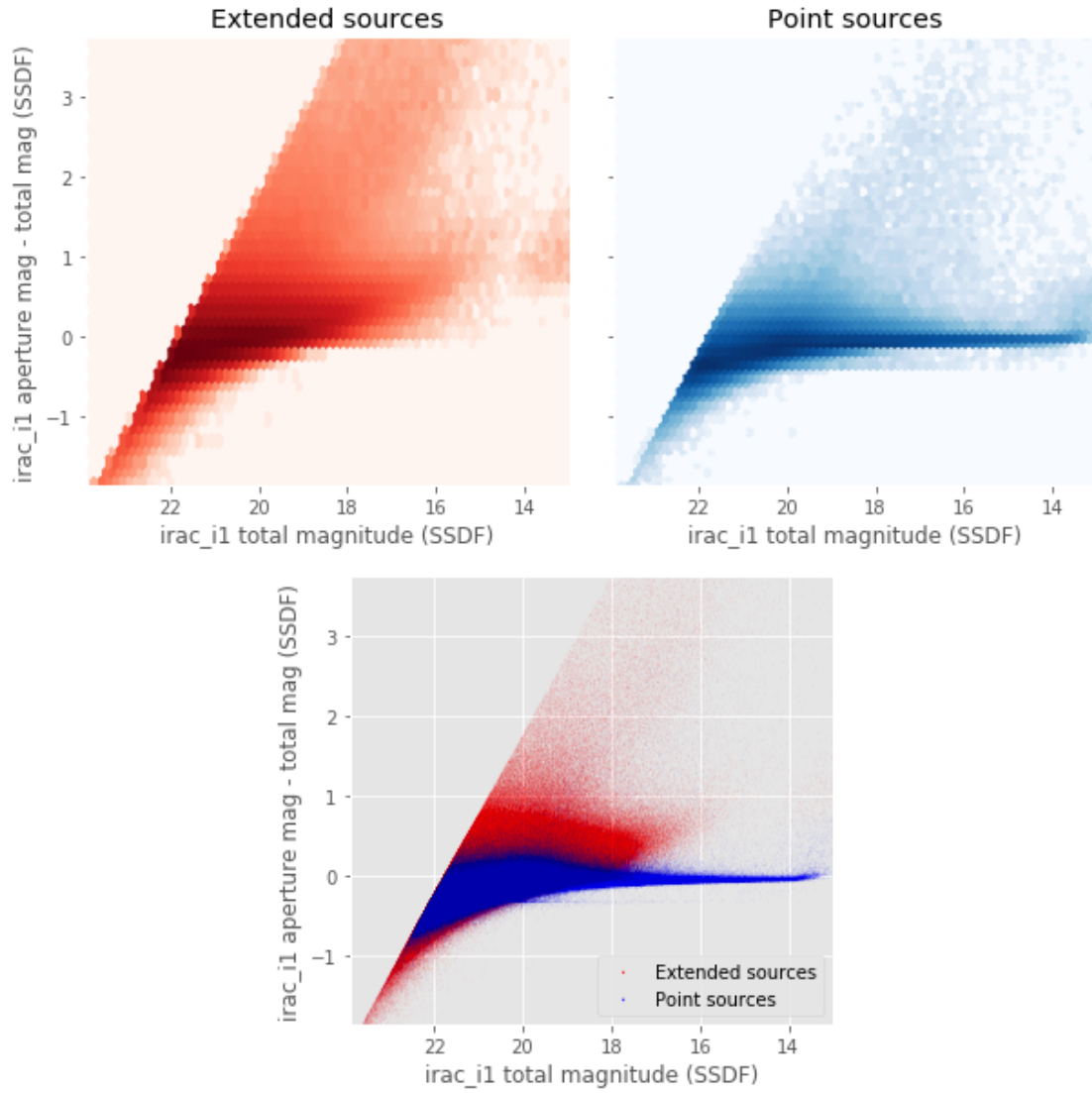


## 1.6 IV - Comparing aperture magnitudes to total ones.

Number of source used: 2057720 / 12661903 (16.25%)

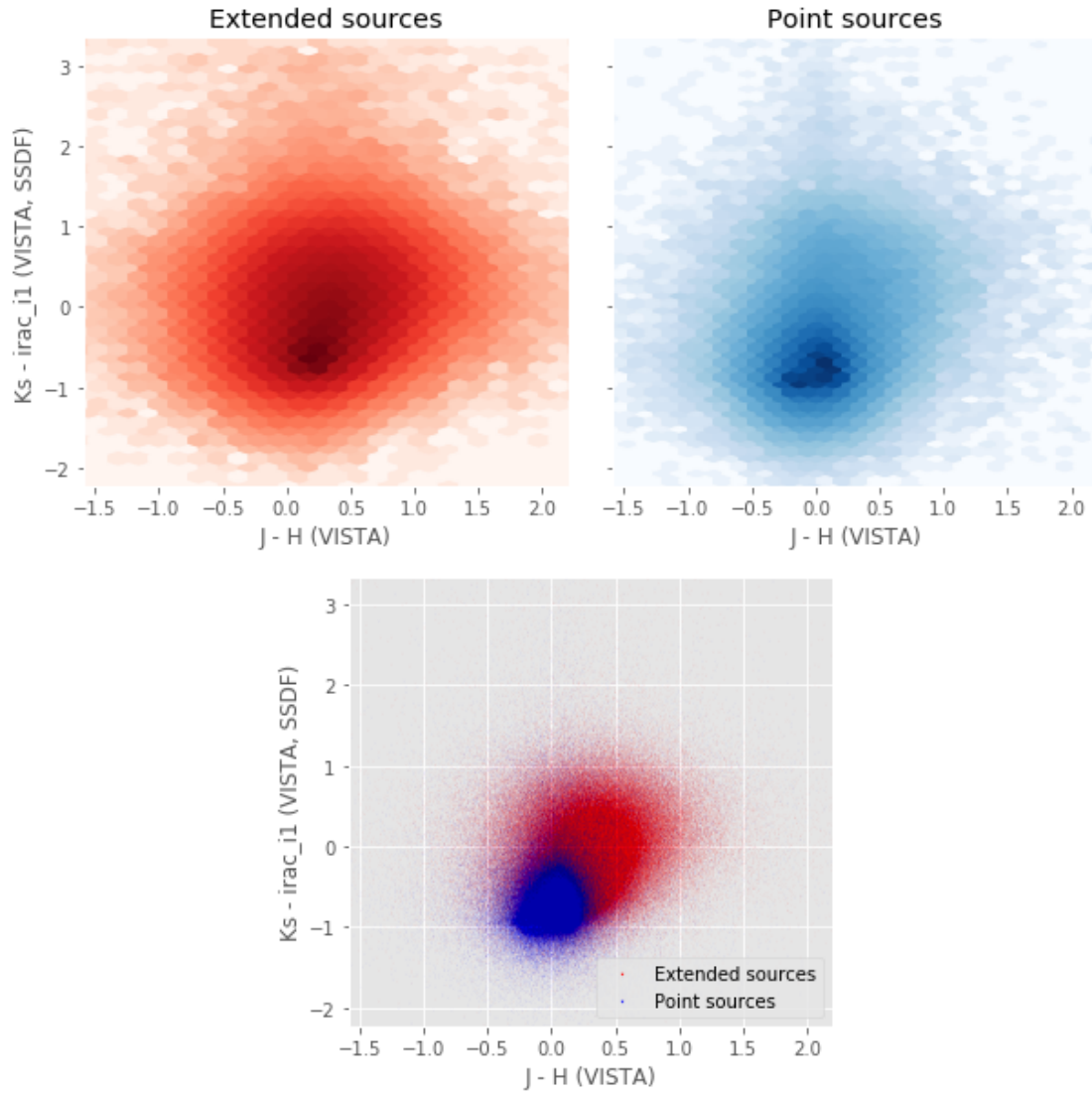


Number of source used: 4239455 / 12661903 (33.48%)

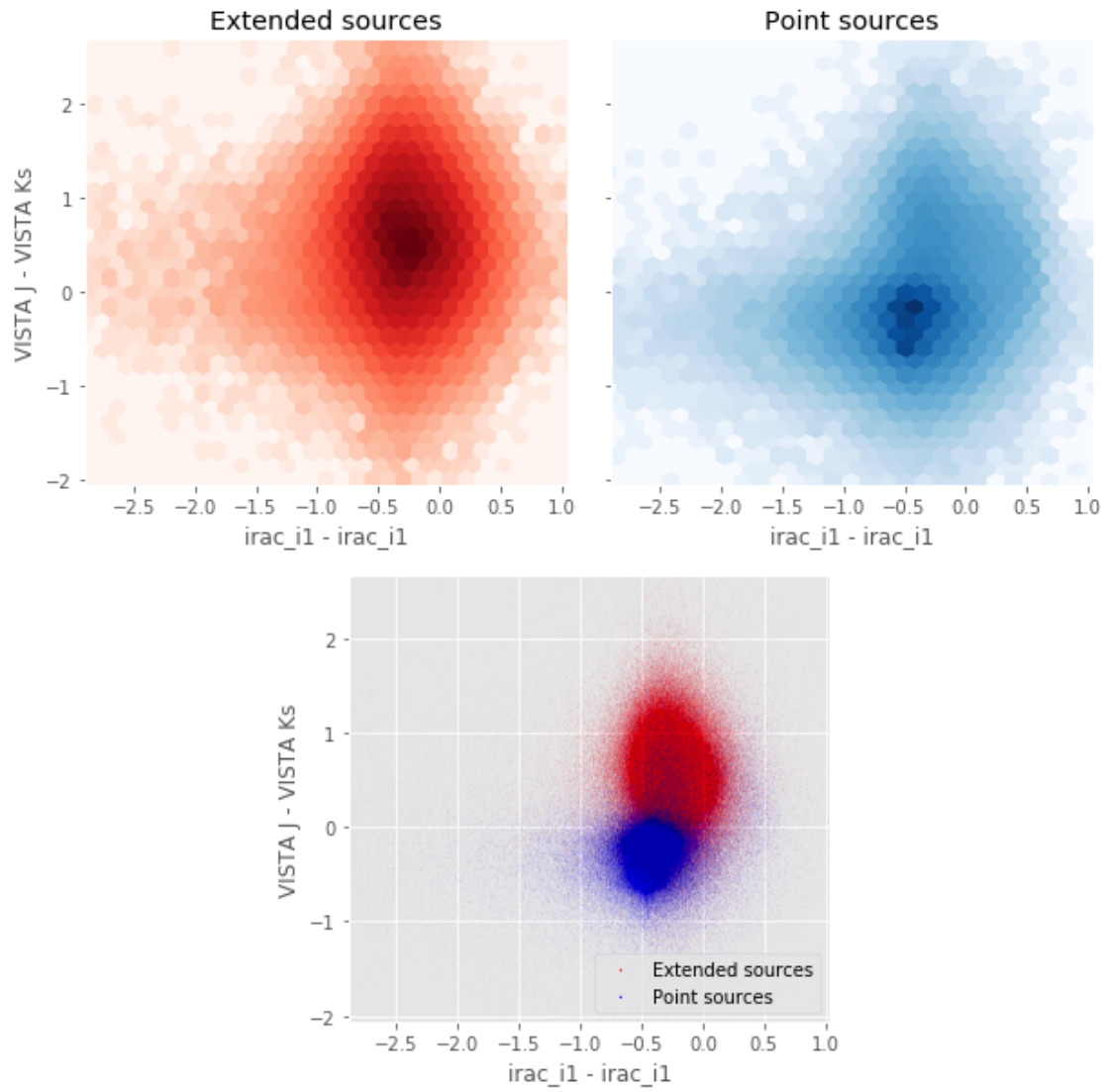


## 1.7 V - Color-color and magnitude-color plots

Number of source used: 745557 / 12661903 (5.89%)



Number of source used: 797742 / 12661903 (6.30%)





# 4\_Selection\_function

March 8, 2018

## 1 SSDF Selection Functions

### 1.1 Depth maps and selection functions

The simplest selection function available is the field MOC which specifies the area for which there is Herschel data. Each pristine catalogue also has a MOC defining the area for which that data is available.

The next stage is to provide mean flux standard deviations which act as a proxy for the catalogue's  $5\sigma$  depth

```
This notebook was run with herschelhelp_internal version:  
0246c5d (Thu Jan 25 17:01:47 2018 +0000) [with local modifications]  
This notebook was executed on:  
2018-02-28 11:56:29.478992
```

Depth maps produced using: master\_catalogue\_ssdf\_20180221.fits

### 1.2 I - Group masterlist objects by healpix cell and calculate depths

We add a column to the masterlist catalogue for the target order healpix cell per object.

### 1.3 II Create a table of all Order=13 healpix cells in the field and populate it

We create a table with every order=13 healpix cell in the field MOC. We then calculate the healpix cell at lower order that the order=13 cell is in. We then fill in the depth at every order=13 cell as calculated for the lower order cell that that the order=13 cell is inside.

```
Out[9]: <IPython.core.display.HTML object>
```

```
Out[11]: <IPython.core.display.HTML object>
```

```
Out[12]: <IPython.core.display.HTML object>
```

## 1.4 III - Save the depth map table

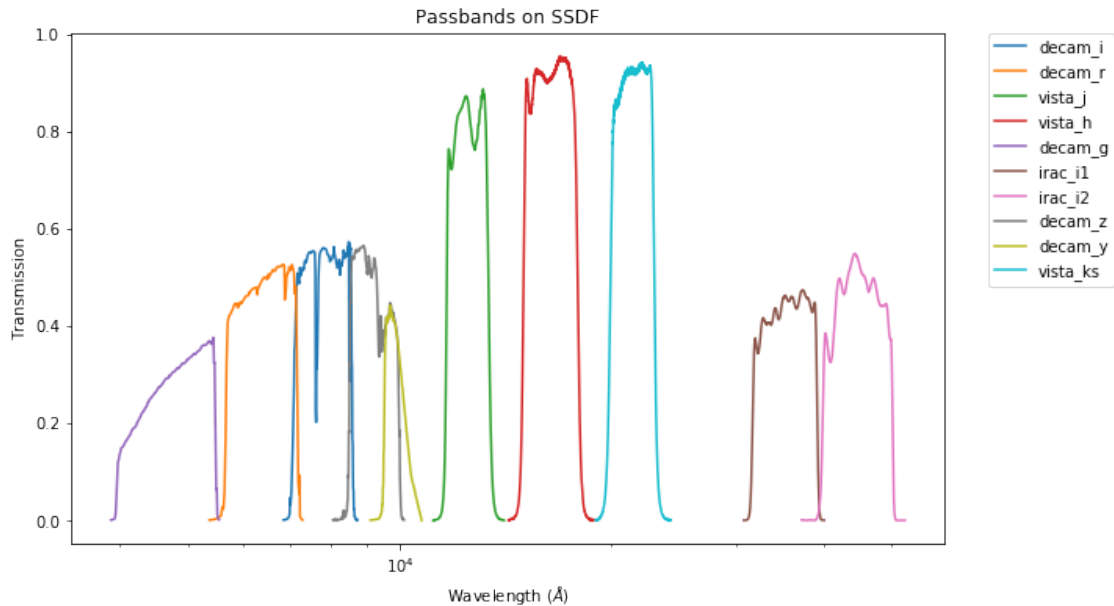
## 1.5 IV - Overview plots

### 1.5.1 IV.a - Filters

First we simply plot all the filters available on this field to give an overview of coverage.

```
Out[14]: {'decam_g',  
          'decam_i',  
          'decam_r',  
          'decam_y',  
          'decam_z',  
          'irac_i1',  
          'irac_i2',  
          'vista_h',  
          'vista_j',  
          'vista_ks'}
```

```
Out[15]: <matplotlib.text.Text at 0x7f9e9fd69e48>
```



### 1.5.2 IV.a - Depth overview

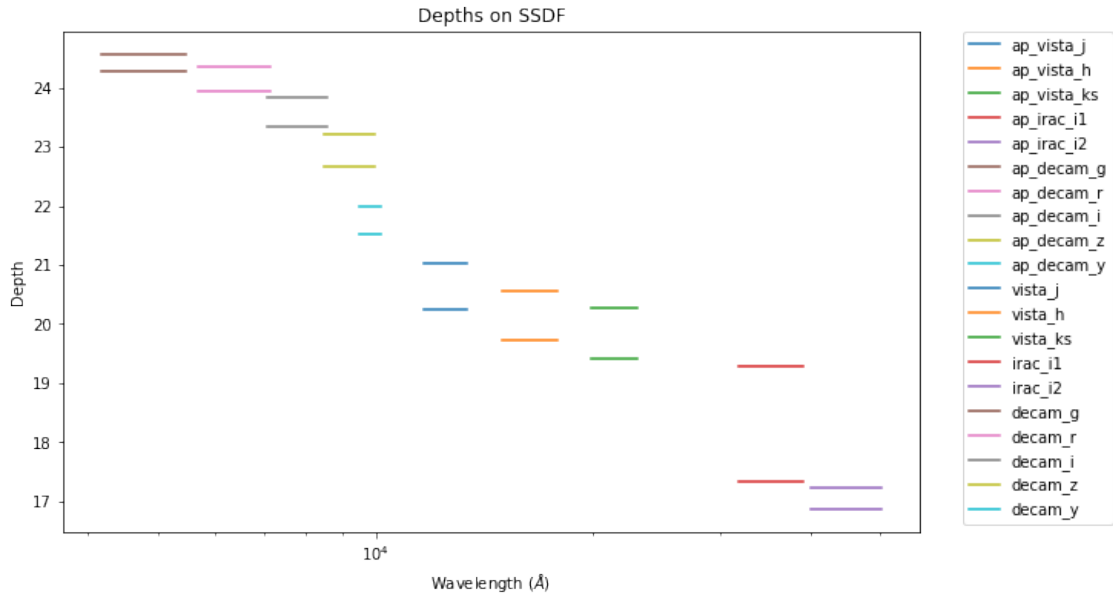
Then we plot the mean depths available across the area a given band is available

```
vista_j: mean flux error: 2.8469159603118896, 3sigma in AB mag (Aperture): 21.57126024530843  
vista_h: mean flux error: 4.345226764678955, 3sigma in AB mag (Aperture): 21.112165748362862  
vista_ks: mean flux error: 5.582312107086182, 3sigma in AB mag (Aperture): 20.84016157760791  
irac_i1: mean flux error: 85.45731353759766, 3sigma in AB mag (Aperture): 17.877823772998816
```

irac\_i2: mean flux error: 92.01163482666016, 3sigma in AB mag (Aperture): 17.797589995338804  
decam\_g: mean flux error: 0.10788271051841146, 3sigma in AB mag (Aperture): 25.12481723963301  
decam\_r: mean flux error: 0.13242671804157846, 3sigma in AB mag (Aperture): 24.90225782326261  
decam\_i: mean flux error: 0.21237251026390613, 3sigma in AB mag (Aperture): 24.389451111995264  
decam\_z: mean flux error: 0.3743137466144476, 3sigma in AB mag (Aperture): 23.77410742118841  
decam\_y: mean flux error: 1.151574241760951, 3sigma in AB mag (Aperture): 22.55396700799789  
vista\_j: mean flux error: 5.782756328582764, 3sigma in AB mag (Total): 20.801859631785398  
vista\_h: mean flux error: 9.392768859863281, 3sigma in AB mag (Total): 20.275212775160163  
vista\_ks: mean flux error: 12.380193710327148, 3sigma in AB mag (Total): 19.97537826306813  
irac\_i1: mean flux error: 14.00174617767334, 3sigma in AB mag (Total): 19.84174136185556  
irac\_i2: mean flux error: 130.55184936523438, 3sigma in AB mag (Total): 17.41773929242374  
decam\_g: mean flux error: 0.14072265950195587, 3sigma in AB mag (Total): 24.83628677767515  
decam\_r: mean flux error: 0.19366893763945733, 3sigma in AB mag (Total): 24.489546937546088  
decam\_i: mean flux error: 0.33351221299309886, 3sigma in AB mag (Total): 23.899417507911302  
decam\_z: mean flux error: 0.6294928812792083, 3sigma in AB mag (Total): 23.209719805243004  
decam\_y: mean flux error: 1.8069069035025214, 3sigma in AB mag (Total): 22.06485742028314

ap\_vista\_j (11670.0, 13380.0, 1710.0)  
ap\_vista\_h (15000.0, 17900.0, 2900.0)  
ap\_vista\_ks (19930.0, 23010.0, 3080.0)  
ap\_irac\_i1 (31754.0, 39164.801, 7410.8008)  
ap\_irac\_i2 (39980.102, 50052.301, 10072.199)  
ap\_decam\_g (4180.0, 5470.0, 1290.0)  
ap\_decam\_r (5680.0, 7150.0, 1470.0)  
ap\_decam\_i (7090.0, 8560.0, 1470.0)  
ap\_decam\_z (8490.0, 9960.0, 1470.0)  
ap\_decam\_y (9510.0, 10170.0, 660.0)  
vista\_j (11670.0, 13380.0, 1710.0)  
vista\_h (15000.0, 17900.0, 2900.0)  
vista\_ks (19930.0, 23010.0, 3080.0)  
irac\_i1 (31754.0, 39164.801, 7410.8008)  
irac\_i2 (39980.102, 50052.301, 10072.199)  
decam\_g (4180.0, 5470.0, 1290.0)  
decam\_r (5680.0, 7150.0, 1470.0)  
decam\_i (7090.0, 8560.0, 1470.0)  
decam\_z (8490.0, 9960.0, 1470.0)  
decam\_y (9510.0, 10170.0, 660.0)

Out[20]: <matplotlib.text.Text at 0x7f9e9f9770f0>



### 1.5.3 IV.c - Depth vs coverage comparison

How best to do this? Colour/intensity plot over area? Percentage coverage vs mean depth?

Out [21]: <matplotlib.text.Text at 0x7f9e9f9c5668>

