intro_project

April 27, 2022

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[1]: # Using ML to classify LAEs in the NEP Field
     import tables as tb
     import numpy as np
     import matplotlib.pyplot as plt
     from matplotlib.colors import LogNorm
     from astropy import constants as const
     from astropy.table import Table, column, join
     from astropy.io import fits
     import astropy.units as u
     from astropy.coordinates import SkyCoord
     from astropy.visualization import ZScaleInterval
     from regions import CircleSkyRegion, CirclePixelRegion
     from hetdex_api.survey import Survey, FiberIndex
     from hetdex_api.config import HDRconfig
     from hetdex_api.detections import Detections
     from hetdex_api.elixer_widget_cls import ElixerWidget
     from hetdex_tools.get_spec import get_spectra
     import pandas as pd
     import seaborn as sb
```

[2]: det_object = Detections('hdr2.1', loadtable = False)

```
[3]: # Once it has loaded you want to filter out the data by selecting those that

→ are in the NEP field

# to do this I will give you the verticies of a box that will encompass all the

→ NEP field - Oscar

# The center of the NEP field is given by:

# NEP Central Coordinates:

# R.A. = 18hours00minutes00seconds, decl. = 66 degree 33minute 38.552 arcmin
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# Then make a radius of 3.5 degrees centered above and find all the RA and DEC_{
m L}
     \rightarrow coordinates
     # in the DF that are within this circle
     # creating the circle region in the sky (NEP field)
     ra = '18h00m00s'
     dec = '+66d33m38.552s'
     center_sky_coords = SkyCoord(ra, dec, frame = 'icrs')
     maskregion = det_object.query_by_coords(center_sky_coords, 3.5 * u.deg)
     detects_in_NEP = det_object[maskregion]
                                                # Sources within the NEP footprint
[8]: spec_test = spec_table = detects_in_NEP.get_spectrum(detects_in_NEP.detectid[0])
[9]: spec_test
[9]: <Table length=1036>
     wave1d
                                                     spec1d_err
                         spec1d
     Angstrom 1e-17 erg / (Angstrom cm2 s) 1e-17 erg / (Angstrom cm2 s)
     float32
                       float32
                                                      float32
       3470.0
                                  1.3739702
                                                                 3.301174
       3472.0
                                  1.3739702
                                                                 3.301174
       3474.0
                                  1.3739702
                                                                 3.301174
       3476.0
                                  1.3739702
                                                                 3.301174
       3478.0
                                  1.3739702
                                                                 3.301174
       3480.0
                                 1.3739702
                                                                 3.301174
       3482.0
                                  1.3739702
                                                                 3.301174
       3484.0
                                  1.3739702
                                                                 3.301174
       3486.0
                                  1.3739702
                                                                 3.301174
       5522.0
                                  1.9421544
                                                                1.8177477
       5524.0
                                  1.9345994
                                                               1.8484716
       5526.0
                                  1.9345994
                                                               1.8484716
       5528.0
                                  1.9345994
                                                                1.8484716
       5530.0
                                  1.9345994
                                                               1.8484716
       5532.0
                                  1.9345994
                                                                1.8484716
       5534.0
                                  1.9345994
                                                               1.8484716
       5536.0
                                  1.9345994
                                                                1.8484716
       5538.0
                                  1.9345994
                                                                1.8484716
       5540.0
                                  1.9345994
                                                                1.8484716
[4]: num_detects = np.size(detects_in_NEP.detectid) # qetting the number of
     → detections in NEP which is 69799
     # this part is creating a 2d list which will then be inputting into a_{\sqcup}
      → dictionary to be converted into an astropy Table
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# 2d list because I wanted a list of lists to hold all the spec1d and spec1d_\( \) \( \rightarror values \)

rows, cols = (num_detects, 1)

spec_ls = [[0 for i in range(cols)] for j in range(rows)]

# going from a list to an astropy array

for i in range(len(spec_ls)):

    spec_table = detects_in_NEP.get_spectrum(detects_in_NEP.detectid[i])

    spec_ls[i] = np.array(spec_table['spec1d']) # turned into array because_\( \rightarror \) \( \rightarror \) astropy column ['spec1d'] was confusing to work with
```

Want to make a table where the first row is the detectid, then I have rows for, spect1d (array), spec1d error (array), line info like EW, line flux, magnitude, RA, and DEC.

[7]: detect info

```
[7]: <Table length=69799>
     detects
               wavelength ...
                                spec1d [1036]
                                                          spec1d_err [1036]
      int64
                float32 ...
                                     float32
                                                              float32
                   4795.6 ...
    2100395300
                               1.3739702 .. 1.9345994
                                                        3.301174 .. 1.8484716
    2100395301
                   3684.0 ... 0.25948396 ... 0.104944706 4.9196553 ... 1.5617634
                  4407.54 ... -0.18306294 .. 0.120995596
                                                          3.343359 .. 1.752679
    2100395303
                  3779.01 ...
                               3.5947866 .. 29.83107 5.0278735 .. 3.9272287
    2100395308
                                 7.810404 .. 8.485255 4.6782527 .. 2.2743156
    2100395309
                  3779.08 ...
                  3777.96 ... 5.4102297 .. 16.955706
    2100395310
                                                          5.465975 .. 3.143807
```

```
3540.37 ...
                              7.232119 .. 36.326458
                                                       5.369271 .. 4.3121314
2100395312
              3779.73 ...
                              5.690473 .. 39.514835
                                                         4.060331 .. 5.03607
2100395314
                             6.0763054 .. 14.341705
2100395323
              3783.45 ...
                                                       4.181327 .. 2.8485444
                              3.6374705 .. 9.342827
                                                      3.9804099 .. 2.4529374
2100395329
              3911.59 ...
                ... ...
2102591133
              3621.42 ...
                           2.9745781 .. -0.15291205
                                                       8.845858 .. 2.1556869
                               6.340416 .. 43.41358
                                                        4.712274 .. 6.132177
              3798.57 ...
2102591135
2102591136
              4496.66 ... 0.41709733 .. -0.05633008
                                                      4.158766 .. 0.99008995
                           -0.2628514 .. 0.08616114 5.2677093 .. 1.1210852
2102591137
              3960.42 ...
                           -0.28486162 .. 0.3934254 3.0511537 .. 0.97738653
2102591139
               4664.4 ...
              4358.55 ... 0.25720462 ... -0.06428938
                                                       4.398488 .. 1.6613194
2102591140
                           0.8538195 .. 0.037515007 4.3777156 .. 1.1644461
2102591141
              4198.65 ...
                                                       3.8486855 .. 0.986437
2102591142
               5337.0 ... -0.72364336 .. 0.034434147
2102591144
              4173.24 ...
                              1.5328524 .. 24.75373 4.7959304 .. 4.3725624
              3507.24 ...
                             2.864891 .. 0.13571836
                                                       7.040993 .. 1.7914823
2102591145
```

[]: # Once you have selected sources within the NEP footprint we can then go ahead

→ and find some

spectra from these sources - Oscar

spectra = detects_in_NEP.hdfile.root.Spectra

center_ksy.separations(skycoords entire) return indeces return distance return 3d separate by dist mask

main goal of algorithm want it to distinguish lae vs o2 emitter. wouldn't impose cuts unless training.

 $\mathrm{cut} = \mathrm{filter} \ \mathrm{cut} = \mathrm{signal} \ \mathrm{to} \ \mathrm{noise} \ \mathrm{could} \ \mathrm{do} \ \mathrm{plya}$

might need cuts for taining for confident lya and o2

increase confidence by visually inspecting

could visually inspect to increase confidence.

plotting histograms to look for outliers

hetdex isn't perfect and it catches emission lines that aren't real. visual inspections helps

no need for dataframes if i found another way

save as csv with astropy table

csv into get spec()

has nice documentation

get_spectrum good for ids ****USE*** returns all fiber spec with corresponding weights.

try to see if can get LAE samples. 02 samples. and ambigious samples. Clasify some of them. Signetection object filter by fields, turn to astropy table and then filter.

Want psf weighted.

[]:[