Using_PyMySQL

November 25, 2015

1 J Using PyMySQL to access MySQL databases

This package contains a pure-Python MySQL client library. In this sense, it does not need to have access to mysql header or library, which is the case for the mysqldb package. The goal of PyMySQL is to be a drop-in replacement for MySQLdb and work on CPython, PyPy, IronPython and Jython.

It is installed with "pip install pymysql"

We first import the usual libraries

```
In [1]: %matplotlib inline
        import numpy as np
        import matplotlib.pyplot as plt
```

This is the import of the library used to connect to MySQl database

```
In [2]: import pymysql
```

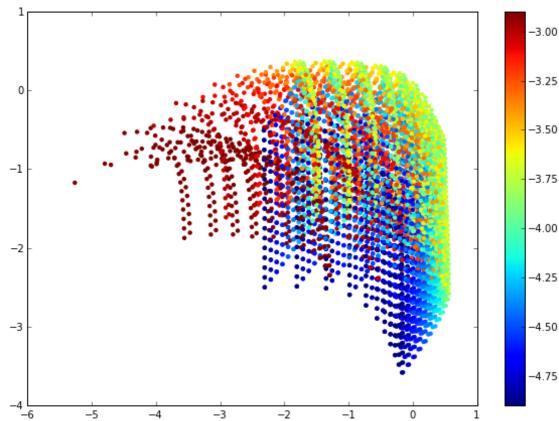
First you need to connect to a database. In our example, we will use the 3MdB database, which needs a password. https://sites.google.com/site/mexicanmillionmodels/

1.0.1 Connect to the database

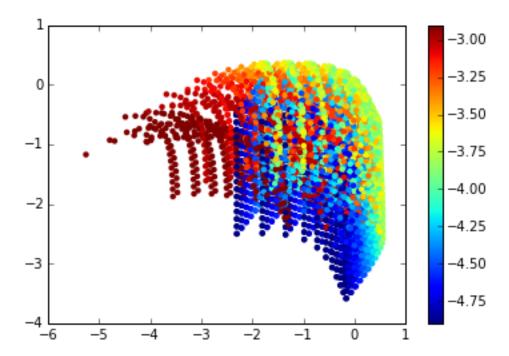
```
Out[20]: ((u'N1', 8, None, 20, 20, 0, False),
          (u'label', 253, None, 15, 15, 0, True),
          (u'id', 253, None, 20, 20, 0, True),
          (u'lambda', 5, None, 22, 22, 31, True),
          (u'name', 253, None, 40, 40, 0, False),
          (u'used', 3, None, 2, 2, 0, True))
In [21]: # fech all the resulting data into a variable
         lines = cur.fetchall()
In [22]: # close the cursor once used
         cur.close()
In [23]: # the result is in a form of tuple of tuples
         print lines
((1, 'BAC___3646A', 'Bac', 3646.0, 'BalmHead', 1), (2, 'COUT__3646A', 'cout', 3646.0, 'OutwardBalmPeak',
In [24]: # Each element of the first level tuple is a tuple corresponding to a row of the query results
         print len(lines)
         print lines[0]
(1, 'BAC___3646A', 'Bac', 3646.0, 'BalmHead', 1)
1.0.3 Using a cursor that returns a dictionary
In [25]: cur_dic = connector.cursor(pymysql.cursors.DictCursor)
In [26]: cur_dic.execute('select * from 'lines' limit 15')
Out[26]: 15
In [27]: lines_dic = cur_dic.fetchall()
In [28]: print lines_dic
[{u'used': 1, u'Nl': 1, u'name': 'BalmHead', u'label': 'BAC__3646A', u'id': 'Bac', u'lambda': 3646.0},
In [29]: # Each element of the table is a dictionary corresponding to a row od the guery results
         print lines_dic[0]
{u'used': 1, u'Nl': 1, u'name': 'BalmHead', u'label': 'BAC__3646A', u'id': 'Bac ', u'lambda': 3646.0}
In [30]: # One can easily create a new dictionary than hold the data in columns, better for plotting.
         new_dic = {k:np.array([d[k] for d in lines_dic]) for k in lines_dic[0].keys()}
In [31]: # The names of the columns are the names use in the database
         new_dic['lambda']
Out[31]: array([ 3.64600000e+03,
                                    3.64600000e+03,
                                                      3.64600000e+03,
                  4.86100000e+03,
                                    4.86100000e+03,
                                                      6.56300000e+03,
                  4.34000000e+03,
                                    4.10200000e+03,
                                                      3.97000000e+03,
                  3.83500000e+03,
                                    1.21600000e+03,
                                                      4.05100000e+00,
                                    7.45800000e+00,
                  2.62500000e+00,
                                                      5.87600000e+03])
In [32]: # One can also transform the results into a numpy recarray.
         # First step: create a table from the dictionnary
         lines_tab = [e.values() for e in lines_dic]
         lines_tab
```

```
Out[32]: [[1, 1, 'BalmHead', 'BAC___3646A', 'Bac', 3646.0],
         [1, 2, 'OutwardBalmPeak', 'COUT_3646A', 'cout', 3646.0],
         [1, 3, 'ReflectedBalmPeak', 'CREF_3646A', 'cref', 3646.0],
         [1, 4, 'H I 4861', 'H_1_4861A', 'H 1', 4861.0],
         [1, 5, 'H I 4861', 'TOTL_4861A', 'TOTL', 4861.0],
         [1, 6, 'H I 6563', 'H_1_6563A', 'H 1', 6563.0],
         [1, 7, 'H I 4340', 'H_1_4340A', 'H 1', 4340.0],
         [1, 8, 'H I 4102', 'H_1_4102A', 'H 1', 4102.0],
         [1, 9, 'H I 3970', 'H_1_3970A', 'H 1', 3970.0],
         [1, 10, 'H I 3835', 'H_1_3835A', 'H 1', 3835.0],
         [1, 11, 'H I 1216', 'H_1_1216A', 'H 1', 1216.0],
         [1, 12, 'H I 4.051m', 'H_1_4051M', 'H 1', 4.051],
         [1, 13, 'H I 2.625m', 'H_{-1}_{-2}625M', 'H 1', 2.625],
         [1, 14, 'H I 7.458m', 'H_1_7458M', 'H 1', 7.458],
         [1, 15, 'He I 5876', 'HE_1_5876A', 'He 1', 5876.0]]
In [33]: # Second step: transform the table into a numpy recarray, using the names from the dictionnary
        res = np.rec.fromrecords(lines_tab, names = lines_dic[0].keys())
In [34]: res
Out[34]: rec.array([(1, 1, 'BalmHead', 'BAC___3646A', 'Bac', 3646.0),
               (1, 2, 'OutwardBalmPeak', 'COUT_3646A', 'cout', 3646.0),
               (1, 3, 'ReflectedBalmPeak', 'CREF_3646A', 'cref', 3646.0),
               (1, 4, 'H I 4861', 'H_1_4861A', 'H 1', 4861.0),
               (1, 5, 'H I 4861', 'TOTL_4861A', 'TOTL', 4861.0),
               (1, 6, 'H I 6563', 'H_1_6563A', 'H 1', 6563.0),
               (1, 7, 'H I 4340', 'H_1_4340A', 'H 1', 4340.0),
               (1, 8, 'H I 4102', 'H<sub>-1</sub>-4102A', 'H 1', 4102.0),
               (1, 9, 'H I 3970', 'H<sub>-1</sub>-3970A', 'H 1', 3970.0),
               (1, 10, 'H I 3835', 'H_1_3835A', 'H 1', 3835.0),
               (1, 11, 'H I 1216', 'H_1_1_1216A', 'H 1', 1216.0),
               (1, 12, 'H I 4.051m', 'H_1_4051M', 'H 1', 4.051),
               (1, 13, 'H I 2.625m', 'H_1_2625M', 'H 1', 2.625),
               (1, 14, 'H I 7.458m', 'H_1_7458M', 'H 1', 7.458),
               (1, 15, 'He I 5876', 'HE_1_5876A', 'He 1', 5876.0)],
              dtype=[(u'used', '<i8'), (u'Nl', '<i8'), (u'name', 'S17'), (u'label', 'S11'), (u'id', 'S
In [35]: res['lambda']
Out[35]: array([ 3.64600000e+03,
                                  3.64600000e+03,
                                                    3.64600000e+03,
                 4.86100000e+03,
                                  4.86100000e+03,
                                                    6.56300000e+03,
                 4.34000000e+03,
                                                    3.97000000e+03,
                                  4.10200000e+03,
                 3.83500000e+03,
                                  1.21600000e+03,
                                                    4.05100000e+00,
                                  7.45800000e+00, 5.87600000e+03])
                 2.62500000e+00,
1.0.4 Example of plotting the result of a query
In [36]: # Send the query
        In [25]: print N
7854
In [37]: # obtain the results as a dictionnary
        res = cur_dic.fetchall()
```

```
In [38]: # transform the disctionary into a recarray
         data = np.rec.fromrecords([e.values() for e in res], names = res[0].keys())
In [39]: # check the data
         data
Out[39]: rec.array([(1.13306243836e+58, 8.465943086e+58, -3.1, 3.15741653467e+58),
                (3.42011987292e+59, 3.82678097448e+59, -4.7, 1.96658128904e+58),
                (1.9919317079e+55, 2.95364632532e+58, -2.9, 8.79993595982e+57), \ldots,
                (1.75269190656e+60, 5.79356475056e+59, -3.7, 5.08981089096e+58),
                (1.37202884837e+60, 5.15976659165e+59, -4.1, 3.20261785304e+57),
                (1.52244147812e+60, 5.27404255136e+59, -4.0, 3.89222406128e+58)],
               dtype=[(u'0_3_5007A', '<f8'), (u'H_1_6563A', '<f8'), (u'oxygen', '<f8'), (u'N_2_6584A'
In [40]: data['0_3_5007A']
Out[40]: array([ 1.13306244e+58,
                                   3.42011987e+59, 1.99193171e+55, ...,
                 1.75269191e+60,
                                   1.37202885e+60,
                                                     1.52244148e+60])
In [42]: \# Plot the results, using a column as color code
         fig, ax = plt.subplots(figsize=(10,7))
         scat = ax.scatter(np.log10(data['0_3_5007A'] / data['H_1_6563A']), np.log10(data['N_2_650])
                     c=data['oxygen'], edgecolor='none')
         fig.colorbar(scat)
Out[42]: <matplotlib.colorbar.Colorbar instance at 0x108b6f4d0>
```



```
In [43]: # Disconnect cursor and connector
         cur dic.close()
         connector.close()
1.0.5 Easier way using pyCloudy library
In [44]: # Import pyCloudy
         import pyCloudy as pc
         # pyCloudy version must be > 0.8.43
         print pc.__version__
0.8.57ъ
In [45]: pc.config.db_connector = 'PyMySQL'
         # Define the parameters of the connection in a dictionnary
         OVN_dic= {'host' : '132.248.1.102',
                   'user_name' : 'OVN_user',
                   'user_passwd' : '***',
                   'base_name' : '3MdB'}
         # Instantiate an object that will deal with the database connections and queries
         db = pc.MdB(OVN_dic)
In [46]: res, N = db.exec_dB('select ref, count(*) from tab group by ref')
         print res
         print N
[{u'count(*)': 85800, u'ref': 'CALIFA'}, {u'count(*)': 20793, u'ref': 'CALIFA_ah'}, {u'count(*)': 41327,
In [47]: # Obtain the result of a select command directly as a recarray
         data, N = db.select_dB(select_='0__3_5007A, N_2_6584A, H_1_6563A, oxygen', from_='tab', w
                             limit_=None, format_='rec')
In [48]: # Check the data
         data
Out [48]: rec.array([(1.13306243836e+58, 8.465943086e+58, -3.1, 3.15741653467e+58),
                (3.42011987292e+59, 3.82678097448e+59, -4.7, 1.96658128904e+58),
                (1.9919317079e+55, 2.95364632532e+58, -2.9, 8.79993595982e+57), \ldots,
                (1.75269190656e+60, 5.79356475056e+59, -3.7, 5.08981089096e+58),
                (1.37202884837e+60, 5.15976659165e+59, -4.1, 3.20261785304e+57),
                (1.52244147812e+60, 5.27404255136e+59, -4.0, 3.89222406128e+58)],
               dtype=[(u'0_3_5007A', '<f8'), (u'H_1_6563A', '<f8'), (u'oxygen', '<f8'), (u'N_2_6584A'
In [47]: # Make the same plot
         fig, ax = plt.subplots()
         scat = ax.scatter(np.log10(data['0_3_5007A'] / data['H_1_6563A']), np.log10(data['N_2_6563A'])
                     c=data['oxygen'], edgecolor='none')
         cb = fig.colorbar(scat)
```



<class 'pandas.core.frame.DataFrame'>

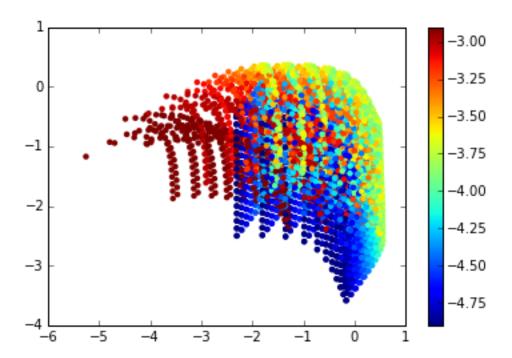
```
Out[54]:
               O__3_5007A
                            N_{-2}_{-6584A}
                                         H_1_6563A oxygen
        0
              1.133062e+58 3.157417e+58 8.465943e+58
                                                          -3.1
                                                          -4.7
        1
              3.420120e+59 1.966581e+58 3.826781e+59
                                                          -2.9
        2
              1.991932e+55 8.799936e+57
                                          2.953646e+58
        3
              1.094455e+59 1.517612e+57
                                          7.586338e+58
                                                          -3.9
        4
              1.300403e+56 9.594305e+56 1.593603e+58
                                                          -3.0
        5
              4.547320e+59 6.234337e+56 2.780188e+59
                                                          -4.3
        6
              3.482729e+59 4.703550e+56 3.828982e+59
                                                          -4.7
        7
              4.364881e+58 1.791885e+58
                                                          -4.5
                                          7.633109e+58
        8
              7.768910e+58 6.520140e+57
                                          1.388706e+59
                                                          -3.2
                                                          -4.0
        9
              6.521827e+59 3.655320e+58
                                          2.850901e+59
              5.827025e+58 2.315942e+57
                                          7.668195e+58
                                                          -4.4
        10
        11
              9.819971e+58 3.263721e+58
                                          1.369624e+59
                                                          -3.3
        12
              1.216272e+58 1.322807e+58
                                         4.584762e+58
                                                          -4.7
        13
              4.319418e+59 2.171798e+56 4.962521e+59
                                                          -4.8
                                                          -3.2
        14
              8.732360e+59 6.007494e+58
                                          7.277213e+59
        15
              1.948541e+59 9.667164e+57
                                          1.312013e+59
                                                          -3.5
        16
              1.218185e+59 3.149880e+57
                                          5.708760e+59
                                                          -3.0
        17
              1.943526e+58 3.791206e+57
                                          4.586533e+58
                                                          -4.5
              4.011769e+55 3.750527e+57 4.869870e+57
                                                          -3.3
        18
```

```
20
                                                         -3.6
              7.564738e+59 3.233375e+57
                                         3.091887e+59
        21
              8.823437e+57 2.065070e+56 4.700258e+58
                                                         -4.9
              3.226162e+57 4.962798e+58
                                                         -3.1
        22
                                         8.706764e+58
        23
              2.365680e+55 1.141878e+57
                                         4.890240e+57
                                                         -4.9
        24
              8.174137e+59 1.582741e+57
                                         3.000036e+59
                                                         -3.7
        25
              1.150354e+60 1.139883e+59 4.314785e+59
                                                         -3.7
        26
              2.211890e+56 5.419716e+57
                                         5.111543e+58
                                                         -2.9
        27
              7.803436e+59 1.548125e+58
                                         3.012500e+59
                                                         -3.7
        28
              5.340369e+58 1.015961e+58
                                         1.260066e+59
                                                         -4.8
        29
              1.435415e+56 3.100620e+58
                                         1.592877e+59
                                                         -2.9
         . . .
                                                          . . .
        7824 1.972405e+60 8.609873e+57
                                         5.901718e+59
                                                         -3.6
        7825
              4.310580e+57 1.923966e+56
                                         2.697913e+58
                                                         -4.7
        7826 1.008846e+59
                                                         -4.0
                            2.983662e+57
                                         7.566729e+58
        7827
              4.018091e+59
                            9.599803e+58
                                         7.998521e+59
                                                         -3.1
        7828 9.895354e+57 1.213889e+59
                                                         -3.0
                                         6.651922e+59
        7829 8.789083e+59
                            6.046069e+56
                                         3.819032e+59
                                                         -4.2
        7830 2.937330e+57
                            2.126139e+57
                                         1.492873e+59
                                                         -2.9
        7831 9.264124e+58
                            2.688380e+57
                                         7.855083e+58
                                                         -3.5
        7832 2.407456e+58 2.687464e+56 4.575668e+58
                                                         -4.4
        7833 8.588467e+58 7.640683e+58 7.745671e+58
                                                         -3.7
        7834 1.322239e+55 9.198548e+57
                                         2.967463e+58
                                                         -2.9
        7835
              2.374835e+57 1.561329e+58 1.500132e+58
                                                         -3.5
        7836 2.630657e+59 5.677059e+57
                                         2.792541e+59
                                                         -4.6
        7837 8.598621e+59 9.858896e+58
                                        4.227971e+59
                                                         -3.9
        7838 1.459214e+59 4.216686e+57
                                                         -4.3
                                         1.234976e+59
        7839 2.409456e+55 6.488206e+56 4.905731e+57
                                                         -4.9
        7840 2.478336e+59 2.729407e+57
                                         1.253125e+59
                                                         -3.8
        7841 3.159983e+58 1.275284e+58 8.214903e+58
                                                         -3.2
        7842 9.530954e+57 1.277263e+58
                                         2.599781e+58
                                                         -4.2
        7843 9.218411e+58 3.574314e+57 7.856340e+58
                                                         -3.5
        7844 3.804036e+59 2.093539e+58
                                         2.776236e+59
                                                         -4.4
        7845 5.177660e+56
                            2.514500e+58
                                         1.553671e+58
                                                         -3.3
        7846
              1.956973e+58
                            2.850425e+57
                                         4.590933e+58
                                                         -4.5
        7847 8.346006e+55 1.451319e+57 4.784272e+57
                                                         -3.4
        7848 6.879302e+59 3.689720e+57 3.181142e+59
                                                         -3.5
        7849 3.386232e+59 8.655237e+58 3.477404e+59
                                                         -3.3
                                                         -3.6
        7850 5.123895e+56 1.386879e+58
                                         8.461529e+57
        7851 1.752692e+60 5.089811e+58 5.793565e+59
                                                         -3.7
                                                         -4.1
        7852 1.372029e+60 3.202618e+57 5.159767e+59
        7853 1.522441e+60 3.892224e+58 5.274043e+59
                                                         -4.0
        [7854 rows x 4 columns]
In [50]: # Make the same plot
        fig, ax = plt.subplots()
        scat = ax.scatter(np.log10(data['0_3_5007A'] / data['H_1_6563A']), np.log10(data['N_2_6563A'])
                    c=data['oxygen'], edgecolor='none')
        cb = fig.colorbar(scat)
```

-4.2

19

1.677233e+59 2.198075e+57 1.232347e+59



In [43]: db.close_dB()

1.0.6 Using pyCloudy to save the result in a file

In [56]: from pyCloudy.db.MdB import MdB_subproc

In [57]: db = MdB_subproc(OVN_dic)

db.close_dB()

O35007A	N_{-2}_{-6584A}	$H_{-}1_{-}6563A$	oxygen	
1.13306243836e58	3.157416	53467e58	8.465943086e58	-3.1
3.42011987292e59	1.966581	28904e58	3.82678097448e59	-4.7
1.9919317079e55	8.7999359	5982e57	2.95364632532e58	-2.9
1.09445528168e59	1.517612	18935e57	7.58633813601e58	-3.9
1.3004028118e56	9.5943049	8831e56	1.59360285671e58	-3
4.54731969943e59	6.234336	65474e56	2.780187567e59	-4.3
3.48272851916e59	4.703549	86736e56	3.82898210273e59	-4.7
4.36488135054e58	1.791885	30745e58	7.63310885003e58	-4.5
7.76890989905e58	6.520139	85859e57	1.38870636951e59	-3.2

In [62]: data = np.genfromtxt('query_res.dat', names=True, dtype=None)

In [63]: data

```
Out[63]: array([(1.13306243836e+58, 3.15741653467e+58, 8.465943086e+58, -3.1),
                (3.42011987292e+59, 1.96658128904e+58, 3.82678097448e+59, -4.7),
                (1.9919317079e+55, 8.79993595982e+57, 2.95364632532e+58, -2.9), \ldots,
                (1.75269190656e+60, 5.08981089096e+58, 5.79356475056e+59, -3.7),
                (1.37202884837e+60, 3.20261785304e+57, 5.15976659165e+59, -4.1),
                (1.52244147812e+60, 3.89222406128e+58, 5.27404255136e+59, -4.0)],
               dtype=[('0_3_5007A', '<f8'), ('N_2_6584A', '<f8'), ('H_1_6563A', '<f8'), ('oxygen', '<
In [64]: # Make the same plot
         fig, ax = plt.subplots()
         scat = ax.scatter(np.log10(data['0_3_5007A'] / data['H_1_6563A']), np.log10(data['N_2_650])
                     c=data['oxygen'], edgecolor='none')
         cb = fig.colorbar(scat)
                                                                         -3.00
                                                                         -3.25
             0
                                                                         -3.50
            -1
                                                                         -3.75
                                                                         -4.00
            -2
                                                                         -4.25
            -3
                                                                         -4.50
                                                                         -4.75
```

-2

-1

0

In []: