

```
In [365]: """
HOMEWORK 5 SOLUTION CREATED BY
OLABODE ALAMU
1498663
GUIDE TO ENGINEERING DATA SCIENCE
"""
```

```
Out[365]: '\nHOMEWORK 5 SOLUTION CREATED BY\nOLABODE ALAMU\n1498663\nGUIDE TO ENGINEERING DATA SCIENCE\n\n'
```

```
In [212]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [213]: # Import the dataset
#head_column = ['Date', 'Q-E', 'ZN-E', 'PH-E', 'DBO-E', 'DQO-E', 'SS-E', 'SSV-E', 'SED-E', 'COND-E', 'PH-P', 'DBO-P', 'SS-P', 'SSV-P', 'SED-P', 'COND-P', 'PH-D', 'DBO-D', 'DQO-D', 'SS-D', 'SSV-D', 'SED-D', 'COND-D', 'PH-S', 'DBO-S', 'DQO-S', 'SS-S', 'SSV-S', 'SED-S', 'COND-S', 'RD-DBO-P', 'RD-SS-P', 'RD-SED-P', 'RD-DBO-S', 'RD-DQO-S', 'RD-DBO-G', 'RD-DQO-G', 'RD-SS-G', 'RD-SED-G']

Water = pd.read_csv('water-treatment.data', header = None, na_values = '?', index_col = 0)
```

```
In [324]: # Prints out the top 5 rows of the dataset
print(Water.head())
```

	1	2	3	4	5	6	7	8	9	10	...
\											
0											...
D-1/3/90	44101.0	1.5	7.8	NaN	407.0	166.0	66.3	4.5	2110	7.9	...
D-2/3/90	39024.0	3.0	7.7	NaN	443.0	214.0	69.2	6.5	2660	7.7	...
D-4/3/90	32229.0	5.0	7.6	NaN	528.0	186.0	69.9	3.4	1666	7.7	...
D-5/3/90	35023.0	3.5	7.9	205.0	588.0	192.0	65.6	4.5	2430	7.8	...
D-6/3/90	36924.0	1.5	8.0	242.0	496.0	176.0	64.8	4.0	2110	7.9	...
	29	30	31	32	33	34	35	36	37	38	
0											
D-1/3/90	2000.0	NaN	58.8	95.5	NaN	70.0	NaN	79.4	87.3	99.6	
D-2/3/90	2590.0	NaN	60.7	94.8	NaN	80.8	NaN	79.5	92.1	100.0	
D-4/3/90	1888.0	NaN	58.2	95.6	NaN	52.9	NaN	75.8	88.7	98.5	
D-5/3/90	1840.0	33.1	64.2	95.3	87.3	72.3	90.2	82.3	89.6	100.0	
D-6/3/90	2120.0	NaN	62.7	95.6	NaN	71.0	92.1	78.2	87.5	99.5	

[5 rows x 38 columns]

```
In [325]: print(Water.shape) # Shows the shape of the dataframe  
(527, 38)
```

```
In [215]: X = Water.columns # This shows all the columns present in the dataset
```

```
In [216]: print(X)  
Int64Index([ 1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12, 13, 14, 15, 16, 1  
7,  
            18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 3  
4,  
            35, 36, 37, 38],  
           dtype='int64')
```

```
In [326]: # This for loop was created to iterate through the difference columns of the dataset, split out each column  
# and calculate the different statistical parameters for each column and print them out  
for column in Water.columns:  
    Maximum = Water[i].max()  
    Minimum = Water[i].min()  
    Mean_value = Water[i].mean()  
    Median_value = Water[i].median()  
    Standard_dev = Water[i].std()  
    print('Column ',i,'\nMaximum value is ', Maximum, '\nMinimum value is ', Minimum,  
          '\nMean value is ', Mean_value, '\nMedian is ', Median_value,  
          '\nStandard deviation is ', Standard_dev)
```

[illegible]

[illegible]

[illegible]

[illegible]

```
In [327]: # Computes the number of missing values in each column and prints out a statement
Missing_values = Water.isnull().sum()
for column in Water.columns: # iterates through each column
    Missing_value = Missing_values[column]
    print('Column ',column, ' has ',Missing_value, ' number of missing values.')
```

```
Column 1 has 18 number of missing values.
Column 2 has 3 number of missing values.
Column 3 has 0 number of missing values.
Column 4 has 23 number of missing values.
Column 5 has 6 number of missing values.
Column 6 has 1 number of missing values.
Column 7 has 11 number of missing values.
Column 8 has 25 number of missing values.
Column 9 has 0 number of missing values.
Column 10 has 0 number of missing values.
Column 11 has 40 number of missing values.
Column 12 has 0 number of missing values.
Column 13 has 11 number of missing values.
Column 14 has 24 number of missing values.
Column 15 has 0 number of missing values.
Column 16 has 0 number of missing values.
Column 17 has 28 number of missing values.
Column 18 has 9 number of missing values.
Column 19 has 2 number of missing values.
Column 20 has 13 number of missing values.
Column 21 has 25 number of missing values.
Column 22 has 0 number of missing values.
Column 23 has 1 number of missing values.
Column 24 has 23 number of missing values.
Column 25 has 18 number of missing values.
Column 26 has 5 number of missing values.
Column 27 has 17 number of missing values.
Column 28 has 28 number of missing values.
Column 29 has 1 number of missing values.
Column 30 has 62 number of missing values.
Column 31 has 4 number of missing values.
Column 32 has 27 number of missing values.
Column 33 has 40 number of missing values.
Column 34 has 26 number of missing values.
Column 35 has 36 number of missing values.
Column 36 has 25 number of missing values.
Column 37 has 8 number of missing values.
Column 38 has 31 number of missing values.
```



```
In [328]: # Import missing value using the median
# Step 1: Computes the medians across all the columns and passes them into a list
Value_list = [] # Creates an empty list

for column in Water.columns:
    Median_value = Water[column].median()
    Value_list.append(Median_value)
print(Value_list)

# Step 2 Transform the list of median values into a dictionary where the column number is the key and the median
# values are the value in the key: value pair of the dictionary
Median_dict = {}
for column in Water.columns:
    Median_dict[column]= Value_list[column-1]

print(Median_dict)

# Step 3 Fill the missing value with the medians in the dictionary

Water_cleaned = Water.fillna(value = Median_dict)

print(Water_cleaned)
```

```

[35990.0, 1.5, 7.8, 182.5, 400.0, 196.0, 64.3, 4.5, 1406.0, 7.8, 197.0, 220.
0, 62.8, 4.5, 1420.0, 7.8, 119.0, 274.0, 90.0, 74.5, 0.3, 1428.0, 7.7, 18.0,
84.0, 19.0, 81.15, 0.01, 1432.0, 39.6, 59.5, 93.3, 85.4, 69.9, 90.2, 79.15,
90.7, 99.7]
{1: 35990.0, 2: 1.5, 3: 7.8, 4: 182.5, 5: 400.0, 6: 196.0, 7: 64.3, 8: 4.5,
9: 1406.0, 10: 7.8, 11: 197.0, 12: 220.0, 13: 62.8, 14: 4.5, 15: 1420.0, 16:
7.8, 17: 119.0, 18: 274.0, 19: 90.0, 20: 74.5, 21: 0.3, 22: 1428.0, 23: 7.7,
24: 18.0, 25: 84.0, 26: 19.0, 27: 81.15, 28: 0.01, 29: 1432.0, 30: 39.6, 31:
59.5, 32: 93.3, 33: 85.4, 34: 69.9, 35: 90.2, 36: 79.15, 37: 90.7, 38: 99.7}
      1      2      3      4      5      6      7      8      9      10 \
0
D-1/3/90    44101.0  1.50  7.8   182.5  407.0  166.0  66.3  4.5  2110  7.9
D-2/3/90    39024.0  3.00  7.7   182.5  443.0  214.0  69.2  6.5  2660  7.7
D-4/3/90    32229.0  5.00  7.6   182.5  528.0  186.0  69.9  3.4  1666  7.7
D-5/3/90    35023.0  3.50  7.9   205.0  588.0  192.0  65.6  4.5  2430  7.8
D-6/3/90    36924.0  1.50  8.0   242.0  496.0  176.0  64.8  4.0  2110  7.9
D-7/3/90    38572.0  3.00  7.8   202.0  372.0  186.0  68.8  4.5  1644  7.8
D-8/3/90    41115.0  6.00  7.8   182.5  552.0  262.0  64.1  5.0  1603  7.8
D-9/3/90    36107.0  5.00  7.7   215.0  489.0  334.0  40.7  6.0  1613  7.6
D-11/3/90   29156.0  2.50  7.7   206.0  451.0  194.0  69.1  4.5  1249  7.7
D-12/3/90   39246.0  2.00  7.8   172.0  506.0  200.0  69.0  5.0  1865  7.8
D-13/3/90   42393.0  0.70  7.9   189.0  478.0  230.0  67.0  5.5  1410  8.1
D-14/3/90   42857.0  1.50  7.7   238.0  319.0  292.0  33.8  3.5  1261  7.6
D-15/3/90   42911.0  0.70  7.6   114.0  252.0  116.0  58.6  1.2  1238  7.9
D-16/3/90   40376.0  1.50  8.1   204.0  333.0  174.0  67.8  3.0  2390  7.8
D-18/3/90   40923.0  3.50  7.6   146.0  329.0  188.0  57.4  2.5  1300  7.6
D-19/3/90   43830.0  1.50  7.8   177.0  512.0  214.0  58.9  5.5  1605  7.7
D-20/3/90   39165.0  1.20  7.4   250.0  447.0  252.0  61.1  7.0  1533  7.4
D-21/3/90   35791.0  1.20  7.8   277.0  466.0  246.0  63.4  4.0  1556  7.7
D-22/3/90   37419.0  1.20  7.6   219.0  446.0  222.0  61.3  5.5  1600  7.7
D-23/3/90   40983.0  3.00  7.6   182.0  431.0  214.0  57.0  7.0  1591  7.5
D-25/3/90   42217.0  8.50  7.5   138.0  333.0  240.0  55.0  3.8  1087  7.5
D-26/3/90   47665.0  1.20  7.7   156.0  405.0  200.0  74.0  4.0  1856  7.6
D-27/3/90   44314.0  3.00  7.8   155.0  389.0  308.0  49.4  6.0  1927  7.7
D-28/3/90   40841.0  1.00  7.6   179.0  389.0  168.0  69.0  3.5  1240  7.8
D-29/3/90   41157.0  3.00  8.0   145.0  398.0  192.0  66.7  4.5  2240  8.0
D-30/3/90   40078.0  1.40  7.9   198.0  464.0  228.0  64.9  4.6  1431  7.6
D-1/2/90    44365.0  7.50  7.9   182.5  365.0  212.0  62.3  3.5  1339  7.9
D-2/2/90    43080.0  4.25  7.8    95.0  349.0  136.0  76.5  2.5  1063  7.8
D-4/2/90    29414.0  3.00  7.6   160.0  374.0  168.0  69.0  3.1  1042  7.6
D-5/2/90    37312.0  1.00  8.1   205.0  492.0  192.0  70.8  4.0  1454  8.1
...
D-25/10/91  35400.0  0.70  7.6   156.0  364.0  194.0  63.9  5.5  1680  7.6
D-26/10/91  30964.0  3.30  7.7   220.0  540.0  184.0  62.0  3.5  1445  7.7
D-27/10/91  35573.0  7.30  7.6   176.0  333.0  178.0  64.0  3.5  1627  7.7
D-29/10/91  29801.0  1.60  7.7   172.0  400.0  136.0  70.1  1.5  1402  7.7
D-30/10/91  31524.0  1.60  7.9   182.5  478.0  204.0  64.7  6.0  1798  7.9
D-1/8/91    29834.0  3.00  7.4   160.0  348.0  194.0  61.9  3.0  1720  7.5
D-2/8/91    28492.0  2.60  7.5   124.0  281.0  172.0  66.3  3.0  1520  7.5
D-4/8/91    24978.0  0.50  7.3   146.0  288.0  124.0  67.7  2.0  1210  7.4
D-5/8/91    29719.0  0.20  7.6   133.0  284.0  186.0  71.0  5.0  1114  7.6
D-6/8/91    29741.0  0.45  7.9   151.0  316.0  196.0  64.3  2.5   948  7.8
D-7/8/91    29027.0  0.40  7.6   136.0  328.0  186.0  67.7  3.0   899  7.6
D-8/8/91    30211.0  0.50  7.6   114.0  521.0  506.0  44.3  7.5   866  7.5
D-9/8/91    30848.0  0.20  7.7   142.0  376.0  144.0  70.8  3.0   940  7.6
D-11/8/91   17527.0  0.55  7.5   150.0  171.0  172.0  37.2  1.4   732  7.5
D-12/8/91   33331.0  0.23  7.6    92.0  233.0  234.0  37.6  1.4   829  7.6

```

D-13/8/91	27998.0	0.62	7.5	138.0	268.0	154.0	66.2	1.7	890	7.5
D-14/8/91	32845.0	0.23	7.6	84.0	251.0	98.0	71.4	2.0	866	7.6
D-16/8/91	27933.0	0.20	7.6	158.0	375.0	178.0	60.7	3.5	1049	7.7
D-18/8/91	27527.0	0.20	7.3	191.0	240.0	166.0	73.5	3.0	1072	7.4
D-19/8/91	32363.0	0.10	7.6	159.0	310.0	146.0	68.5	1.6	1096	7.6
D-20/8/91	31437.0	0.47	7.6	132.0	304.0	148.0	64.9	2.0	939	7.7
D-21/8/91	31914.0	2.00	7.7	127.0	274.0	144.0	72.2	2.0	1031	7.6
D-22/8/91	28088.0	0.20	7.5	153.0	307.0	124.0	82.3	2.5	1044	7.6
D-23/8/91	27838.0	0.13	7.6	179.0	265.0	128.0	71.9	1.8	992	7.6
D-25/8/91	29271.0	0.36	7.5	99.0	585.0	140.0	71.4	4.5	962	7.6
D-26/8/91	32723.0	0.16	7.7	93.0	252.0	176.0	56.8	2.3	894	7.7
D-27/8/91	33535.0	0.32	7.8	192.0	346.0	172.0	68.6	4.0	988	7.8
D-28/8/91	32922.0	0.30	7.4	139.0	367.0	180.0	64.4	3.0	1060	7.5
D-29/8/91	32190.0	0.30	7.3	200.0	545.0	258.0	65.1	4.0	1260	7.4
D-30/8/91	30488.0	0.21	7.5	152.0	300.0	132.0	69.7	4.5	1073	7.4

	...	29	30	31	32	33	34	35	36	37 \
0	...									
D-1/3/90	...	2000.0	39.6	58.8	95.5	85.4	70.0	90.2	79.40	87.3
D-2/3/90	...	2590.0	39.6	60.7	94.8	85.4	80.8	90.2	79.50	92.1
D-4/3/90	...	1888.0	39.6	58.2	95.6	85.4	52.9	90.2	75.80	88.7
D-5/3/90	...	1840.0	33.1	64.2	95.3	87.3	72.3	90.2	82.30	89.6
D-6/3/90	...	2120.0	39.6	62.7	95.6	85.4	71.0	92.1	78.20	87.5
D-7/3/90	...	1764.0	39.6	59.7	96.5	86.7	78.3	90.1	73.10	84.9
D-8/3/90	...	1703.0	39.6	61.9	93.8	89.1	79.8	90.2	86.20	90.1
D-9/3/90	...	1606.0	39.6	70.4	95.6	90.6	53.7	92.1	66.90	94.6
D-11/3/90	...	1338.0	46.1	43.6	92.5	85.6	58.2	92.2	73.80	90.2
D-12/3/90	...	1616.0	21.2	59.7	90.8	88.4	66.1	89.0	69.00	86.5
D-13/3/90	...	1575.0	0.6	45.8	92.0	11.6	25.7	19.6	36.00	43.0
D-14/3/90	...	1304.0	31.8	61.2	92.9	85.4	69.9	90.2	79.15	18.5
D-15/3/90	...	1221.0	46.6	48.5	91.7	85.4	20.4	26.3	31.70	10.3
D-16/3/90	...	2550.0	41.1	50.0	84.0	76.5	52.9	84.3	54.10	43.7
D-18/3/90	...	1545.0	32.7	33.3	90.0	82.6	61.3	87.0	71.40	78.2
D-19/3/90	...	1110.0	28.0	72.7	92.7	78.8	36.6	85.9	60.40	90.7
D-20/3/90	...	1402.0	49.8	58.3	92.3	89.9	96.8	94.4	98.10	92.1
D-21/3/90	...	1606.0	39.6	39.6	78.3	85.5	70.4	91.3	73.40	89.4
D-22/3/90	...	1780.0	35.3	57.5	92.0	75.6	49.3	80.8	60.80	76.1
D-23/3/90	...	1597.0	20.1	64.5	93.6	86.9	68.1	87.4	72.20	88.3
D-25/3/90	...	1223.0	29.4	55.4	91.3	85.2	68.0	88.4	81.40	92.9
D-26/3/90	...	1706.0	28.1	50.0	91.4	85.2	76.5	87.8	82.50	88.5
D-27/3/90	...	1869.0	52.0	64.9	90.8	87.6	71.2	90.3	77.60	92.5
D-28/3/90	...	1416.0	26.7	66.2	95.0	89.2	72.5	91.1	78.70	88.1
D-29/3/90	...	2290.0	34.3	65.0	94.2	89.3	69.7	89.7	78.10	89.1
D-30/3/90	...	1475.0	27.2	67.6	97.3	90.4	68.0	91.4	78.00	90.4
D-1/2/90	...	1377.0	39.6	50.0	89.4	85.4	76.6	90.2	79.50	90.6
D-2/2/90	...	1220.0	6.8	47.9	80.0	84.6	78.9	80.0	80.80	82.4
D-4/2/90	...	1087.0	42.7	54.5	95.7	84.9	73.6	88.1	78.90	83.3
D-5/2/90	...	1275.0	39.6	33.0	94.5	85.3	73.7	84.4	76.80	80.7
...
D-25/10/91	...	1840.0	47.3	61.3	94.0	76.4	69.9	86.5	82.40	90.7
D-26/10/91	...	1337.0	39.6	38.6	93.3	85.4	87.0	92.7	95.00	91.8
D-27/10/91	...	1799.0	39.6	40.4	95.0	85.4	72.9	90.9	79.90	90.7
D-29/10/91	...	1468.0	32.4	40.4	88.0	87.8	77.6	91.3	85.30	83.8
D-30/10/91	...	1568.0	39.6	43.9	65.3	85.4	75.3	90.2	81.20	89.7
D-1/8/91	...	1772.0	29.1	53.5	92.0	85.7	64.2	90.6	72.70	85.6
D-2/8/91	...	1549.0	34.2	47.7	99.0	84.4	66.1	90.3	73.30	88.4
D-4/8/91	...	1259.0	39.6	52.3	92.3	75.3	69.9	86.3	79.15	76.6

D-5/8/91	...	1100.0	55.1	73.2	96.0	82.0	62.5	91.7	78.90	88.7
D-6/8/91	...	951.0	44.8	64.5	95.0	82.2	80.0	89.4	86.10	88.8
D-7/8/91	...	898.0	39.6	64.1	94.0	84.9	60.4	91.9	76.80	87.6
D-8/8/91	...	884.0	48.7	86.9	98.8	81.0	67.9	90.4	91.60	96.0
D-9/8/91	...	947.0	39.6	47.6	97.3	85.4	77.6	93.7	84.80	90.3
D-11/8/91	...	728.0	62.8	77.7	90.0	73.8	65.5	92.7	77.20	90.7
D-12/8/91	...	929.0	36.9	43.6	86.7	87.7	71.9	91.3	79.80	92.3
D-13/8/91	...	858.0	38.1	41.6	86.7	87.7	79.0	94.2	87.70	91.6
D-14/8/91	...	879.0	50.9	36.5	80.0	87.0	69.6	91.7	80.50	82.7
D-16/8/91	...	828.0	68.0	66.7	96.7	81.6	41.8	94.3	72.50	83.1
D-18/8/91	...	999.0	38.5	54.5	88.0	90.0	69.9	95.8	81.70	93.4
D-19/8/91	...	1083.0	25.2	61.4	91.2	78.6	65.1	86.8	81.00	89.0
D-20/8/91	...	1012.0	45.6	60.3	94.4	82.5	72.9	89.4	86.20	91.2
D-21/8/91	...	1053.0	39.6	59.5	93.3	85.4	77.7	90.2	79.15	90.7
D-22/8/91	...	1038.0	40.5	54.4	94.0	89.7	75.5	93.5	85.00	90.3
D-23/8/91	...	1044.0	13.7	45.0	95.0	87.5	71.3	93.9	79.60	89.1
D-25/8/91	...	968.0	40.8	71.6	93.3	59.0	26.4	74.7	83.80	81.4
D-26/8/91	...	942.0	39.6	62.3	93.3	69.8	75.9	79.6	78.60	96.6
D-27/8/91	...	950.0	39.6	58.3	97.8	83.0	59.1	91.1	74.60	90.7
D-28/8/91	...	1136.0	39.6	65.0	97.1	76.2	66.4	82.0	77.10	88.9
D-29/8/91	...	1326.0	39.8	65.9	97.1	81.7	70.9	89.5	87.00	89.5
D-30/8/91	...	1224.0	39.6	69.5	93.3	81.7	76.4	90.2	81.70	86.4

38

0

D-1/3/90	99.6
D-2/3/90	100.0
D-4/3/90	98.5
D-5/3/90	100.0
D-6/3/90	99.5
D-7/3/90	100.0
D-8/3/90	99.0
D-9/3/90	100.0
D-11/3/90	99.4
D-12/3/90	99.6
D-13/3/90	36.4
D-14/3/90	42.9
D-15/3/90	95.4
D-16/3/90	100.0
D-18/3/90	99.2
D-19/3/90	100.0
D-20/3/90	100.0
D-21/3/90	99.4
D-22/3/90	99.6
D-23/3/90	100.0
D-25/3/90	100.0
D-26/3/90	99.8
D-27/3/90	100.0
D-28/3/90	100.0
D-29/3/90	100.0
D-30/3/90	100.0
D-1/2/90	98.6
D-2/2/90	100.0
D-4/2/90	100.0
D-5/2/90	100.0
...	...
D-25/10/91	99.8

D-26/10/91	95.7
D-27/10/91	98.6
D-29/10/91	96.7
D-30/10/91	99.2
D-1/8/91	99.3
D-2/8/91	99.3
D-4/8/91	99.0
D-5/8/91	99.6
D-6/8/91	100.0
D-7/8/91	99.3
D-8/8/91	99.7
D-9/8/91	99.3
D-11/8/91	98.6
D-12/8/91	99.3
D-13/8/91	98.8
D-14/8/91	99.0
D-16/8/91	98.6
D-18/8/91	99.3
D-19/8/91	99.4
D-20/8/91	99.5
D-21/8/91	99.7
D-22/8/91	100.0
D-23/8/91	100.0
D-25/8/91	99.7
D-26/8/91	99.6
D-27/8/91	100.0
D-28/8/91	99.0
D-29/8/91	99.8
D-30/8/91	99.7

[527 rows x 38 columns]

In [329]: `print(Water_cleaned)`

	1	2	3	4	5	6	7	8	9	10	\
0											
D-1/3/90	44101.0	1.50	7.8	182.5	407.0	166.0	66.3	4.5	2110	7.9	
D-2/3/90	39024.0	3.00	7.7	182.5	443.0	214.0	69.2	6.5	2660	7.7	
D-4/3/90	32229.0	5.00	7.6	182.5	528.0	186.0	69.9	3.4	1666	7.7	
D-5/3/90	35023.0	3.50	7.9	205.0	588.0	192.0	65.6	4.5	2430	7.8	
D-6/3/90	36924.0	1.50	8.0	242.0	496.0	176.0	64.8	4.0	2110	7.9	
D-7/3/90	38572.0	3.00	7.8	202.0	372.0	186.0	68.8	4.5	1644	7.8	
D-8/3/90	41115.0	6.00	7.8	182.5	552.0	262.0	64.1	5.0	1603	7.8	
D-9/3/90	36107.0	5.00	7.7	215.0	489.0	334.0	40.7	6.0	1613	7.6	
D-11/3/90	29156.0	2.50	7.7	206.0	451.0	194.0	69.1	4.5	1249	7.7	
D-12/3/90	39246.0	2.00	7.8	172.0	506.0	200.0	69.0	5.0	1865	7.8	
D-13/3/90	42393.0	0.70	7.9	189.0	478.0	230.0	67.0	5.5	1410	8.1	
D-14/3/90	42857.0	1.50	7.7	238.0	319.0	292.0	33.8	3.5	1261	7.6	
D-15/3/90	42911.0	0.70	7.6	114.0	252.0	116.0	58.6	1.2	1238	7.9	
D-16/3/90	40376.0	1.50	8.1	204.0	333.0	174.0	67.8	3.0	2390	7.8	
D-18/3/90	40923.0	3.50	7.6	146.0	329.0	188.0	57.4	2.5	1300	7.6	
D-19/3/90	43830.0	1.50	7.8	177.0	512.0	214.0	58.9	5.5	1605	7.7	
D-20/3/90	39165.0	1.20	7.4	250.0	447.0	252.0	61.1	7.0	1533	7.4	
D-21/3/90	35791.0	1.20	7.8	277.0	466.0	246.0	63.4	4.0	1556	7.7	
D-22/3/90	37419.0	1.20	7.6	219.0	446.0	222.0	61.3	5.5	1600	7.7	
D-23/3/90	40983.0	3.00	7.6	182.0	431.0	214.0	57.0	7.0	1591	7.5	
D-25/3/90	42217.0	8.50	7.5	138.0	333.0	240.0	55.0	3.8	1087	7.5	
D-26/3/90	47665.0	1.20	7.7	156.0	405.0	200.0	74.0	4.0	1856	7.6	
D-27/3/90	44314.0	3.00	7.8	155.0	389.0	308.0	49.4	6.0	1927	7.7	
D-28/3/90	40841.0	1.00	7.6	179.0	389.0	168.0	69.0	3.5	1240	7.8	
D-29/3/90	41157.0	3.00	8.0	145.0	398.0	192.0	66.7	4.5	2240	8.0	
D-30/3/90	40078.0	1.40	7.9	198.0	464.0	228.0	64.9	4.6	1431	7.6	
D-1/2/90	44365.0	7.50	7.9	182.5	365.0	212.0	62.3	3.5	1339	7.9	
D-2/2/90	43080.0	4.25	7.8	95.0	349.0	136.0	76.5	2.5	1063	7.8	
D-4/2/90	29414.0	3.00	7.6	160.0	374.0	168.0	69.0	3.1	1042	7.6	
D-5/2/90	37312.0	1.00	8.1	205.0	492.0	192.0	70.8	4.0	1454	8.1	
...	
D-25/10/91	35400.0	0.70	7.6	156.0	364.0	194.0	63.9	5.5	1680	7.6	
D-26/10/91	30964.0	3.30	7.7	220.0	540.0	184.0	62.0	3.5	1445	7.7	
D-27/10/91	35573.0	7.30	7.6	176.0	333.0	178.0	64.0	3.5	1627	7.7	
D-29/10/91	29801.0	1.60	7.7	172.0	400.0	136.0	70.1	1.5	1402	7.7	
D-30/10/91	31524.0	1.60	7.9	182.5	478.0	204.0	64.7	6.0	1798	7.9	
D-1/8/91	29834.0	3.00	7.4	160.0	348.0	194.0	61.9	3.0	1720	7.5	
D-2/8/91	28492.0	2.60	7.5	124.0	281.0	172.0	66.3	3.0	1520	7.5	
D-4/8/91	24978.0	0.50	7.3	146.0	288.0	124.0	67.7	2.0	1210	7.4	
D-5/8/91	29719.0	0.20	7.6	133.0	284.0	186.0	71.0	5.0	1114	7.6	
D-6/8/91	29741.0	0.45	7.9	151.0	316.0	196.0	64.3	2.5	948	7.8	
D-7/8/91	29027.0	0.40	7.6	136.0	328.0	186.0	67.7	3.0	899	7.6	
D-8/8/91	30211.0	0.50	7.6	114.0	521.0	506.0	44.3	7.5	866	7.5	
D-9/8/91	30848.0	0.20	7.7	142.0	376.0	144.0	70.8	3.0	940	7.6	
D-11/8/91	17527.0	0.55	7.5	150.0	171.0	172.0	37.2	1.4	732	7.5	
D-12/8/91	33331.0	0.23	7.6	92.0	233.0	234.0	37.6	1.4	829	7.6	
D-13/8/91	27998.0	0.62	7.5	138.0	268.0	154.0	66.2	1.7	890	7.5	
D-14/8/91	32845.0	0.23	7.6	84.0	251.0	98.0	71.4	2.0	866	7.6	
D-16/8/91	27933.0	0.20	7.6	158.0	375.0	178.0	60.7	3.5	1049	7.7	
D-18/8/91	27527.0	0.20	7.3	191.0	240.0	166.0	73.5	3.0	1072	7.4	
D-19/8/91	32363.0	0.10	7.6	159.0	310.0	146.0	68.5	1.6	1096	7.6	
D-20/8/91	31437.0	0.47	7.6	132.0	304.0	148.0	64.9	2.0	939	7.7	
D-21/8/91	31914.0	2.00	7.7	127.0	274.0	144.0	72.2	2.0	1031	7.6	
D-22/8/91	28088.0	0.20	7.5	153.0	307.0	124.0	82.3	2.5	1044	7.6	
D-23/8/91	27838.0	0.13	7.6	179.0	265.0	128.0	71.9	1.8	992	7.6	

D-25/8/91	29271.0	0.36	7.5	99.0	585.0	140.0	71.4	4.5	962	7.6
D-26/8/91	32723.0	0.16	7.7	93.0	252.0	176.0	56.8	2.3	894	7.7
D-27/8/91	33535.0	0.32	7.8	192.0	346.0	172.0	68.6	4.0	988	7.8
D-28/8/91	32922.0	0.30	7.4	139.0	367.0	180.0	64.4	3.0	1060	7.5
D-29/8/91	32190.0	0.30	7.3	200.0	545.0	258.0	65.1	4.0	1260	7.4
D-30/8/91	30488.0	0.21	7.5	152.0	300.0	132.0	69.7	4.5	1073	7.4

	...	29	30	31	32	33	34	35	36	37 \
0	...									
D-1/3/90	...	2000.0	39.6	58.8	95.5	85.4	70.0	90.2	79.40	87.3
D-2/3/90	...	2590.0	39.6	60.7	94.8	85.4	80.8	90.2	79.50	92.1
D-4/3/90	...	1888.0	39.6	58.2	95.6	85.4	52.9	90.2	75.80	88.7
D-5/3/90	...	1840.0	33.1	64.2	95.3	87.3	72.3	90.2	82.30	89.6
D-6/3/90	...	2120.0	39.6	62.7	95.6	85.4	71.0	92.1	78.20	87.5
D-7/3/90	...	1764.0	39.6	59.7	96.5	86.7	78.3	90.1	73.10	84.9
D-8/3/90	...	1703.0	39.6	61.9	93.8	89.1	79.8	90.2	86.20	90.1
D-9/3/90	...	1606.0	39.6	70.4	95.6	90.6	53.7	92.1	66.90	94.6
D-11/3/90	...	1338.0	46.1	43.6	92.5	85.6	58.2	92.2	73.80	90.2
D-12/3/90	...	1616.0	21.2	59.7	90.8	88.4	66.1	89.0	69.00	86.5
D-13/3/90	...	1575.0	0.6	45.8	92.0	11.6	25.7	19.6	36.00	43.0
D-14/3/90	...	1304.0	31.8	61.2	92.9	85.4	69.9	90.2	79.15	18.5
D-15/3/90	...	1221.0	46.6	48.5	91.7	85.4	20.4	26.3	31.70	10.3
D-16/3/90	...	2550.0	41.1	50.0	84.0	76.5	52.9	84.3	54.10	43.7
D-18/3/90	...	1545.0	32.7	33.3	90.0	82.6	61.3	87.0	71.40	78.2
D-19/3/90	...	1110.0	28.0	72.7	92.7	78.8	36.6	85.9	60.40	90.7
D-20/3/90	...	1402.0	49.8	58.3	92.3	89.9	96.8	94.4	98.10	92.1
D-21/3/90	...	1606.0	39.6	39.6	78.3	85.5	70.4	91.3	73.40	89.4
D-22/3/90	...	1780.0	35.3	57.5	92.0	75.6	49.3	80.8	60.80	76.1
D-23/3/90	...	1597.0	20.1	64.5	93.6	86.9	68.1	87.4	72.20	88.3
D-25/3/90	...	1223.0	29.4	55.4	91.3	85.2	68.0	88.4	81.40	92.9
D-26/3/90	...	1706.0	28.1	50.0	91.4	85.2	76.5	87.8	82.50	88.5
D-27/3/90	...	1869.0	52.0	64.9	90.8	87.6	71.2	90.3	77.60	92.5
D-28/3/90	...	1416.0	26.7	66.2	95.0	89.2	72.5	91.1	78.70	88.1
D-29/3/90	...	2290.0	34.3	65.0	94.2	89.3	69.7	89.7	78.10	89.1
D-30/3/90	...	1475.0	27.2	67.6	97.3	90.4	68.0	91.4	78.00	90.4
D-1/2/90	...	1377.0	39.6	50.0	89.4	85.4	76.6	90.2	79.50	90.6
D-2/2/90	...	1220.0	6.8	47.9	80.0	84.6	78.9	80.0	80.80	82.4
D-4/2/90	...	1087.0	42.7	54.5	95.7	84.9	73.6	88.1	78.90	83.3
D-5/2/90	...	1275.0	39.6	33.0	94.5	85.3	73.7	84.4	76.80	80.7
...
D-25/10/91	...	1840.0	47.3	61.3	94.0	76.4	69.9	86.5	82.40	90.7
D-26/10/91	...	1337.0	39.6	38.6	93.3	85.4	87.0	92.7	95.00	91.8
D-27/10/91	...	1799.0	39.6	40.4	95.0	85.4	72.9	90.9	79.90	90.7
D-29/10/91	...	1468.0	32.4	40.4	88.0	87.8	77.6	91.3	85.30	83.8
D-30/10/91	...	1568.0	39.6	43.9	65.3	85.4	75.3	90.2	81.20	89.7
D-1/8/91	...	1772.0	29.1	53.5	92.0	85.7	64.2	90.6	72.70	85.6
D-2/8/91	...	1549.0	34.2	47.7	99.0	84.4	66.1	90.3	73.30	88.4
D-4/8/91	...	1259.0	39.6	52.3	92.3	75.3	69.9	86.3	79.15	76.6
D-5/8/91	...	1100.0	55.1	73.2	96.0	82.0	62.5	91.7	78.90	88.7
D-6/8/91	...	951.0	44.8	64.5	95.0	82.2	80.0	89.4	86.10	88.8
D-7/8/91	...	898.0	39.6	64.1	94.0	84.9	60.4	91.9	76.80	87.6
D-8/8/91	...	884.0	48.7	86.9	98.8	81.0	67.9	90.4	91.60	96.0
D-9/8/91	...	947.0	39.6	47.6	97.3	85.4	77.6	93.7	84.80	90.3
D-11/8/91	...	728.0	62.8	77.7	90.0	73.8	65.5	92.7	77.20	90.7
D-12/8/91	...	929.0	36.9	43.6	86.7	87.7	71.9	91.3	79.80	92.3
D-13/8/91	...	858.0	38.1	41.6	86.7	87.7	79.0	94.2	87.70	91.6
D-14/8/91	...	879.0	50.9	36.5	80.0	87.0	69.6	91.7	80.50	82.7

D-16/8/91	...	828.0	68.0	66.7	96.7	81.6	41.8	94.3	72.50	83.1
D-18/8/91	...	999.0	38.5	54.5	88.0	90.0	69.9	95.8	81.70	93.4
D-19/8/91	...	1083.0	25.2	61.4	91.2	78.6	65.1	86.8	81.00	89.0
D-20/8/91	...	1012.0	45.6	60.3	94.4	82.5	72.9	89.4	86.20	91.2
D-21/8/91	...	1053.0	39.6	59.5	93.3	85.4	77.7	90.2	79.15	90.7
D-22/8/91	...	1038.0	40.5	54.4	94.0	89.7	75.5	93.5	85.00	90.3
D-23/8/91	...	1044.0	13.7	45.0	95.0	87.5	71.3	93.9	79.60	89.1
D-25/8/91	...	968.0	40.8	71.6	93.3	59.0	26.4	74.7	83.80	81.4
D-26/8/91	...	942.0	39.6	62.3	93.3	69.8	75.9	79.6	78.60	96.6
D-27/8/91	...	950.0	39.6	58.3	97.8	83.0	59.1	91.1	74.60	90.7
D-28/8/91	...	1136.0	39.6	65.0	97.1	76.2	66.4	82.0	77.10	88.9
D-29/8/91	...	1326.0	39.8	65.9	97.1	81.7	70.9	89.5	87.00	89.5
D-30/8/91	...	1224.0	39.6	69.5	93.3	81.7	76.4	90.2	81.70	86.4

38

0	
D-1/3/90	99.6
D-2/3/90	100.0
D-4/3/90	98.5
D-5/3/90	100.0
D-6/3/90	99.5
D-7/3/90	100.0
D-8/3/90	99.0
D-9/3/90	100.0
D-11/3/90	99.4
D-12/3/90	99.6
D-13/3/90	36.4
D-14/3/90	42.9
D-15/3/90	95.4
D-16/3/90	100.0
D-18/3/90	99.2
D-19/3/90	100.0
D-20/3/90	100.0
D-21/3/90	99.4
D-22/3/90	99.6
D-23/3/90	100.0
D-25/3/90	100.0
D-26/3/90	99.8
D-27/3/90	100.0
D-28/3/90	100.0
D-29/3/90	100.0
D-30/3/90	100.0
D-1/2/90	98.6
D-2/2/90	100.0
D-4/2/90	100.0
D-5/2/90	100.0
...	...
D-25/10/91	99.8
D-26/10/91	95.7
D-27/10/91	98.6
D-29/10/91	96.7
D-30/10/91	99.2
D-1/8/91	99.3
D-2/8/91	99.3
D-4/8/91	99.0
D-5/8/91	99.6
D-6/8/91	100.0

```

D-7/8/91      99.3
D-8/8/91      99.7
D-9/8/91      99.3
D-11/8/91     98.6
D-12/8/91     99.3
D-13/8/91     98.8
D-14/8/91     99.0
D-16/8/91     98.6
D-18/8/91     99.3
D-19/8/91     99.4
D-20/8/91     99.5
D-21/8/91     99.7
D-22/8/91    100.0
D-23/8/91    100.0
D-25/8/91     99.7
D-26/8/91     99.6
D-27/8/91    100.0
D-28/8/91     99.0
D-29/8/91     99.8
D-30/8/91     99.7

```

[527 rows x 38 columns]

```

In [330]: # Scale the variables
from sklearn.preprocessing import StandardScaler
scaled = StandardScaler()
scaled.fit(Water_cleaned)
Water_cleaned_scaled = scaled.transform(Water_cleaned) # Scaled dataset

```

```

In [331]: print(Water_cleaned_scaled)

[[ 1.06097737 -0.31176336 -0.04089158 ...,  0.17091754 -0.20723325
   0.11050301]
 [ 0.28219456  0.23571845 -0.44749277 ...,  0.1824499   0.38144415
   0.20304983]
 [-0.76011965  0.96569418 -0.85409395 ..., -0.24424763 -0.03553567
  -0.14400075]
 ...,
 [-0.65381741 -0.7497488  -1.66729633 ..., -0.09432687 -0.01100745
  -0.02831722]
 [-0.76610203 -0.7497488  -2.07389751 ...,  1.04737733  0.06257723
   0.15677642]
 [-1.02717911 -0.7825977  -1.26069514 ...,  0.43616195 -0.31761026
   0.13363972]]

```

```

In [332]: # Determine the two principal components among the features
from sklearn.decomposition import PCA
pca = PCA(n_components = 2) # Request that two principal components be calculated from the features
pca.fit(Water_cleaned_scaled) # fit the dataset
pca_comps = pca.transform(Water_cleaned_scaled) # assign the principal components to a variable

```

In [333]: `print(pca_comps)`

```
[[ 1.61324758 -0.29293352]
 [ 4.30829529 -0.48676991]
 [ 0.45812616 -1.06688971]
 ...,
 [-3.07358511 -0.47657264]
 [-1.68898355  0.39533681]
 [-2.98845529  0.52221904]]
```

In [334]: `print(pca_comps.shape) # Shows the new shape of the dataframe`

notice that instead of 38 columns, it has been reduced to 2 columns

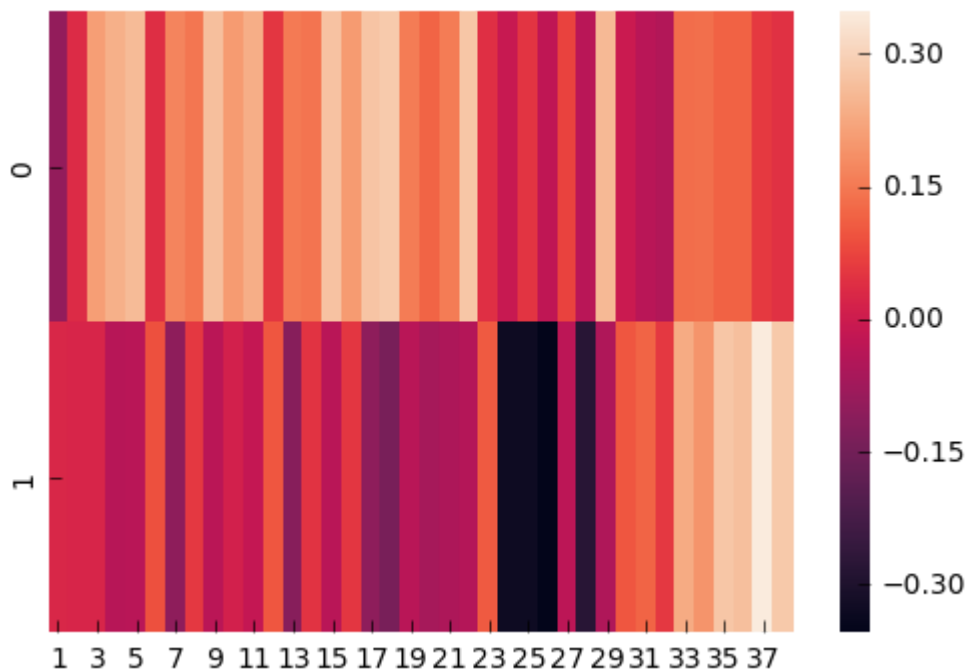
```
(527, 2)
```

In [335]: `# Show the PCA components`
`print(pca.components_)`

```
[[-0.09607625  0.03404058  0.20829207  0.24007351  0.25809293  0.040229
  0.16702766  0.14580396  0.26566311  0.20340852  0.23675963  0.05168039
  0.15127654  0.14275293  0.27136624  0.20246357  0.27376737  0.28617692
  0.15425277  0.11842546  0.153793    0.27793829  0.04171031 -0.00904403
  0.05043745 -0.02302684  0.07194353 -0.03469232  0.25436505 -0.00426902
 -0.03510447 -0.04680218  0.13162004  0.13405186  0.11654486  0.11693155
  0.05690212  0.04311658]
 [ 0.02592851  0.0226068  0.02370611 -0.03887028 -0.03565239  0.09137956
 -0.103618    0.05467435 -0.03317637  0.01134908 -0.0169161  0.10020294
 -0.11585381  0.0467968  -0.03506975  0.05368288 -0.10234408 -0.14042989
 -0.03563338 -0.06542278 -0.05517845 -0.04312377  0.10349285 -0.32903646
 -0.32960542 -0.35439283 -0.03006309 -0.28014293 -0.05164938  0.09859565
  0.1193992   0.05455503  0.23046802  0.19290569  0.27690464  0.2666469
  0.34737629  0.28368542]]
```

```
In [336]: # Create a visual heat map of the prinmcipal components
df_pca_comps = pd.DataFrame(data = pca.components_ , columns =
np.arange(1,39))
sns.heatmap(data = df_pca_comps)
plt.figure(figsize=(40,10))
plt.show()
```

<matplotlib.figure.Figure at 0xc050090>



<matplotlib.figure.Figure at 0xc0cab30>

```
In [337]: np.arange(1,39)
```

```
Out[337]: array([ 1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12, 13, 14, 15, 16, 17,
        18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34,
        35, 36, 37, 38])
```

```
In [339]: print(df_pca_comps)
```

	1	2	3	4	5	6	7	\
0	-0.096076	0.034041	0.208292	0.240074	0.258093	0.040229	0.167028	
1	0.025929	0.022607	0.023706	-0.038870	-0.035652	0.091380	-0.103618	
	8	9	10	...	29	30	31	\
0	0.145804	0.265663	0.203409	...	0.254365	-0.004269	-0.035104	
1	0.054674	-0.033176	0.011349	...	-0.051649	0.098596	0.119399	
	32	33	34	35	36	37	38	
0	-0.046802	0.131620	0.134052	0.116545	0.116932	0.056902	0.043117	
1	0.054555	0.230468	0.192906	0.276905	0.266647	0.347376	0.283685	

[2 rows x 38 columns]

```
In [340]: from scipy.cluster.hierarchy import dendrogram, linkage
```

```
In [341]: Water_matrix = np.array(Water_cleaned_scaled)
```

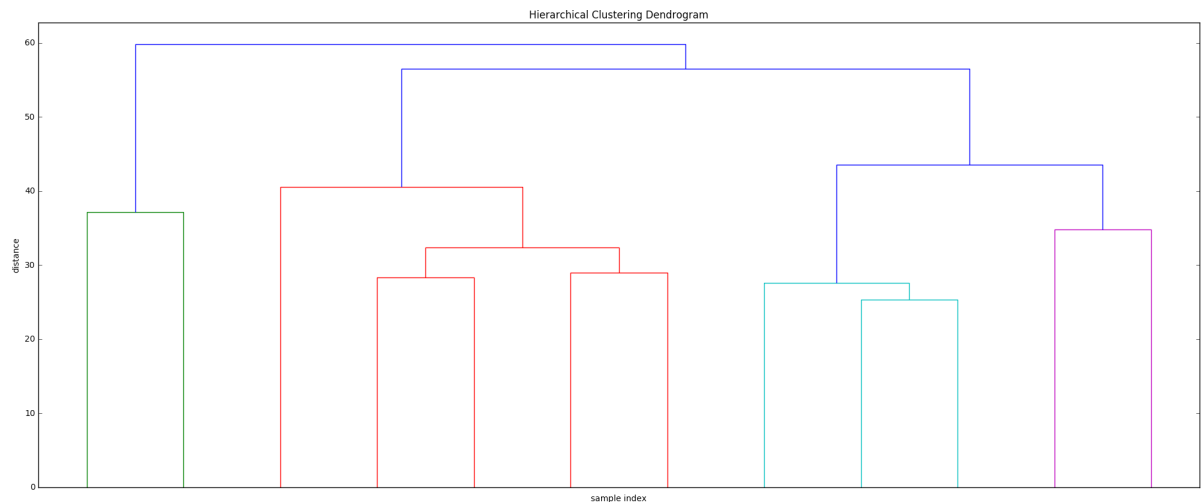
```
In [342]: # generate the Linkage matrix  
Z = linkage(Water_matrix, 'ward')
```

```
In [343]: from scipy.cluster.hierarchy import cophenet  
from scipy.spatial.distance import pdist
```

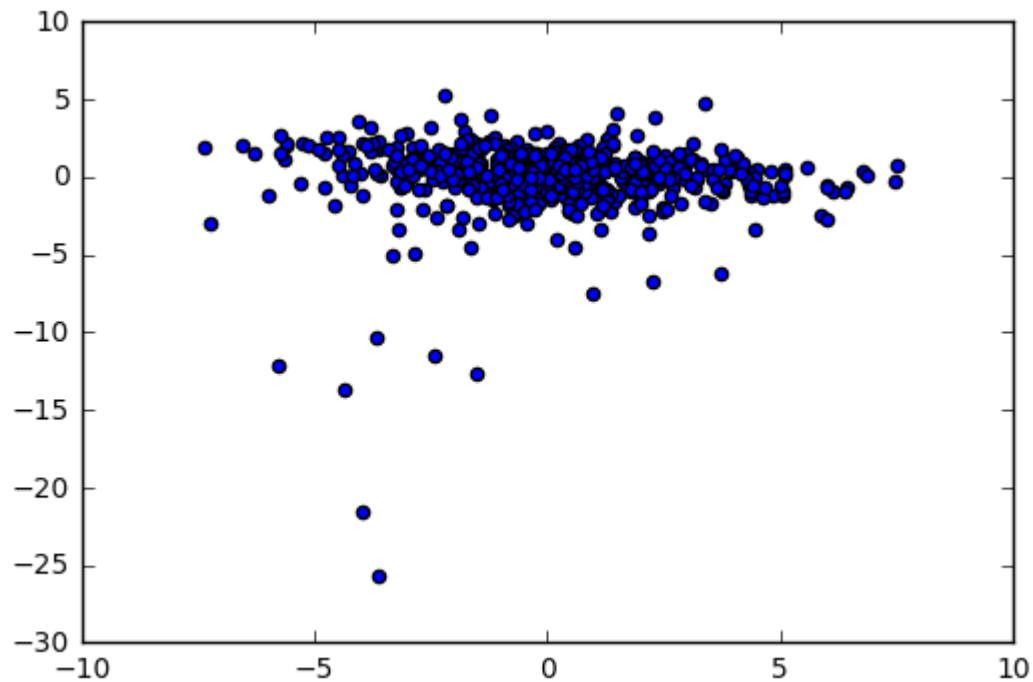
```
In [344]: c, coph_dists = cophenet(Z, pdist(Water_matrix))  
print(c)
```

```
0.298966756872
```

```
In [345]: # calculate full dendrogram  
plt.figure(figsize=(25, 10))  
plt.title('Hierarchical Clustering Dendrogram')  
plt.xlabel('sample index')  
plt.ylabel('distance')  
dendrogram(  
    Z,  
    truncate_mode='lastp', # show only the last p merged clusters  
    p=12, # show only the last p merged clusters  
    show_leaf_counts=False, # otherwise numbers in brackets are counts  
    leaf_rotation=90., # rotates the x axis labels  
    leaf_font_size=8., # font size for the x axis labels  
)  
plt.show()
```



```
In [346]: plt.scatter(pca_comps.transpose()[0], pca_comps.transpose()[1])  
plt.show()
```



```
In [348]: print(Water_matrix.shape)  
  
(527, 38)
```

```
In [349]: # Import Kmeans  
from sklearn.cluster import KMeans
```

```
In [350]: # Create an instance of a Kmeans model with 4 clusters  
kmeans = KMeans(n_clusters = 4)
```

```
In [351]: # fit the data  
kmeans.fit(Water_cleaned_scaled)
```

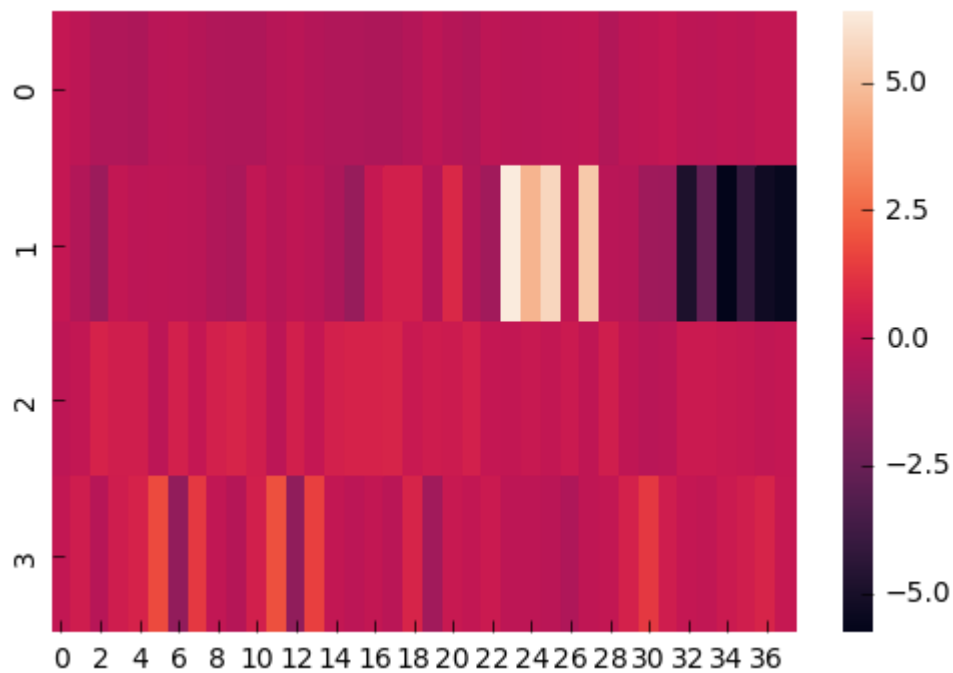
```
Out[351]: KMeans(copy_x=True, init='k-means++', max_iter=300, n_clusters=4, n_init=10,  
              n_jobs=1, precompute_distances='auto', random_state=None, tol=0.0001,  
              verbose=0)
```

```
In [354]: print(kmeans.cluster_centers_.shape)  
  
(4, 38)
```

```
In [355]: print(kmeans.cluster_centers_)
```

```
[[ 7.44631005e-02 -1.04508645e-01 -4.96284909e-01 -4.75560726e-01
  -5.53981098e-01 -2.34544662e-01 -2.00086601e-01 -3.55889313e-01
  -4.61135125e-01 -5.15113282e-01 -5.30001446e-01 -2.76004456e-01
  -1.72102716e-01 -3.96802827e-01 -4.69883380e-01 -5.01264436e-01
  -5.76768057e-01 -5.91260066e-01 -3.80515538e-01 -1.09953146e-01
  -3.31590081e-01 -4.87507083e-01 -1.21214736e-01 -1.82521561e-01
  -2.64345137e-01 -1.38915710e-01 -1.11561273e-01 -7.37445111e-02
  -4.31578861e-01 -8.16547210e-02 -3.26312569e-02 7.86622579e-02
  -1.13640279e-01 -1.48247419e-01 -4.05549479e-02 -1.05720851e-01
  2.34301794e-02 5.42345440e-02]
 [ 6.95462438e-02 -4.31687940e-01 -1.02835160e+00 6.57169026e-02
  -1.49040335e-01 -7.75894413e-02 -1.63763121e-01 -2.09608344e-01
  -4.76286869e-01 -6.36350928e-01 1.08098429e-02 -3.09987192e-01
  -3.78493793e-02 -1.90675922e-01 -5.85427435e-01 -1.20622006e+00
  1.53992793e-01 5.34520413e-01 5.28617416e-01 -3.98734820e-01
  8.54242030e-01 -4.48645017e-01 -8.93101092e-01 6.38031986e+00
  4.64385846e+00 5.68521702e+00 -5.05936155e-02 5.28441679e+00
  -2.42371456e-01 -2.68845718e-01 -9.76339435e-01 -1.00801592e+00
  -4.86510250e+00 -2.67346962e+00 -5.77410168e+00 -4.09688152e+00
  -5.28457576e+00 -5.51832681e+00]
 [ -9.39950896e-02 3.30999512e-02 6.71593068e-01 4.47730285e-01
  4.85033927e-01 -1.53412193e-01 5.56897143e-01 9.53555425e-02
  5.40691397e-01 6.95311497e-01 4.84276675e-01 -1.34753236e-01
  5.32445641e-01 9.50608560e-02 5.50581577e-01 6.50880811e-01
  6.50482264e-01 7.05399471e-01 2.54305186e-01 3.57344789e-01
  3.11869777e-01 5.60096122e-01 9.93218200e-02 2.81574380e-02
  1.76114944e-01 2.09849510e-02 2.58505435e-01 -6.82652271e-02
  4.91709690e-01 -4.25595300e-02 -2.52365974e-01 -1.62027967e-01
  2.64338752e-01 2.56844568e-01 1.68835342e-01 1.46133532e-01
  -2.60844302e-02 8.46068821e-02]
 [ 2.66986670e-02 4.21792067e-01 -2.91107695e-01 4.00480023e-01
  6.50022322e-01 1.78121442e+00 -1.35068433e+00 1.33946305e+00
  1.43666292e-02 -3.52790946e-01 5.16391874e-01 1.93360057e+00
  -1.39966395e+00 1.53484945e+00 2.96548726e-02 -1.56391681e-01
  2.51717541e-02 -1.86609826e-01 6.92116000e-01 -9.15802572e-01
  1.71734897e-01 5.60832949e-02 2.86600671e-01 -9.94263516e-02
  -9.25727447e-02 -1.85429056e-01 -5.40571371e-01 -7.06340433e-02
  4.61269798e-02 6.07184189e-01 1.34630716e+00 4.36781611e-01
  9.30757581e-02 -4.15334194e-03 2.64448924e-01 4.47139874e-01
  7.08094063e-01 1.27410603e-01]]
```

```
In [356]: sns.heatmap(data = kmeans.cluster_centers_)
plt.figure(figsize=(40,10))
plt.show()
```



<matplotlib.figure.Figure at 0xbc11890>

```
In [357]: C = kmeans.cluster_centers_.transpose()
```

```
In [359]: print(C.shape)
```

(38, 4)

```
In [360]: kmeans.labels_.shape
```

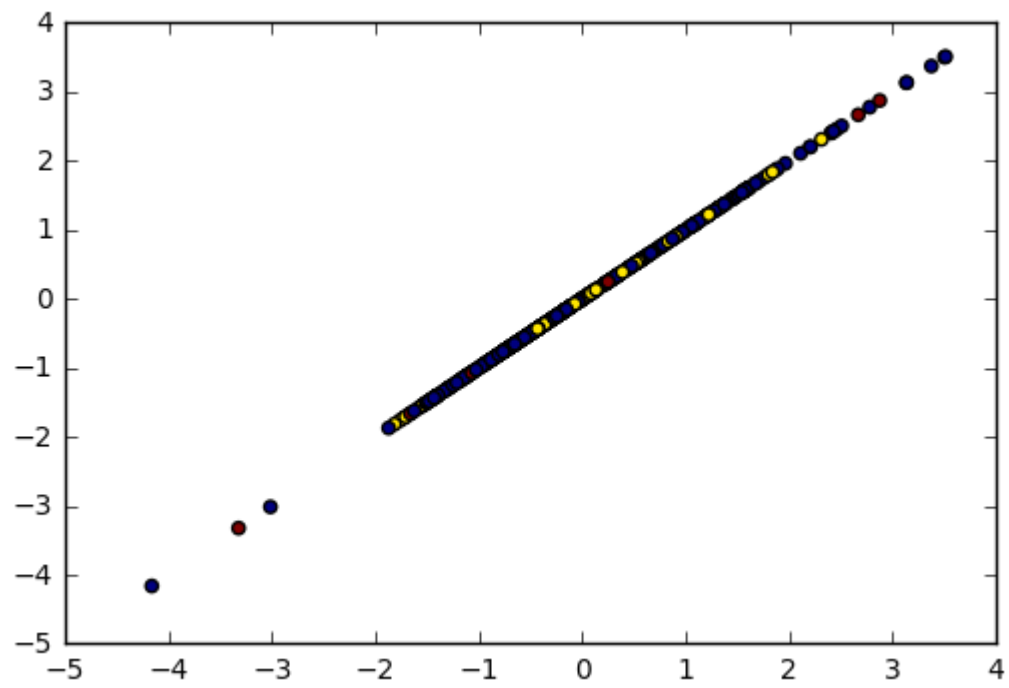
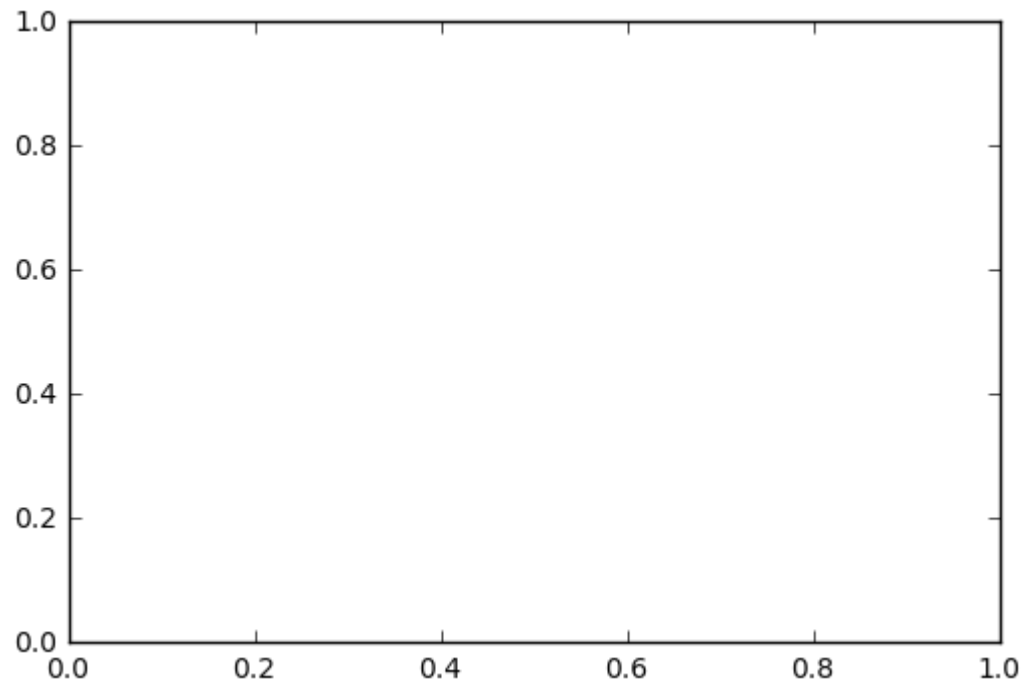
```
Out[360]: (527,)
```

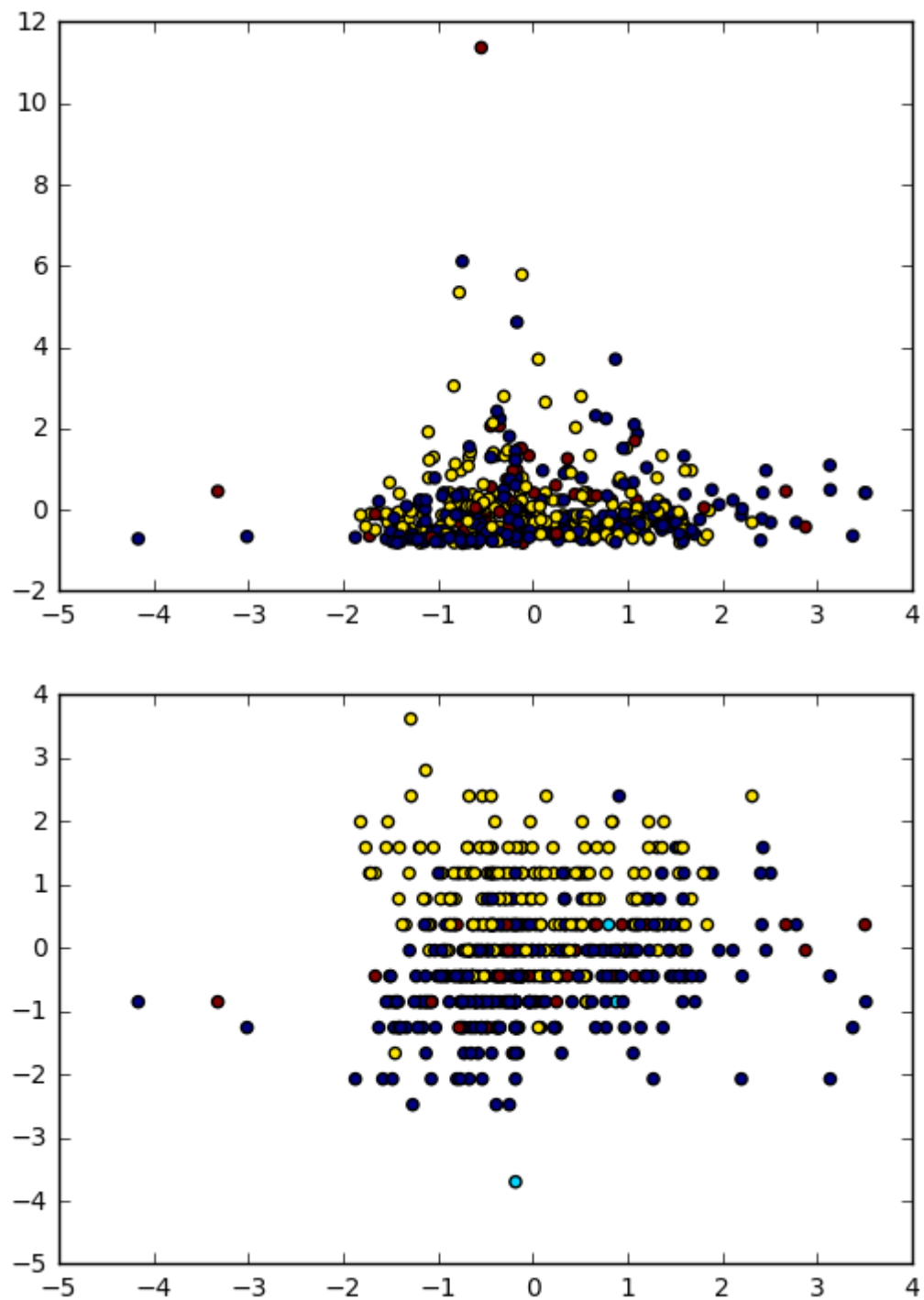
```
In [361]: C[:,0].shape
```

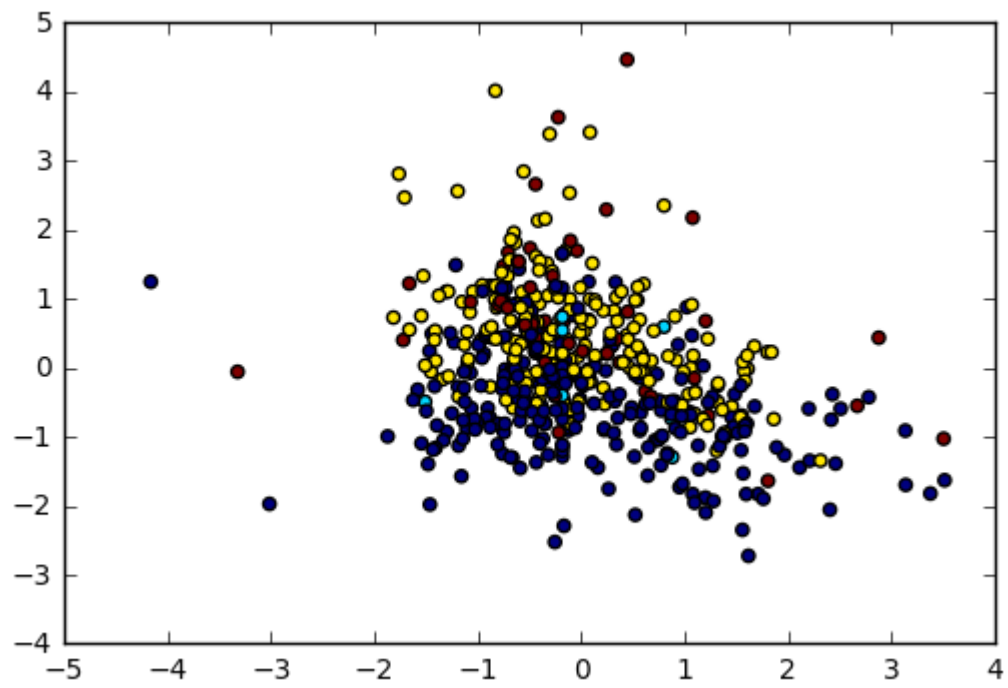
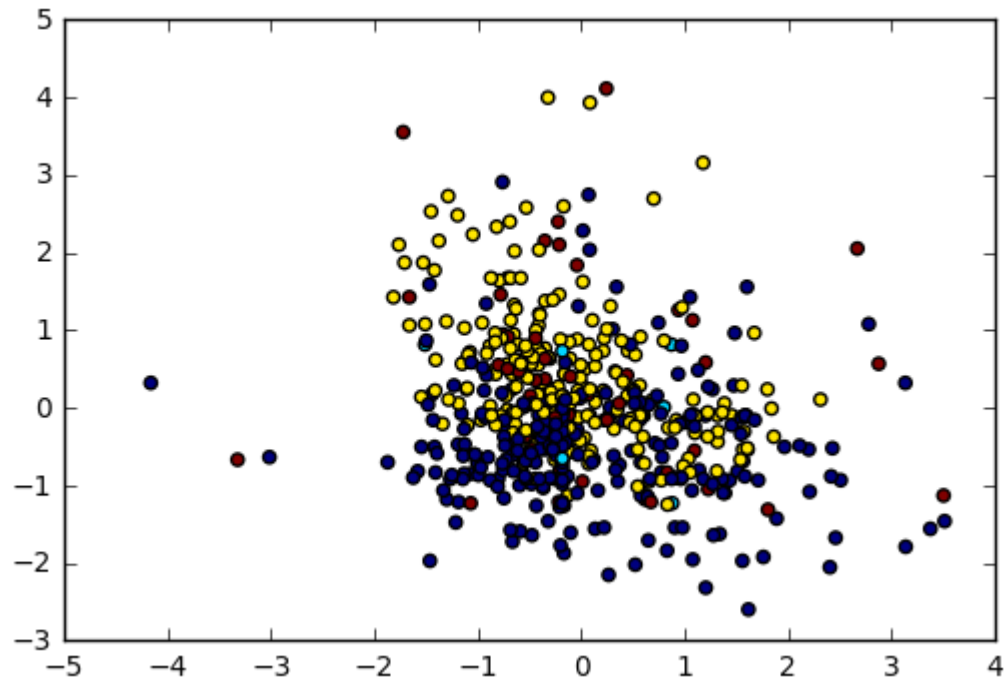
```
Out[361]: (38,)
```

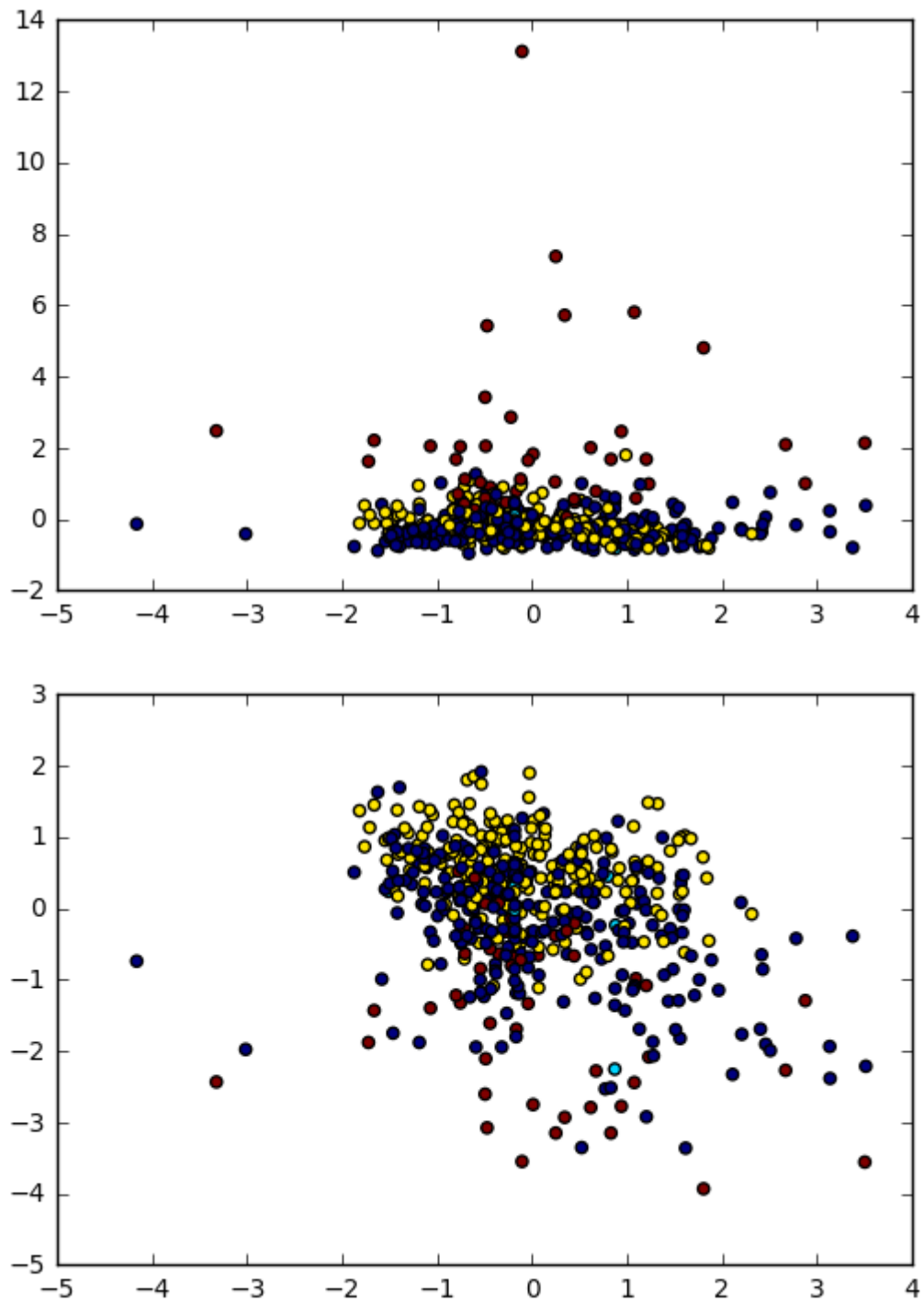


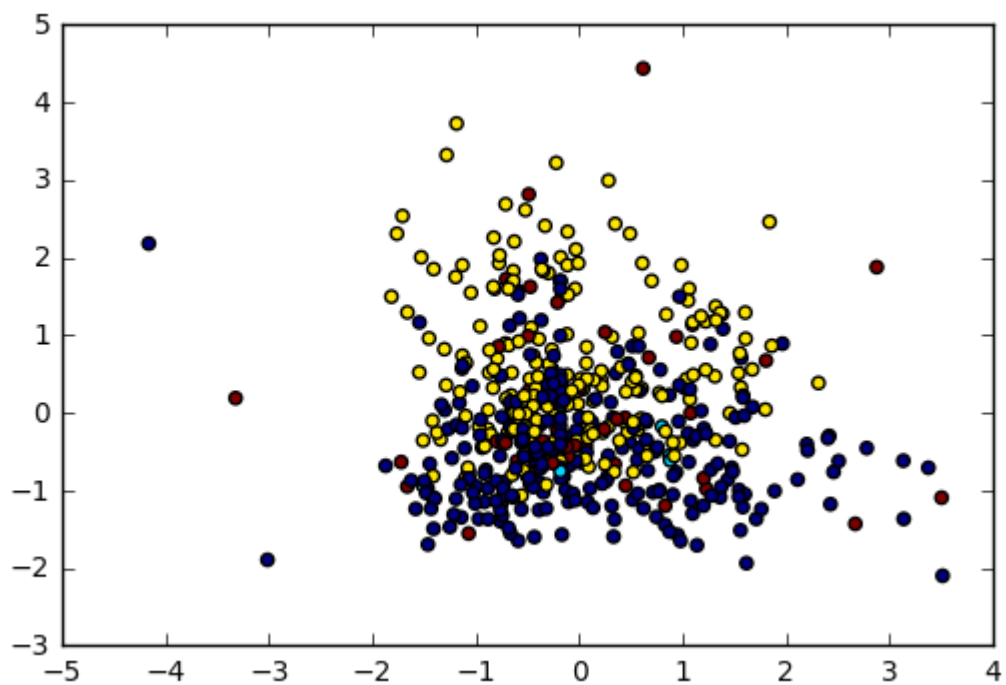
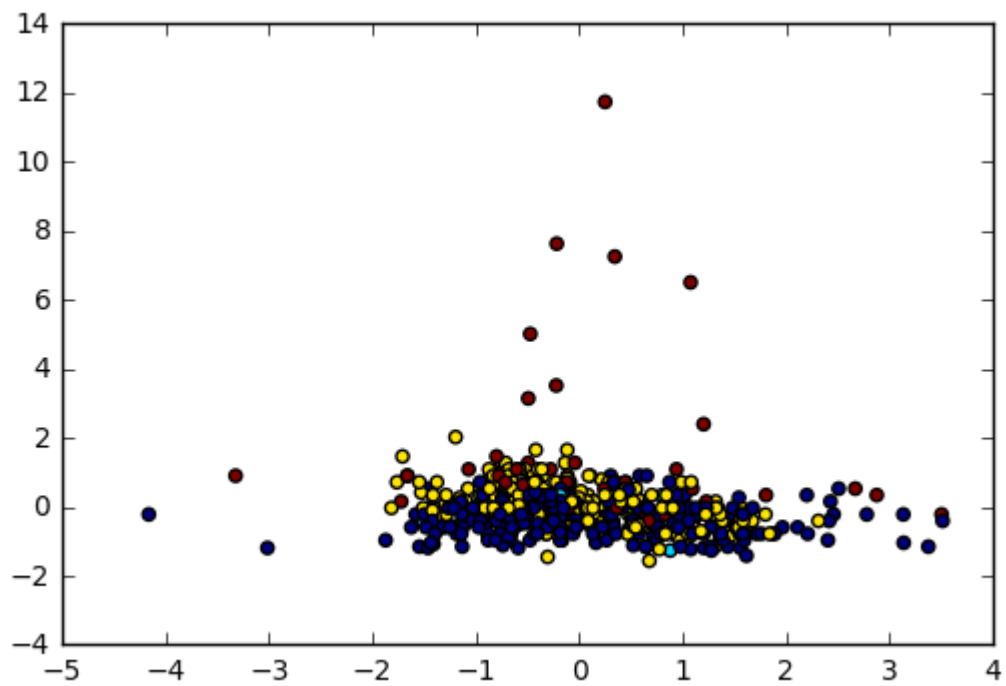
```
In [363]: m,n = Water_matrix.shape
          for i in range(n):
              #print(i)
              plt.figure()
              plt.scatter(Water_matrix[:,0],Water_matrix[:,i], c = kmeans.labels_)
              plt.show()
```

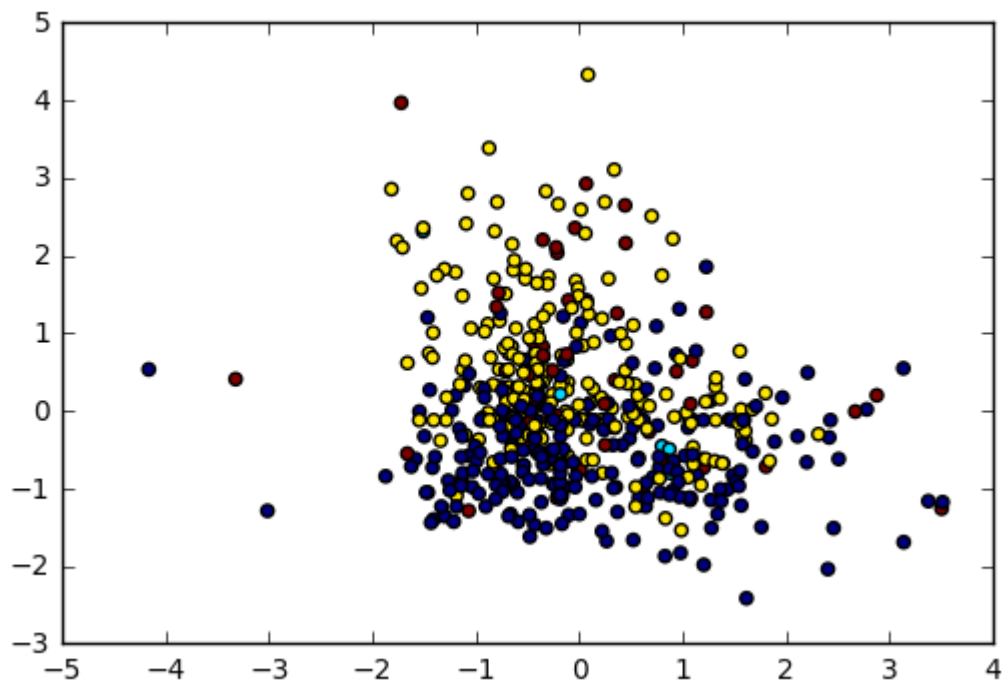
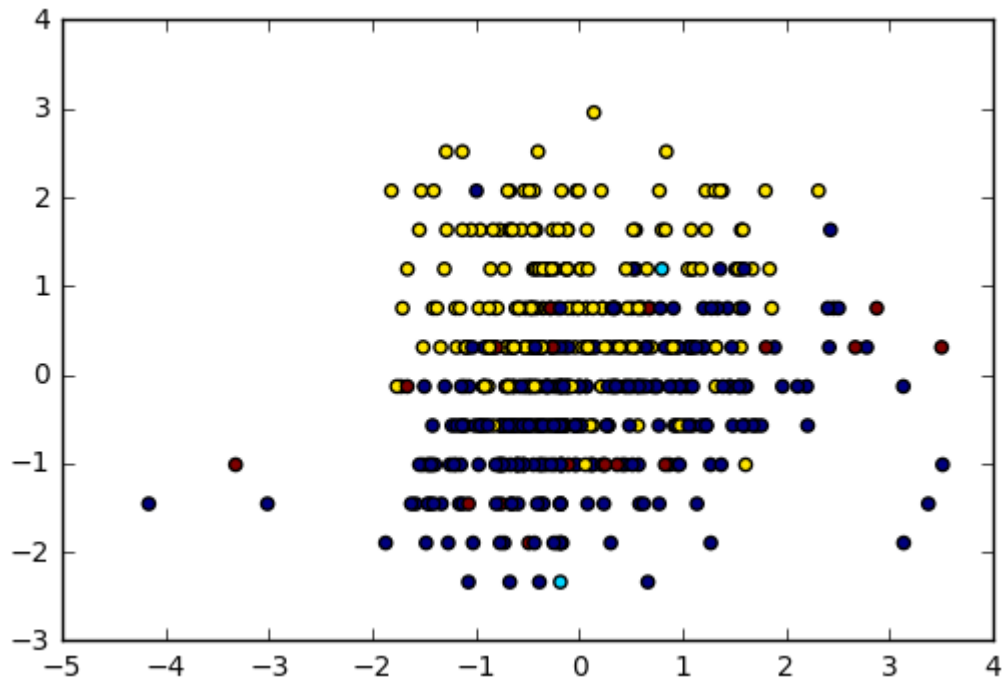


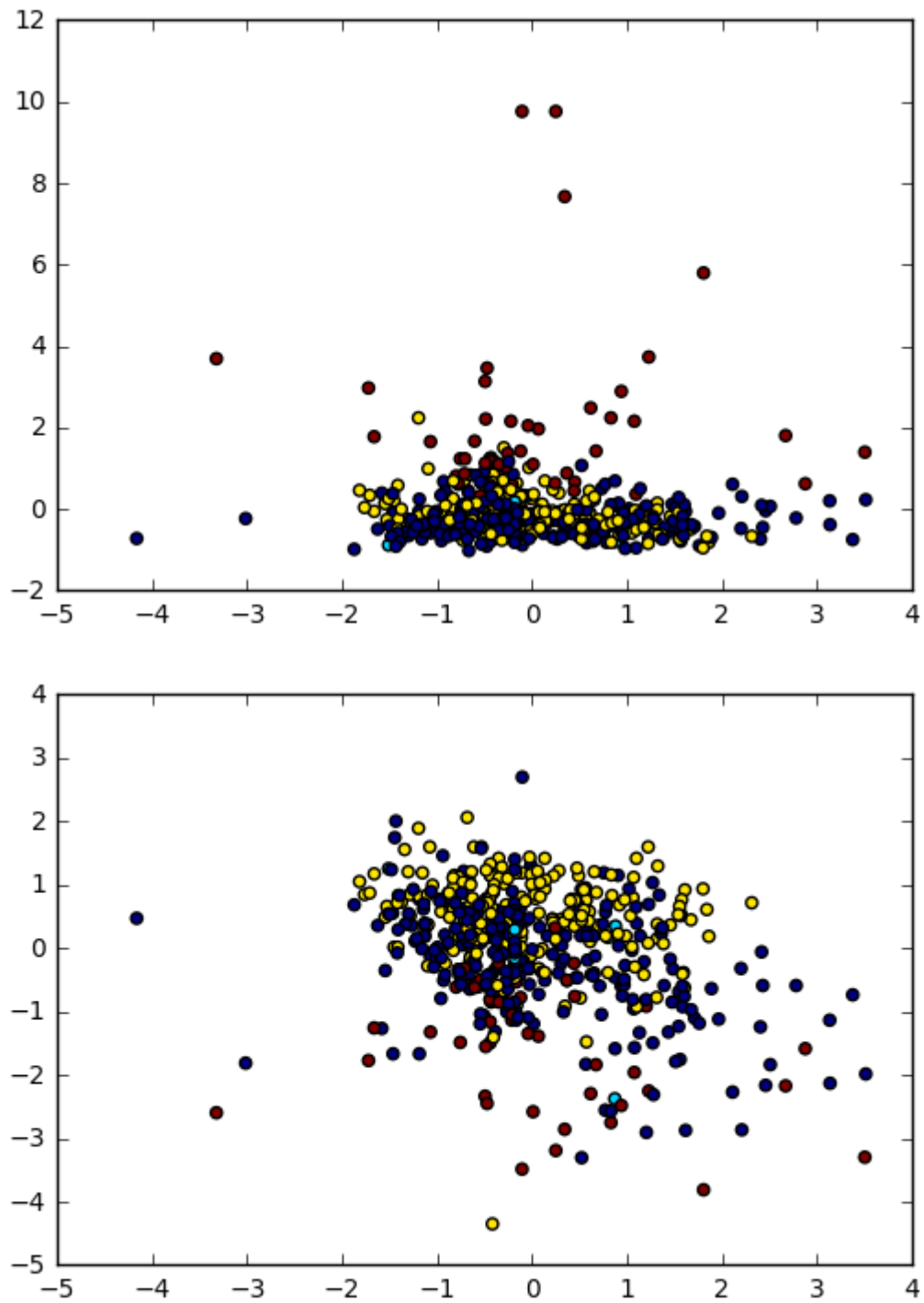


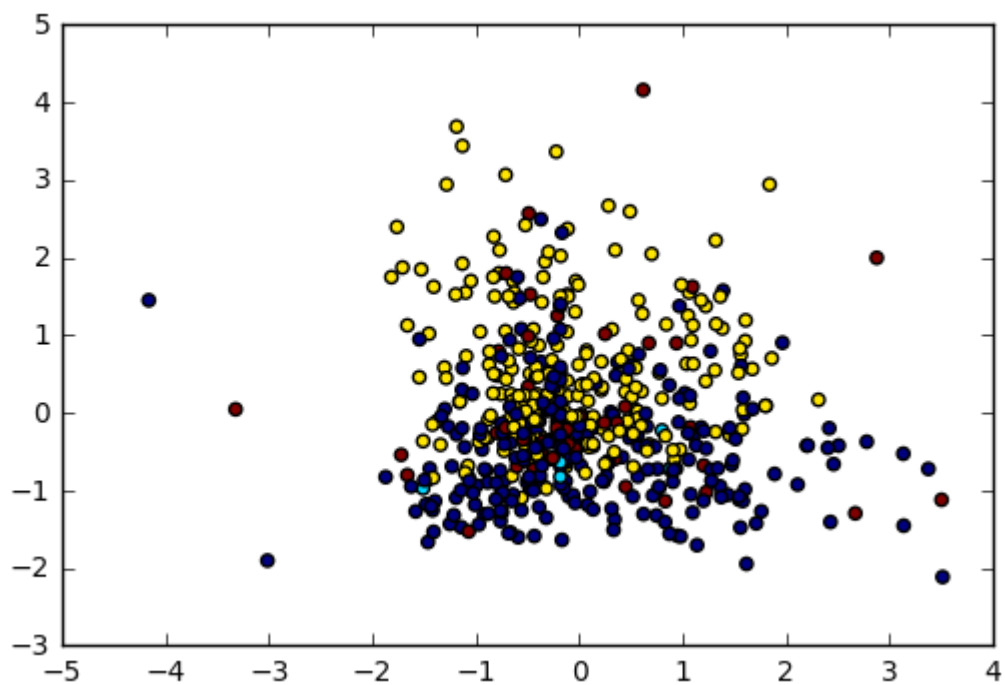
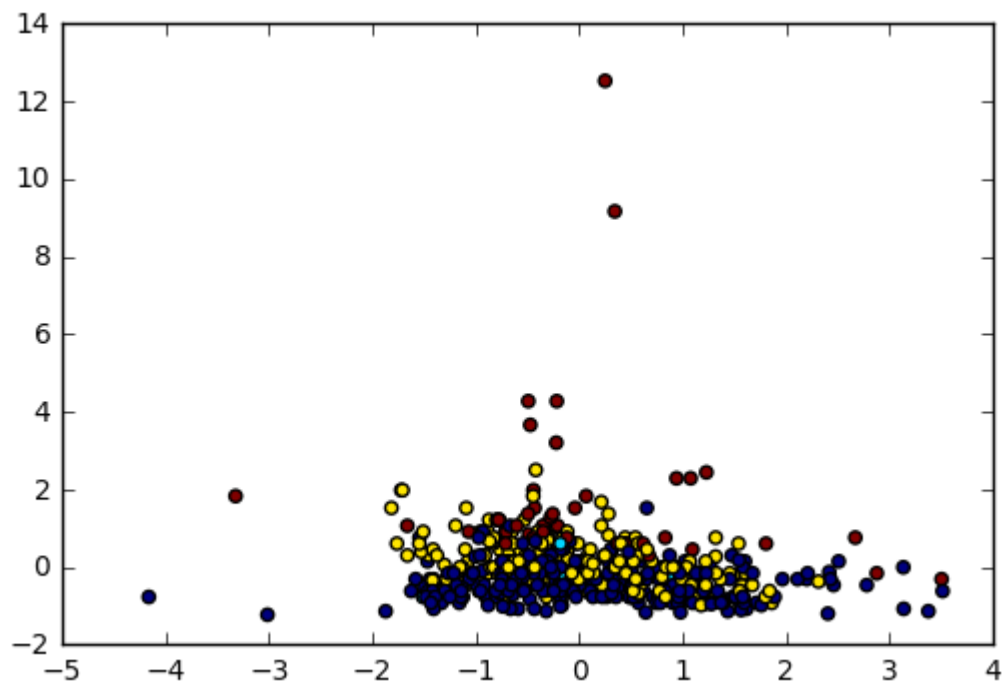


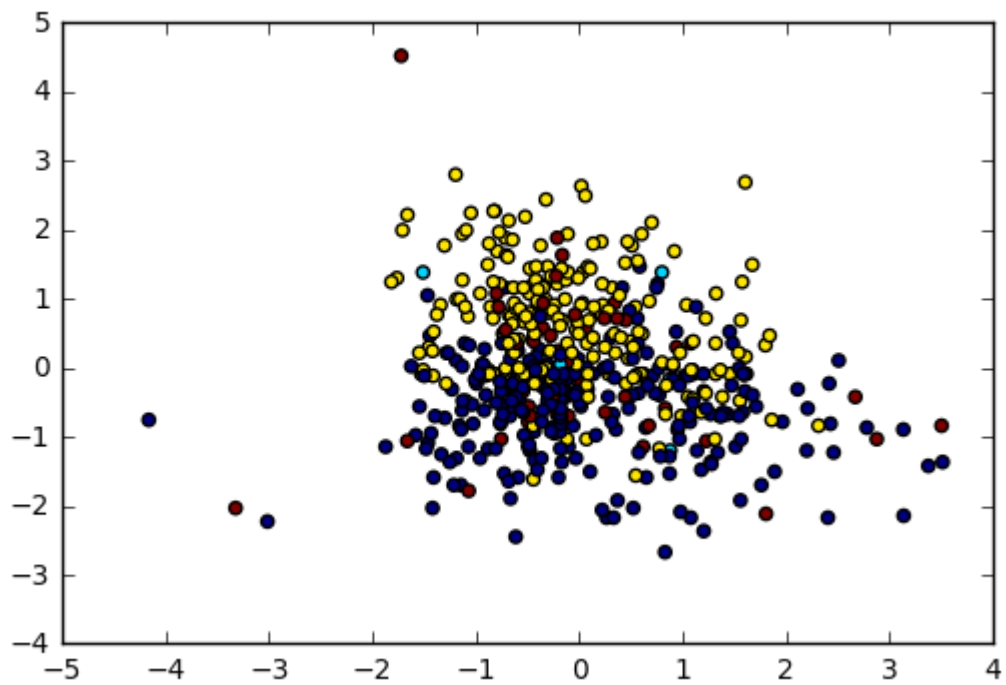
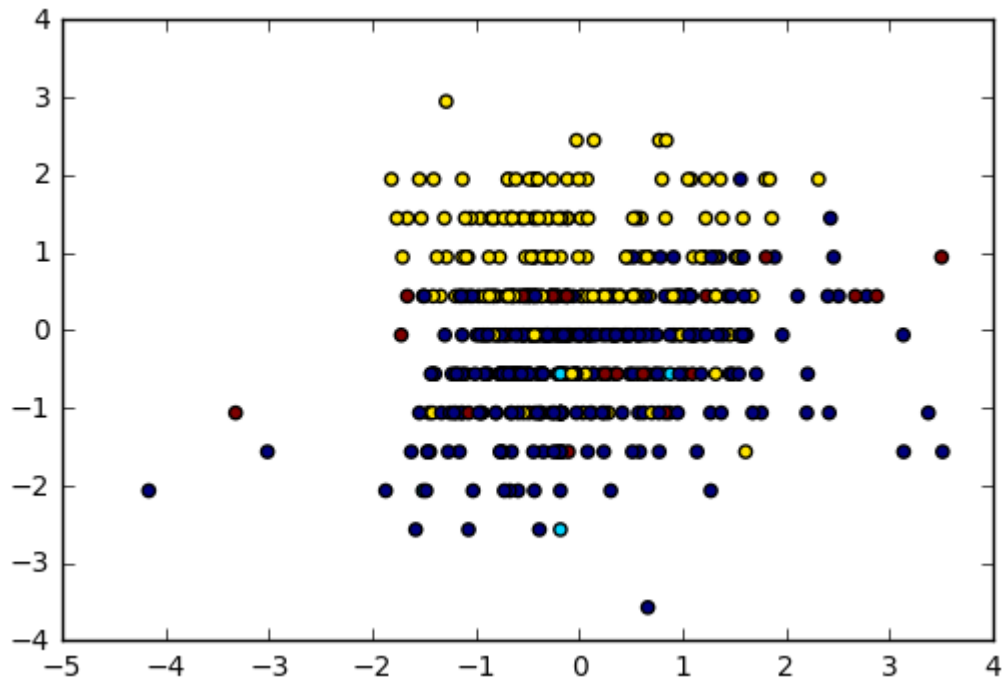


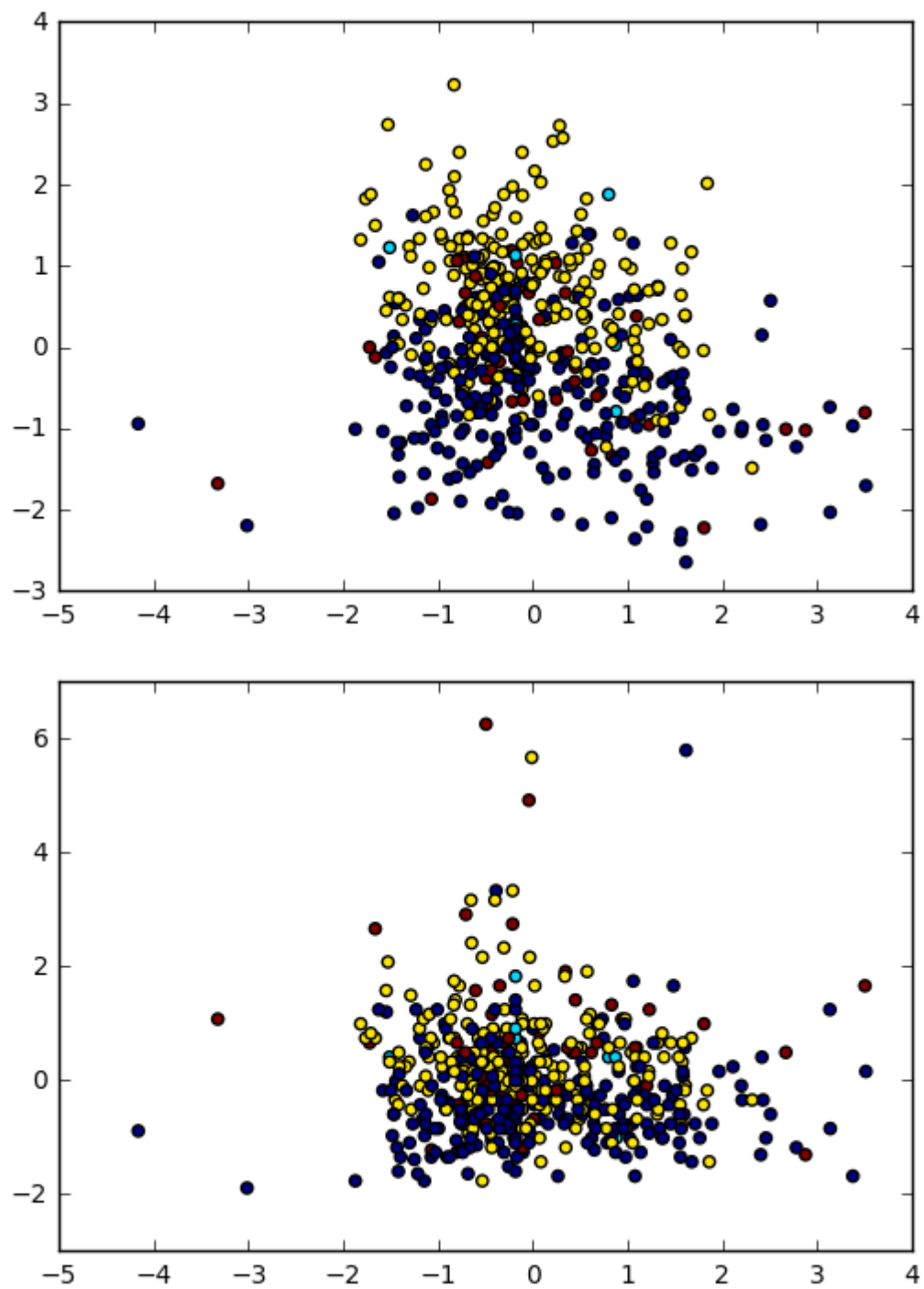


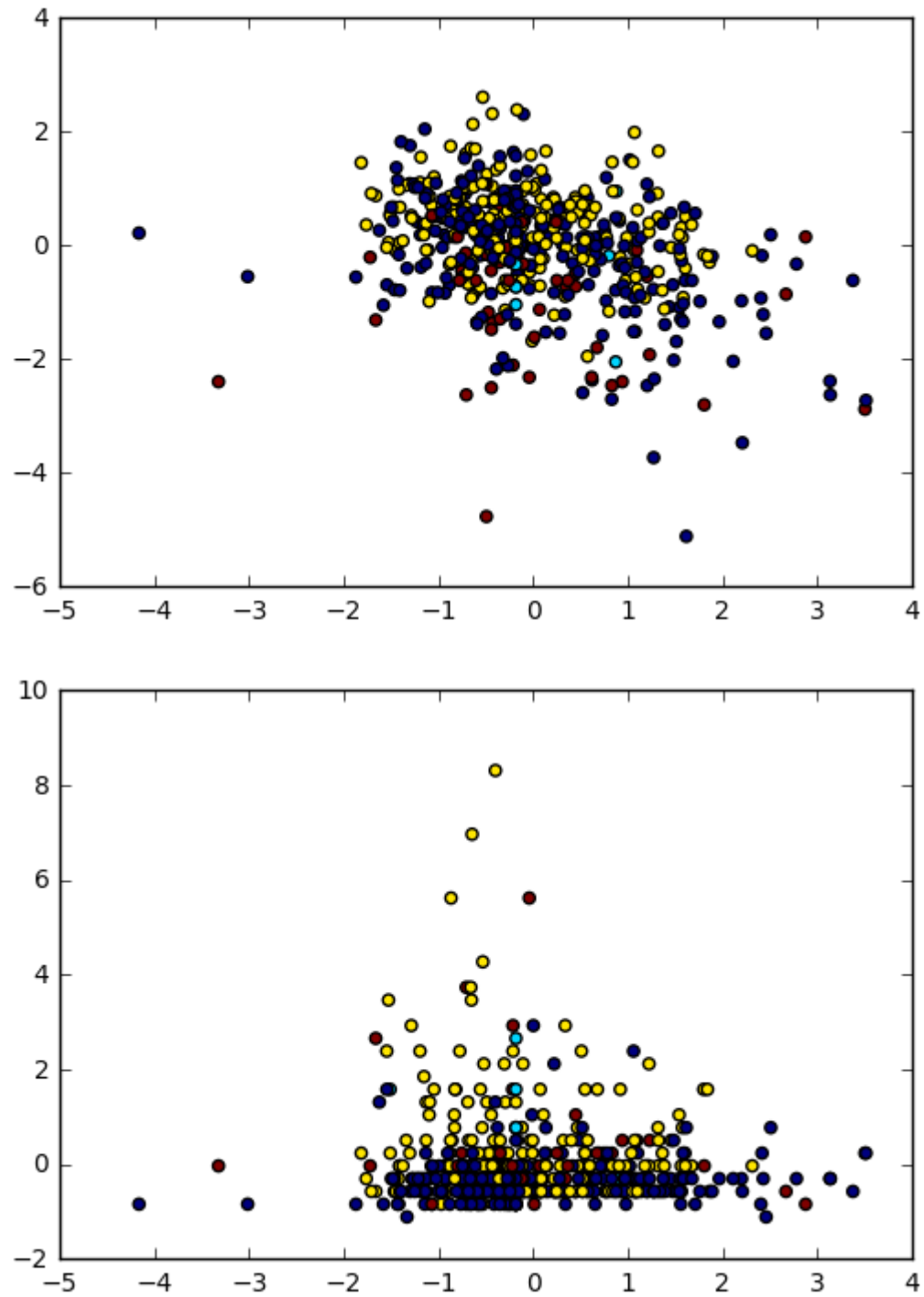


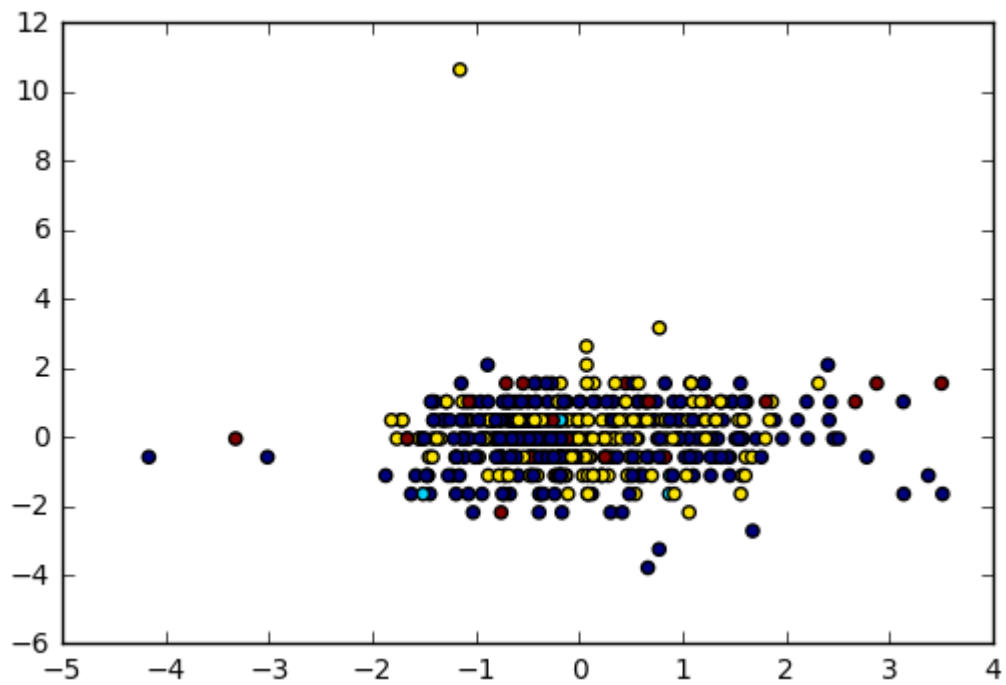
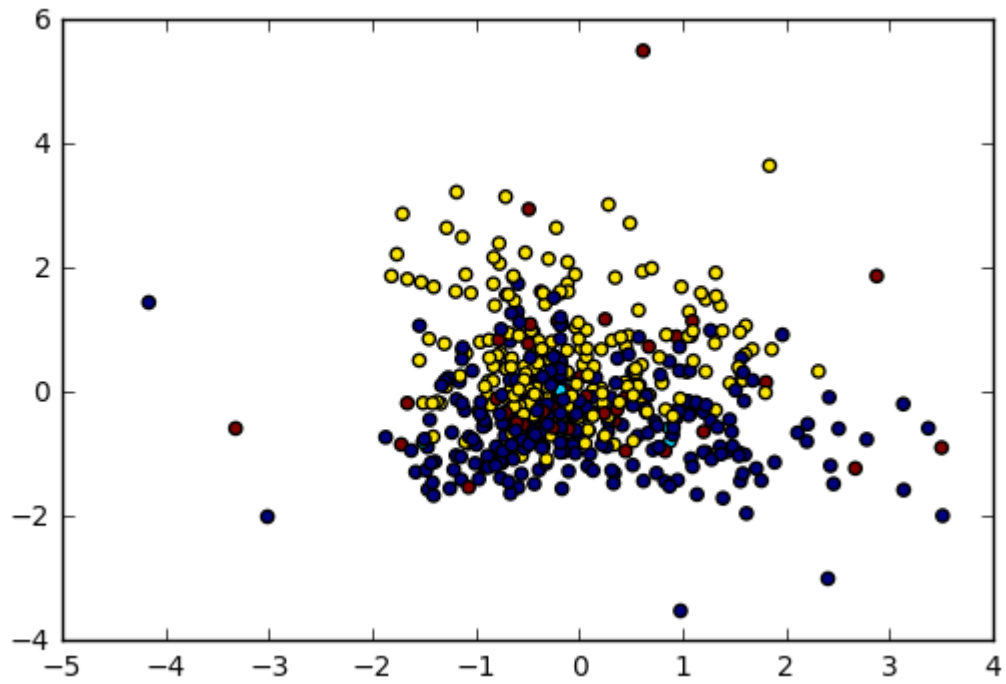


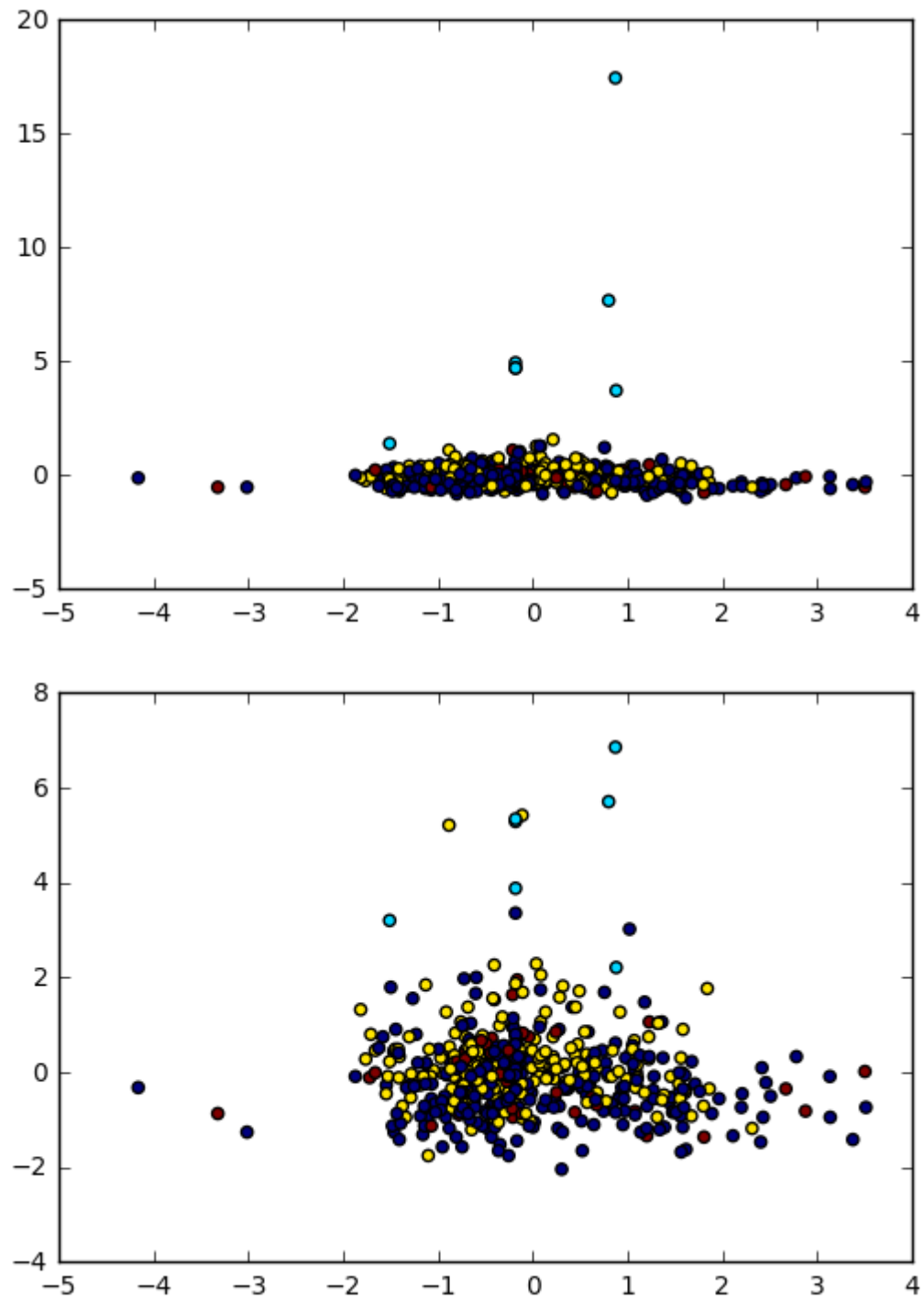


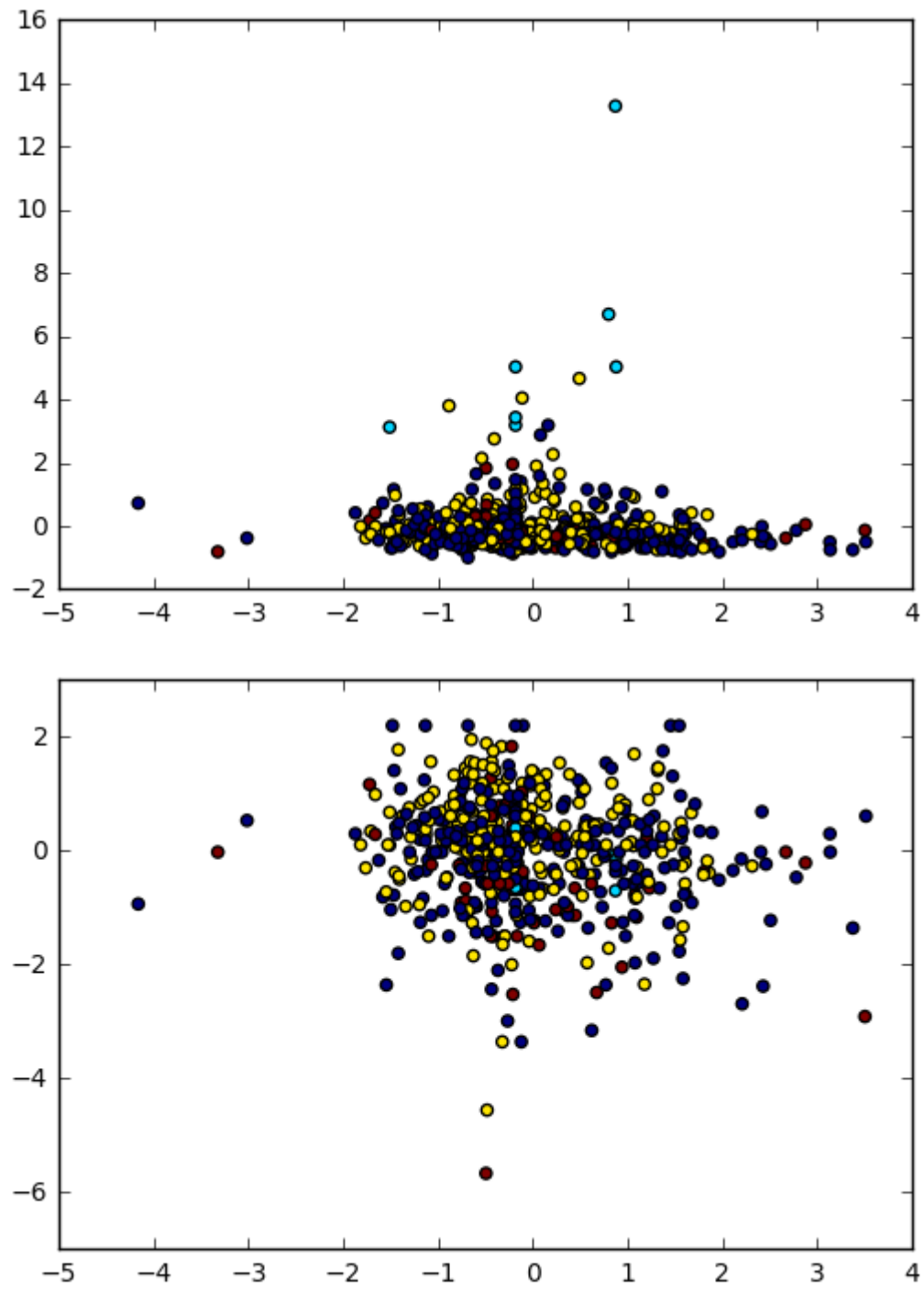


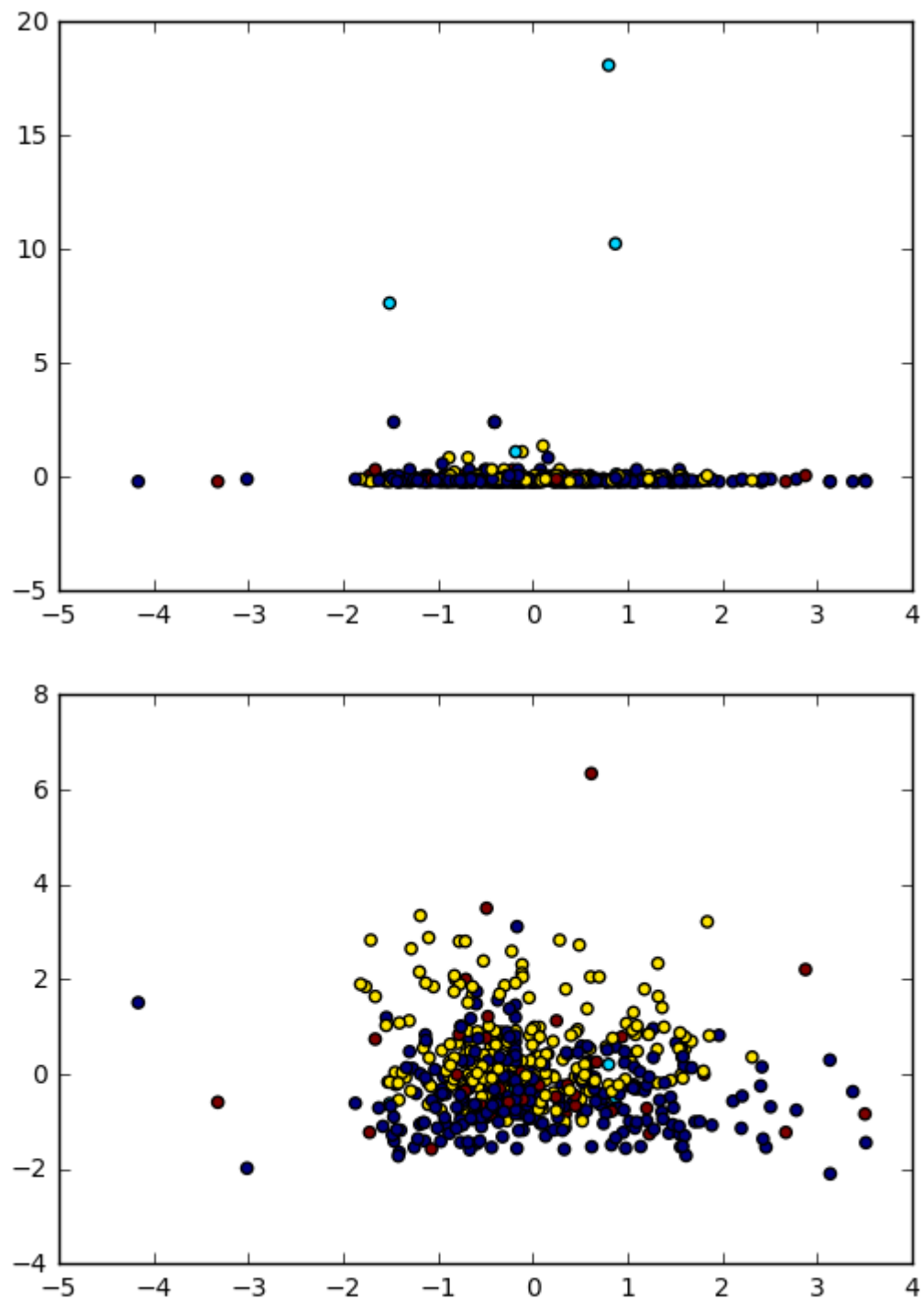


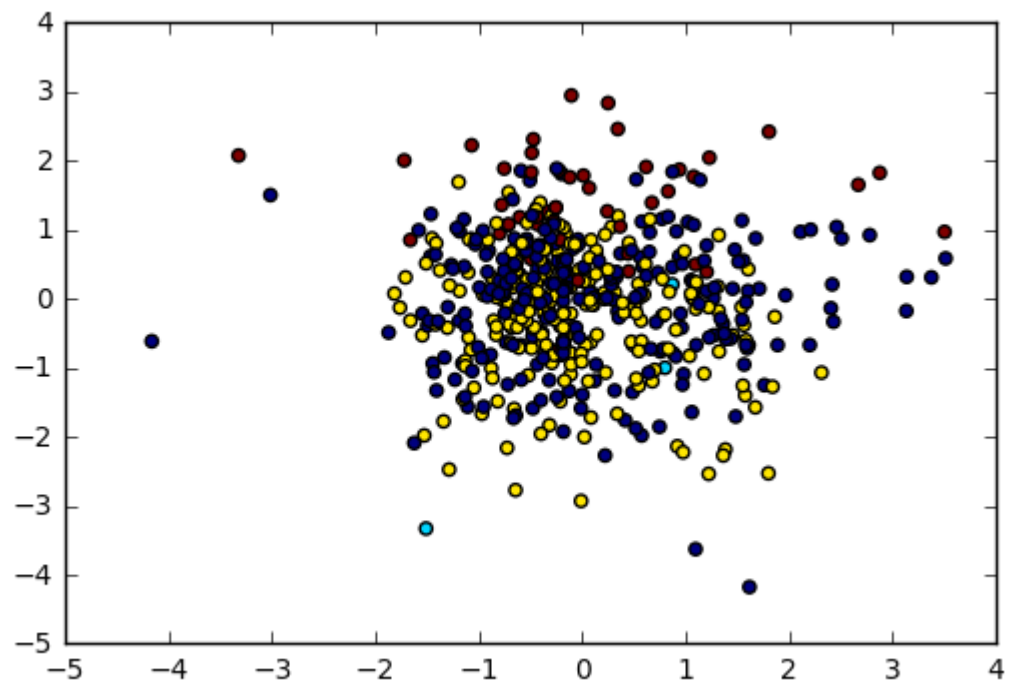
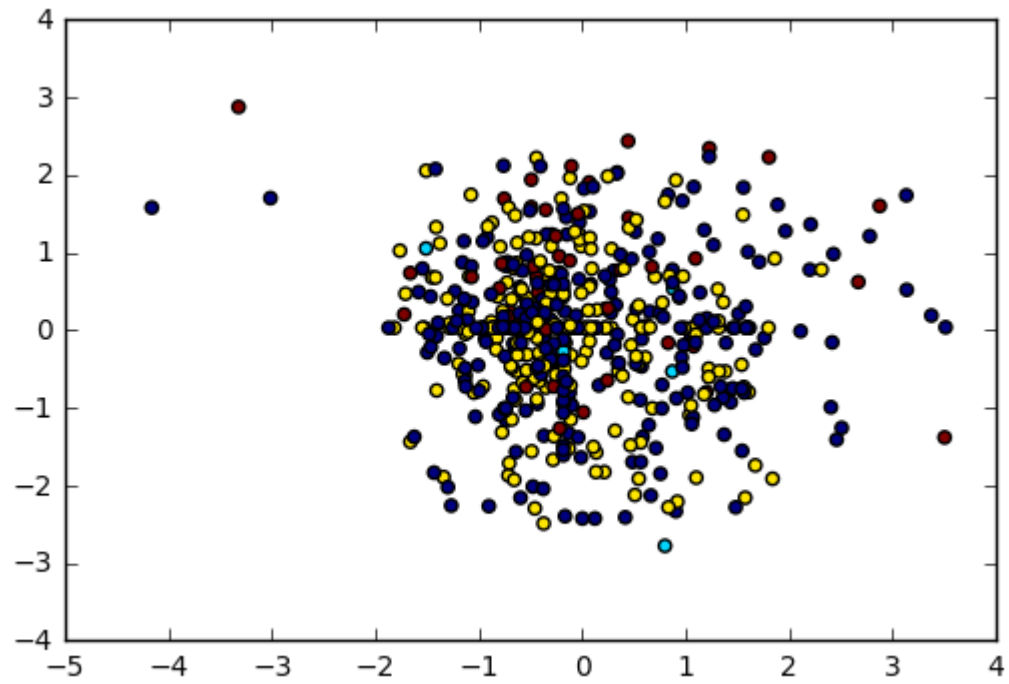


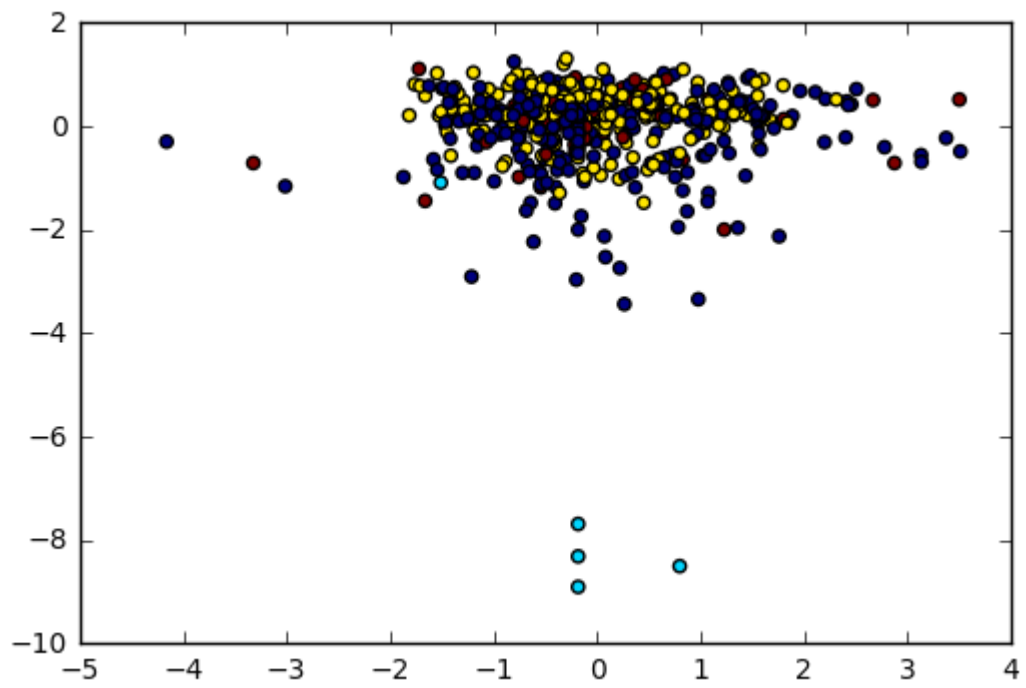
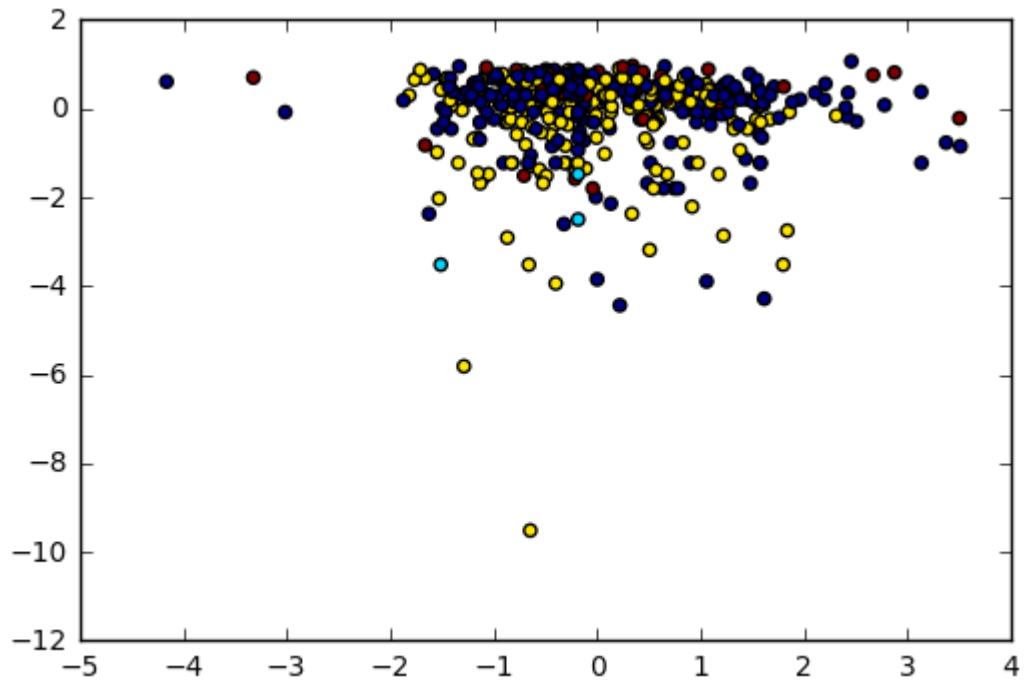


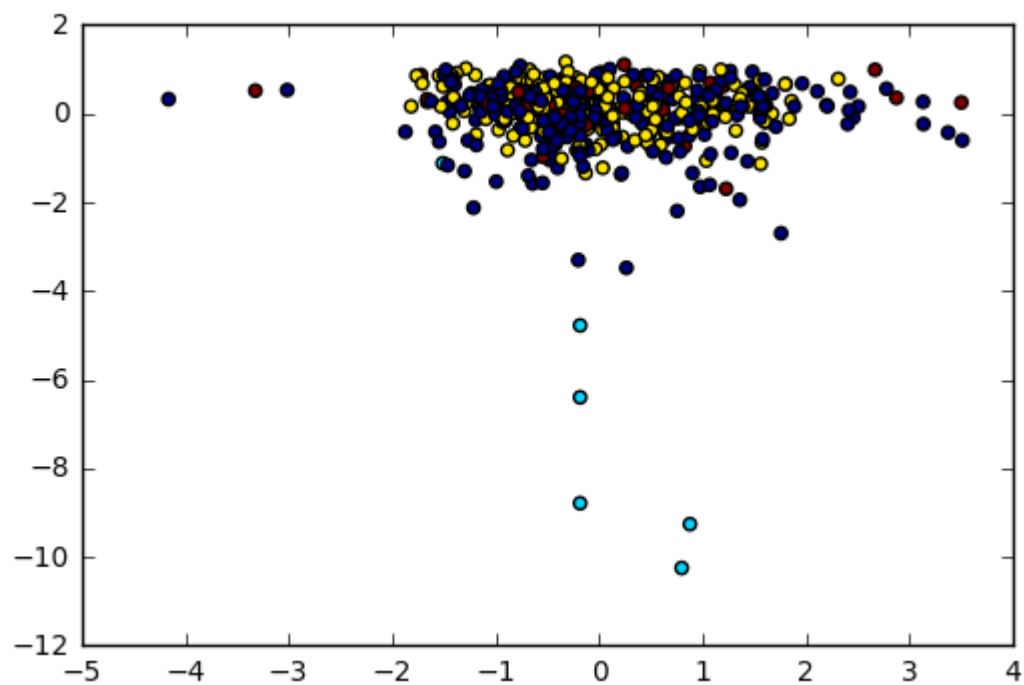
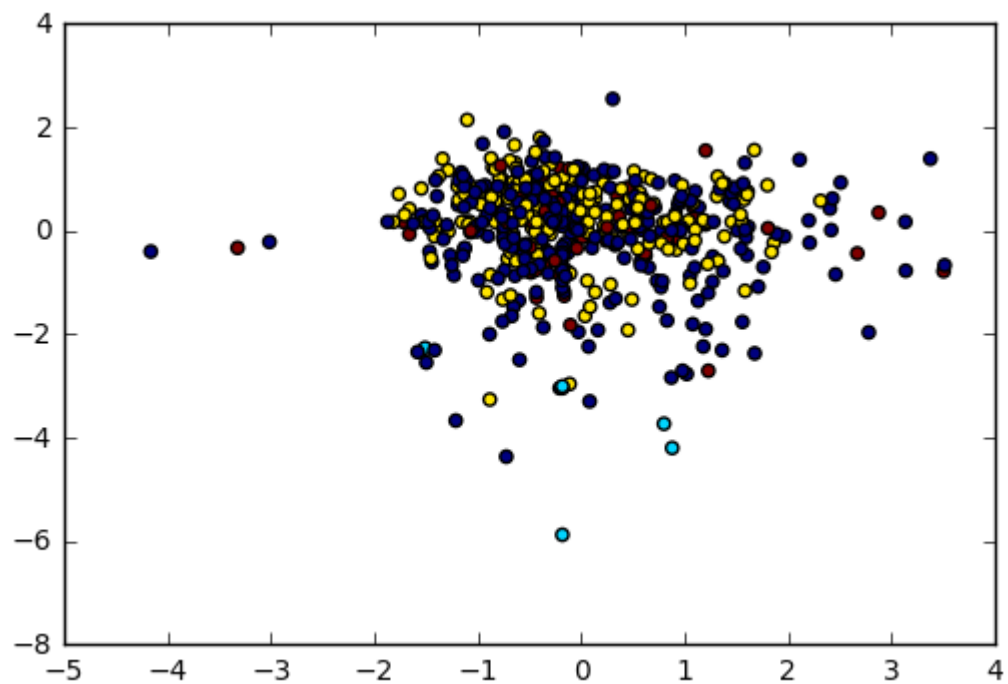


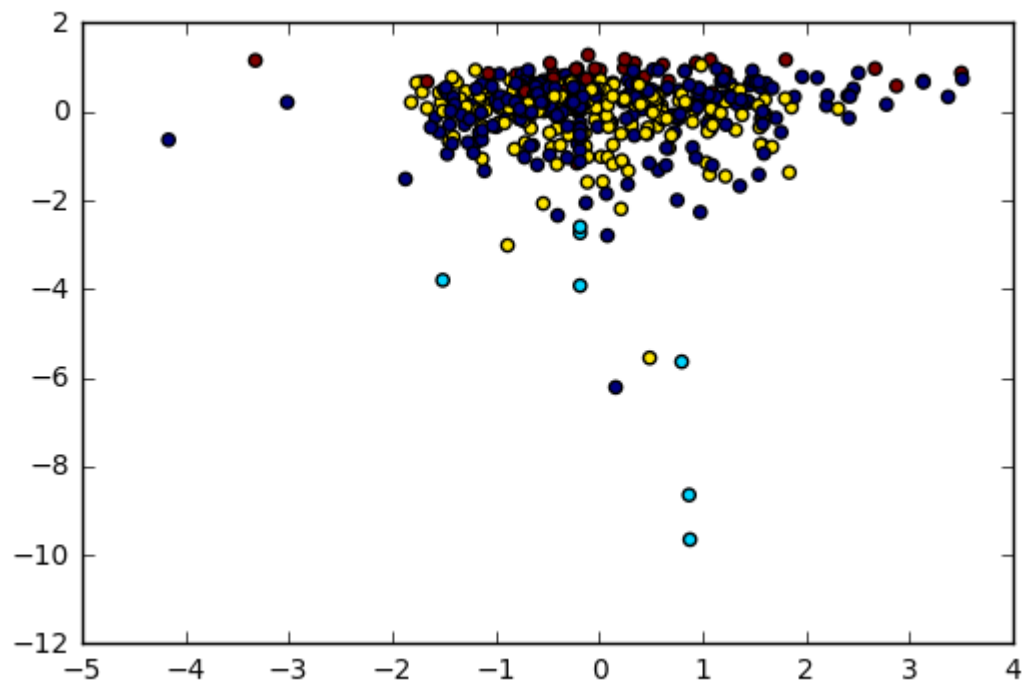
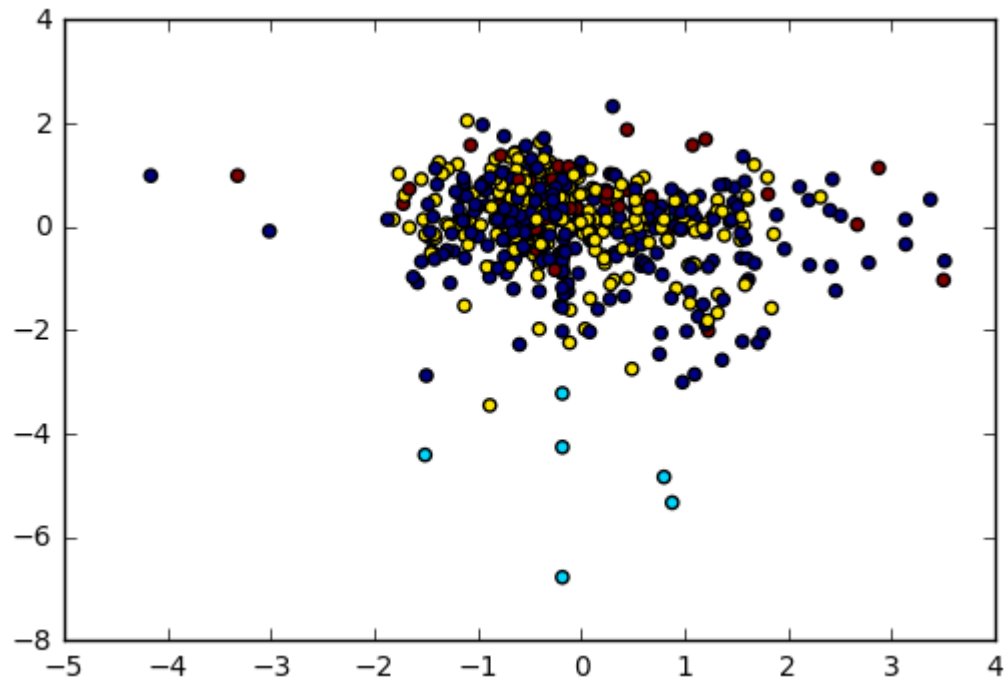


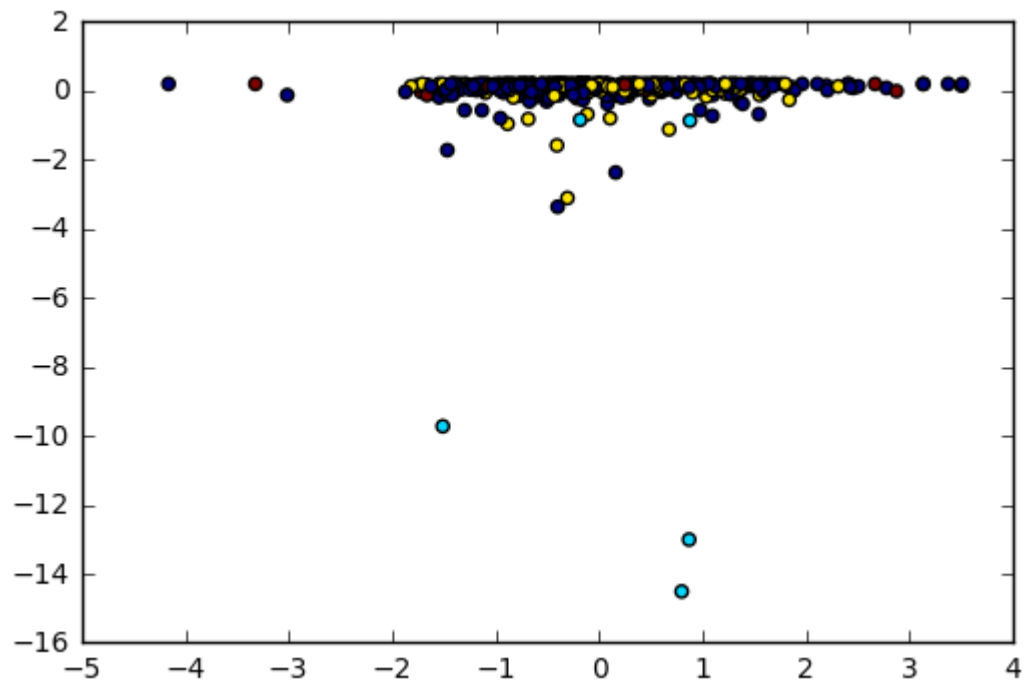




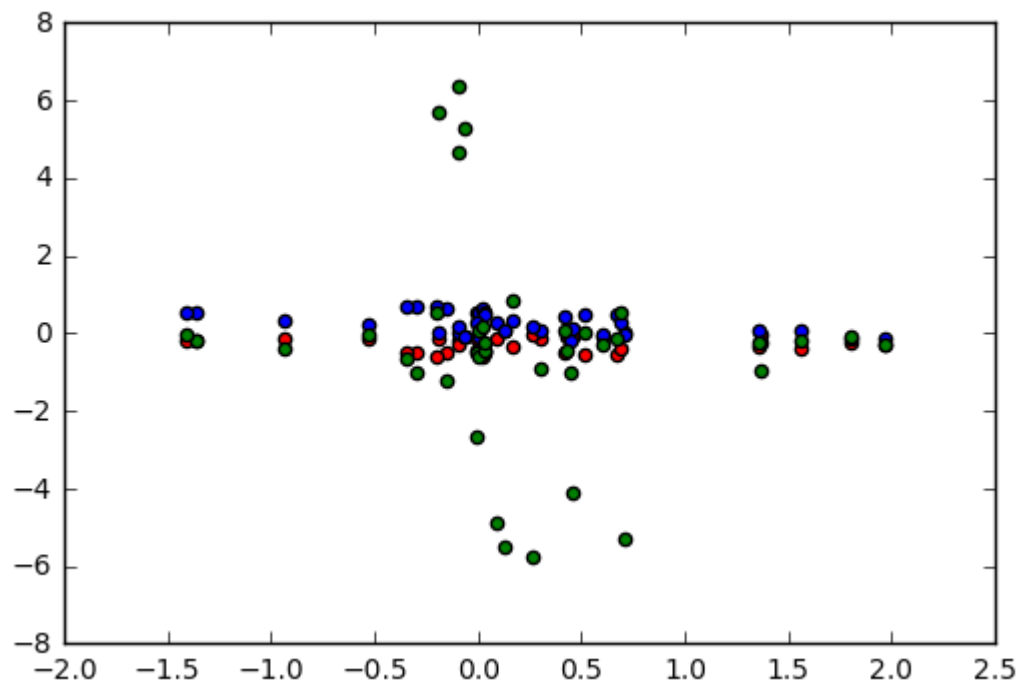








```
In [309]: plt.scatter(C[:,0], C[:,1], c='r')
plt.scatter(C[:,0], C[:,2], c='b')
plt.scatter(C[:,0], C[:,3], c='g')
plt.show()
```



```
In [364]: print('The End')
```

The End

```
In [ ]:
```