DBSCAN

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Quick Start

All programs are tested on macOS 10.14 & Ubuntu 18.04

Requirements

• Python3

How to Run

clustering.py

```
./clustering.py [input file] [number of clusters] [Eps] [MinPts]
python3 clustering.py [input file] [number of clusters] [Eps] [MinPts]
```

In order to execuate program using first command, you must have execution permission for clustering.py.

If it gives permission error, either give it a execution permission or use second line command.

evaluation.py

```
./evaluation.py
python3 evaluation.py
```

Running above command will give you two options to run evaluation program.

- 1. Use existing outputfiles at test/ folder
- 2. Run DBSCAN to create new outputFiles

Either way will use ideal & created output files to evaluate the accuracy. Note that in order to run first command, you **MUST** have output files ready.

Directory of the files may be modified with below global variables.

```
inputFileDir = "data/"
idealFileDir = "test/"
outputFileDir = "test/"
```

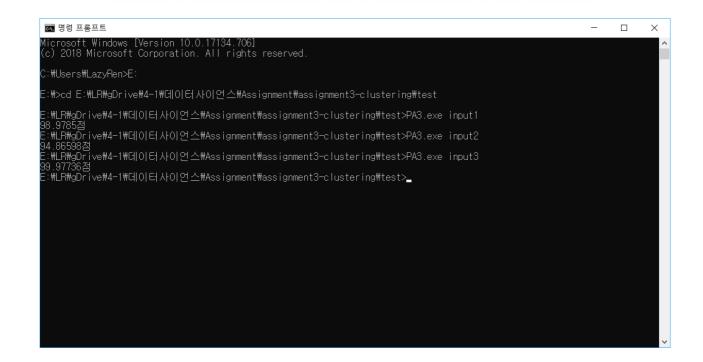
In order to create new tests, you must append appropriate values to below list, and ideal cluster files ready at idealFileDir.

Score of evaluation.py is equal to that of PA3.exe. But it gives extra information about clustered data.

```
🕽 🔵 🛑 assignment3-clustering — LazyRen@DaeIn-MBP — ..t3-clustering — -zsh 🛭 -zsh...
~/GoogleDrive/4-1/데이터사이언스/Assignment/assignment3-clustering master*
[> ./evaluation.py

    Use existing outputfiles at test/

2. Run DBSCAN to create new outputFiles
Input: 1
Starting Evaluation: input1
7676 points are clustered in ideal files
7485 points are clustered in output files
29155750 / 29456650 = 98.97850%
Evaluation took 10.975946 seconds
Starting Evaluation: input2
2000 points are clustered in ideal files
1942 points are clustered in output files
1896371 / 1999000 = 94.86598%
Evaluation took 0.735864 seconds
Starting Evaluation: input3
2100 points are clustered in ideal files
2099 points are clustered in output files
2203451 / 2203950 = 99.97736%
Evaluation took 0.814148 seconds
```



Results

Note that output files will be located on the same folder as input file. (in data folder in example) < If evaluation.py is used to create output files(by choosing option 2), it will generate output files at test folder. >

Result score may differ from each excution because DBSCAN algorithm choose *random point* to generate *density-based clustering*.

```
**GoogleDrive/4-1/데이터사이언스/Assignment/assignment...

**GoogleDrive/4-1/데이터사이언스/Assignment/assignment3-clustering master* †

**Time python3 ./clustering.py data/input1.txt 8 15 22

python3 ./clustering.py data/input1.txt 8 15 22 41.23s user 0.04s system 99% cp u 41.306 total

**GoogleDrive/4-1/데이터사이언스/Assignment/assignment3-clustering master* † 41s

**Time python3 ./clustering.py data/input2.txt 5 2 7

python3 ./clustering.py data/input2.txt 5 2 7 2.84s user 0.02s system 99% cpu 2

.876 total

**GoogleDrive/4-1/데이터사이언스/Assignment/assignment3-clustering master* †

**Time python3 ./clustering.py data/input3.txt 4 5 5

python3 ./clustering.py data/input3.txt 4 5 5 4.43s user 0.01s system 99% cpu 4

.449 total

**GoogleDrive/4-1/데이터사이언스/Assignment/assignment3-clustering master* †

**Time python3 ./clustering.py data/input3.txt 4 5 5 4.43s user 0.01s system 99% cpu 4

.449 total
```

```
E:\LR\gDrive\4-1\G|이터사이언스\Assignment\assignment3-clustering\test>pa3.exe input1
98.9785점
E:\LR\gDrive\4-1\G|이터사이언스\Assignment\assignment3-clustering\test>pa3.exe input2
94.86598점
E:\LR\gDrive\4-1\G|이터사이언스\Assignment\assignment3-clustering\test>pa3.exe input3
99.97736점
```

Implementation

class Point

```
class Point:
    def __init__(self, id, x, y):
        self.id = int(id)
        self.x = float(x)
        self.y = float(y)
        self.isVisited = False
        self.label = -1

def __repr__(self):
    return self.__str__()
```

```
def __str__(self):
    ret = "("
    ret += str(self.id) + ": "
    ret += str(self.x) + ", "
    ret += str(self.y) + ")"
    return ret
```

Each line from input file will be converted to a Point object.

id: unique number that represents Point object.

x: x-coordinate in 2D

y: y-coordinate in 2D

isVisited: boolean value that is used while DBSCAN algorithm.

label: id of cluster that it belongs to.

```
{isVisited, label} == {False, -1} --> not yet identified.
```

{isVisited, label} == {True, -1} --> outlier

Point may become {True, -1} after labelConverter is used to reduce # of clusters.

loadData()

Simply read line to line from input file and generate Point object.

Return value is list of all Point objects that will be used to clustering.

findNeighbor()

Find all neighbor Points from cur point. Neighbor must be positioned within radius of eps. Time Complexity = O(n)

dbscan()

Algorithm: DBSCAN: a density-based clustering algorithm.

Input:

- \blacksquare D: a data set containing n objects,
- \bullet : the radius parameter, and
- MinPts: the neighborhood density threshold.

Output: A set of density-based clusters.

Method:

```
(1) mark all objects as unvisited;
(2)
     do
(3)
           randomly select an unvisited object p;
(4)
           mark p as visited;
           if the \epsilon-neighborhood of \boldsymbol{p} has at least MinPts objects
(5)
                 create a new cluster C, and add p to C;
(6)
                 let N be the set of objects in the \epsilon-neighborhood of p;
(7)
(8)
                 for each point p' in N
(9)
                       if p' is unvisited
                            mark p' as visited;
(10)
                            if the \epsilon-neighborhood of p' has at least MinPts points,
(11)
                            add those points to N;
(12)
                       if p' is not yet a member of any cluster, add p' to C;
                 end for
(13)
                 output C;
(14)
(15)
           else mark p as noise;
(16) until no object is unvisited;
```

pseudo code of DBSCAN ¹

Implementation of DBSCAN algorithm. *Time Complexity* = $O(n^2)$

After creating *density-based clusters*, function also generates <code>labelconverter</code> dictionary. Which is used to remove any extra clusters exceeding <code>N</code> (user given argument of max cluster #).

createOutputFile()

Generates N output files. Each file represents one *density-based cluster*.

labelConverter is used at this point to eliminate extra clusters(labels). Any *Point* with -1 as a label is either outlier or belongs to extra cluster.

Note that if number of created clusters is smaller than $\overline{\mathbb{N}}$, some output files are created as a blank file.