

Self Organizing Map

```
In [1]: # Importing the Libraries
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
```

```
In [2]: # Importing the dataset
dataset = pd.read_csv('Credit_Card_Applications.csv')
dataset.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 690 entries, 0 to 689
Data columns (total 16 columns):
CustomerID    690 non-null int64
A1            690 non-null int64
A2            690 non-null float64
A3            690 non-null float64
A4            690 non-null int64
A5            690 non-null int64
A6            690 non-null int64
A7            690 non-null float64
A8            690 non-null int64
A9            690 non-null int64
A10           690 non-null int64
A11           690 non-null int64
A12           690 non-null int64
A13           690 non-null int64
A14           690 non-null int64
Class         690 non-null int64
dtypes: float64(3), int64(13)
memory usage: 86.3 KB
```

```
In [3]: dataset.head()
```

```
Out[3]:
```

	CustomerID	A1	A2	A3	A4	A5	A6	A7	A8	A9	A10	A11	A12	A13	A14	Class
0	15776156	1	22.08	11.46	2	4	4	1.585	0	0	0	1	2	100	1213	0
1	15739548	0	22.67	7.00	2	8	4	0.165	0	0	0	0	2	160	1	0
2	15662854	0	29.58	1.75	1	4	4	1.250	0	0	0	1	2	280	1	0
3	15687688	0	21.67	11.50	1	5	3	0.000	1	1	11	1	2	0	1	1
4	15715750	1	20.17	8.17	2	6	4	1.960	1	1	14	0	2	60	159	1

```
In [4]: X = dataset.iloc[:, :-1].values
y = dataset.iloc[:, -1].values
```


In [8]: X[10]

Out[8]: array([0.31526975, 1. , 0.29699248, 0.0625 , 0.5 ,
 1. , 0.875 , 0.15789474, 1. , 1. ,
 0.05970149, 1. , 0.5 , 0.1265 , 0.00857])

```
In [9]: # Training the SOM
from minisom_new import MiniSom
som = MiniSom(x = 20, y = 20, input_len = 15, sigma = 1.0, learning_rate = 0.2)
...
```

Parameters

```

-----
x : int
    x dimension of the SOM.

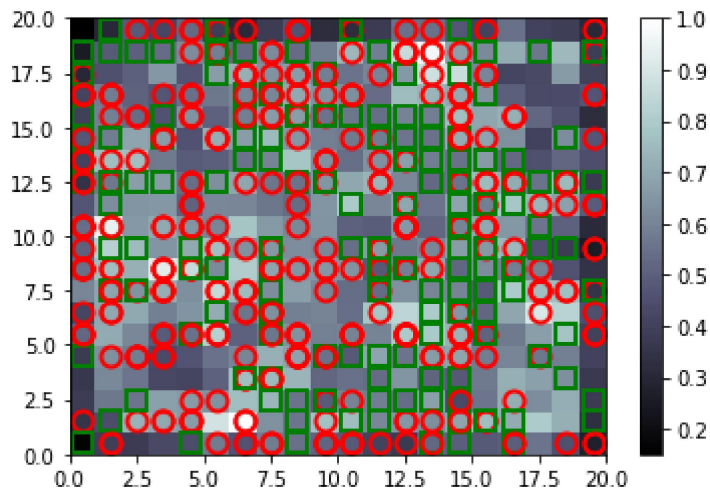
y : int
    y dimension of the SOM.

input_len : int
    Number of the elements of the vectors in input.

sigma : float, optional (default=1.0)
    Spread of the neighborhood function, needs to be adequate
    to the dimensions of the map.
    (at the iteration t we have  $\sigma(t) = \sigma / (1 + t/T)$ 
    where T is #num_iteration/2)
learning_rate, initial learning rate
    (at the iteration t we have
     $\text{learning\_rate}(t) = \text{learning\_rate} / (1 + t/T)$ 
    where T is #num_iteration/2)
random_seed : int, optional (default=None)
    Random seed to use.
...
som.random_weights_init(X)
"""Initializes the weights of the SOM
picking random samples from data"""
som.train_random(data = X, num_iteration = 100)
# """Trains the SOM picking samples at random from data"""

```

```
In [11]: # Visualizing the results
from pylab import bone, pcolor, colorbar, plot, show
bone()
pcolor(som.distance_map().T)
"""Returns the distance map of the weights.
    Each cell is the normalised sum of the distances between
    a neuron and its neighbours."""
colorbar()
markers = ['o', 's']
colors = ['r', 'g']
for i, x in enumerate(X):
    w = som.winner(x) # """Computes the coordinates of the winning neuron for the
    plot(w[0] + 0.5,
         w[1] + 0.5,
         markers[y[i]],
         markeredgecolor = colors[y[i]],
         markerfacecolor = 'None',
         markersize = 10,
         markeredgewidth = 2)
show()
```



```
In [12]: mappings = som.win_map(X)
"""Returns a dictionary wm where wm[(i,j)] is a list
    with all the patterns that have been mapped in the position i,j."""
```

In [13]: mappings

```
Out[13]: defaultdict(list,
                        {(9,
                          4): [array([0.84268147, 1.          , 0.12526316, 0.40928571, 0.5
                                ,
                                0.23076923, 0.375          , 0.05561404, 0.          , 0.
                                ,
                                0.          , 1.          , 0.5          , 0.05          , 0.01212
                                ]), array([0.59042402, 1.          , 0.20556391, 0.44642857, 0.5
                                ,
                                0.38461538, 0.5          , 0.00877193, 0.          , 0.
                                ,
                                0.          , 1.          , 0.5          , 0.36          , 0.
                                ]), array([0.89857005, 1.          , 0.05639098, 0.78571429, 1.
                                ,
                                0.          , 0.75          , 0.          , 0.          , 0.
                                ,
                                0.          , 1.          , 1.          , 0.225          , 1.
                                ]), array([0.67920025, 1.          , 0.19428571, 0.52089286, 0.5
                                ,
                                0.15384615, 0.5          , 0.          , 0.          , 0.
                                ,
                                0.          , 1.          , 0.5          , 0.089          , 0.
                                ]),
                          ...
                        }
```

In [17]: labels_maps = som.labels_map(X, y)

```
"""Returns a dictionary wm where wm[(i,j)] is a dictionary
    that contains the number of samples from a given label
    that have been mapped in position i,j.

    Parameters
    -----
    data : data matrix

    label : list or array that contains the label of each sample in data.
    """
```

In [18]: labels_maps

```
Out[18]: defaultdict(list,
                        {(9, 4): Counter({0: 3, 1: 1}),
                         (7, 6): Counter({0: 5, 1: 1}),
                         (8, 4): Counter({0: 5}),
                         (12, 15): Counter({1: 5}),
                         (17, 10): Counter({1: 1}),
                         (18, 5): Counter({1: 5}),
                         (3, 5): Counter({0: 3}),
                         (8, 0): Counter({1: 4}),
                         (5, 14): Counter({0: 3}),
                         (0, 1): Counter({0: 1}),
                         (14, 3): Counter({1: 4}),
                         (19, 12): Counter({1: 4}),
                         (16, 8): Counter({0: 1, 1: 1}),
                         (12, 4): Counter({1: 1}),
                         (14, 14): Counter({0: 4, 1: 1}),
                         (2, 12): Counter({1: 1}),
                         (15, 7): Counter({1: 2}),
                         (0, 14): Counter({1: 6, 0: 1}),
                         ...
                        }
```

In []: