Department of Computer Engineering Faculty of Engineering, University of Peradeniya ${\rm CO}542$

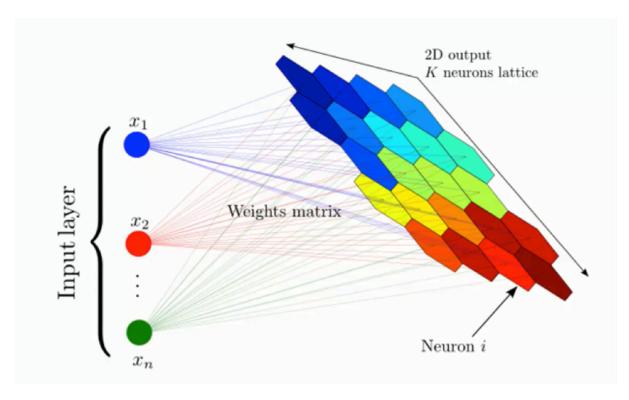
Neural Networks and Fuzzy Systems 2021

Lab 05 – Self Organizing Maps (SOM)

Objectives

- Implement and train Self Organizing Maps (SOM) using python libraries.
- Identify the usage of SOM for real world applications.
- Practice different data visualizations associated with SOM.

Self-Organizing Maps



 $Figure 1: Self-organizing \ map \ (Source: \ https://www.superdatascience.com/blogs/the-ultimate-guide-to-self-organizing-maps-soms)$

Minisom

MiniSom is a minimalistic and Numpy based implementation of the Self Organizing Maps (SOM). SOM is a type of Artificial Neural Network that is able to convert complex, nonlinear statistical relationships between high-dimensional data items into simple geometric relationships on a low-dimensional display. Minisom is designed to allow researchers to easily build on top of it and to give students the ability to quickly grasp its details.

Installation

• Using pip

```
pip install minisom
```

• Or Download MiniSom to a directory of your choice and use the setup script

```
git clone https://github.com/JustGlowing/minisom.git
python setup.py install
```

Usage

• In order to use MiniSom you need your data organized as a Numpy matrix where each row corresponds to an observation or as list of lists like the following:

• Then you can train MiniSom as follows:

```
from minisom import MiniSom
#initialization of 6x6 SOM
som = MiniSom(6, 6, 4, sigma=0.3,learning_rate=0.5)
# trains the SOM with 100 iterations
som.train(data, 100)
```

• You can obtain the position of the winning neuron on the map for a given sample as follows (E.g. For sample data[0]):

```
som.winner(data[0])
```

For more information refer to the,

- GitHub JustGlowing/minisom: MiniSom is a minimalistic implementation of the Self Organizing Maps.
- minisom/examples at master · JustGlowing/minisom · GitHub (Provides an overview of all the features implemented in minisom. Important: Follow these examples as a reference to complete the following tasks).

Exercise

The Pima Indians Diabetes Database contains several medical predictor variables and one target variable, 'Outcome'. Predictor variables include the number of pregnancies the patient has had, their BMI, insulin level, age, and so on.

- 1. Load the data set as a pandas' data frame and generate the feature matrix by selecting the appropriate features in the data set.
- 2. Use sklearn.preprocessing.MinMaxScaler normalize the data between 0 and 1.
- 3. Initialize the weights using MiniSom.pca_weights_init function (Also identify the usage of random_weights_init method and mention an advantage of using pca_weights_init).
- 4. Identify the difference between Minisom.train_batch and Minisom.train_random methods available minisom and use one of the models to train your self-organizing map.
- 5. Study about the distance map and mention what does it indicate related to the self-organizing maps.
- 6. Use Minisom.distance_map function to visualize the results obtained from the training process and use appropriate markers to indicate individual samples matched into each cell (according to the classes; patient who has diabetes/ do not have diabetes define two markers).
- 7. Generate another grid to indicate how often each neuron is activated using, MiniSom.activation_response.
- 8. Visualize the proportion of samples per class falling in a specific neuron using the, matplotlib.gridspec.GridSpec and matplotlib.patches.Patcha

Submission

Submit a single folder containing,

- All the required .py files or. ipynb files.
- A PDF containing all required outputs (E.g., Graphs, maps, etc...) and explanations before the deadline (announced in the course page).

Note: Please submit all the files (pdfs, code files etc.) in a single zipped file renamed as EXXYYY_Lab05, where XX is your batch ID, and YYY is your E No.

Reference

• GitHub - JustGlowing/minisom: MiniSom is a minimalistic implementation of the Self Organizing Maps