

## Shiraz University Neural Networks and Deep Learning (Spring 2021)

### Homework #2: Kohonen Self-Organizing Map (SOM)

Due date: 24th Ordibehesht 1400

In this homework, you have to train and test an SOM network to do cluster analysis of a news collection, from the BBC news website corresponding to stories in five topical areas from 2004-2005. This dataset is a collection of 2225 news document, categorized into 5 classes of 'business', 'entertainment', 'politics', 'sport', and 'tech'.

# Text Clustering using SOM

Text clustering is an unsupervised process, used to separate a document collection into some clusters on the basis of the similarity relationship between documents in the collection. Suppose  $D = \{d_1, ..., d_N\}$  be a collection of N documents to be clustered. The task is to divide D into k clusters  $C_1, ..., C_k$  where  $C_1 \cup ... \cup C_k = D$  and  $C_i \cap C_j = \emptyset$ , for  $i \neq j$ .

SOM text clustering can be done in two main phases. The first phase is document preprocessing, which uses Vector Space Model (VSM) to generate a numeric vector for each text document. In the next phase, SOM is applied on the document vectors to obtain document clusters.

# Phase 1: Document Preprocessing

By means of VSM, each document  $d_i$  can be represented by an n-dimensional feature vector  $v_i = \langle v_{i1}, ..., v_{in} \rangle$ , where  $v_{ij}$  is a representation of term  $t_j$  in document  $d_i$  and n is the number of distinct terms in the document collection D.

An approach for computing  $v_{ij}$  is the Term Frequency - Inverse Document Frequency (TF-IDF) weighting scheme. This method computes  $v_{ij}$  for term  $t_j$  in document  $d_i$  as:

$$v_{ij} = \log(1 + tf_{ij}) \times \log(\frac{N}{df_i})$$

where  $tf_{ij}$  is the frequency of term  $t_j$  in document  $d_i$ , and  $df_j$  is the number of documents in D containing term  $t_i$ .

Read 'bbc-text.csv' file and for each document:

1. Remove all non-letter characters from the documents.

- 2. Extract all words of the document and remove the short words (length  $\leq 2$ ).
- 3. Remove all stop words (e.g., 'a', 'and', 'what', ...), given in file 'stopwords.txt'.
- 4. Compute the feature vector for each document, using TF-IDF weighting scheme.

# Phase 2: SOM Clustering

- a) Winner-takes-all approach
  - 1. Using all documents, build an SOM with one neuron for each class.
  - 2. Depict the SOM-hits plot.
  - 3. Compute and report the confusion matrix.
- b) On-center, off-surround approach
  - 1. Using all documents, build an SOM with 3×3 neurons.
  - 2. Depict the SOM-hits plot.
  - 3. Compute the Euclidean distance of all documents to their winner neurons and sum up the distances.
  - 4. Repeat steps 1-3 for  $4\times4$  and  $5\times5$  topologies.
  - 5. Report and discuss the overall distances of three topologies.

#### **Notes:**

- Pay extra attention to the due date. It will not extend.
- Be advised that submissions after the deadline would not grade.
- Prepare your full report in PDF format and include the figures and results.
- You can use any library for SOM in Matlab or Python.
- Submit your assignment using a zipped file with the name of "StdNum\_FirstName\_LastName.zip" to soroushmehrpou@gmail.com with "NNDL-Spring 2021-HW#2" subject.