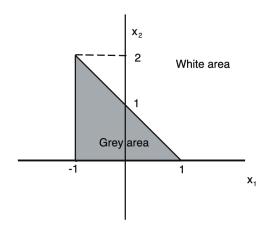
18.01.2022 Tuesday 15:00-18:00

- 1- A planar input pattern region of class 1 and 2 is shown in the figure below. The grey area in the figure below belongs to Category 1 and the rest of the pattern space belongs to Category 2
- i. Is this configuration linearly separable? Why?
- ii. If yes, then construct a single layer perceptron classifier. If no, then construct a multi-layer discreteperceptron classifier.



Category 1: White area Category 2: Grey area

2- Prototype points are given as

$$X_1=\begin{bmatrix}2\\-1\end{bmatrix},\ X_2=\begin{bmatrix}2\\1\end{bmatrix},\ X_3=\begin{bmatrix}1\\0\end{bmatrix},\ X_4=\begin{bmatrix}3\\0\end{bmatrix}: \text{Class 1}$$

$$X_5 = \begin{bmatrix} -2 \\ -1 \end{bmatrix}, \ X_6 = \begin{bmatrix} -2 \\ 1 \end{bmatrix}, \ X_7 = \begin{bmatrix} -3 \\ 0 \end{bmatrix}, \ X_8 = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$
 :Class 2

- i. Determine if the classes of patterns are linearly separable.
- ii. Determine the center of gravity for patterns of each class and find & draw the decision surface in pattern space.
- iii. Using equations (3.9) and (3.11) design the dichotomizer for the given prototype points and determine how it would recognize the following input patterns of unknown class membership.

$$P_1 = \begin{bmatrix} 5 \\ -2 \end{bmatrix}, P_2 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, P_3 = \begin{bmatrix} 12 \\ -8 \end{bmatrix}, P_4 = \begin{bmatrix} -2 \\ -4 \end{bmatrix}$$

3- The truncated energy function of a certain three-neuron single layer network with self-feedback is given as

$$E(\underline{\mathbf{V}}) = v_1^2 - v_2^2 + v_3^2 - 2v_1v_3 - v_2v_3 + 4v_1 + 12 \text{ where } \underline{\mathbf{V}} = \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix}$$

Assuming high gain neurons;

- i. Find the weight matrix $\underline{\mathbf{W}} \in R^{3x3}$ and the bias current vector $\underline{\mathbf{I}} \in R^3$
- ii. Find the gradient vector $\nabla E(\mathbf{V})$
- iii. Find the equilibrium point of Gradient Type Hopfield Neural Network
- iv. Find the Hessian matrix $\nabla^2 E(\mathbf{V})$
- v. Decide on the definiteness of the Hessian matrix
- vi. Find any minima, maxima or saddle points that the energy function may have, what are they?

- 4- An analog input value $x \in R$ is to be converted by a two neuron network proposed by Tank and Hopfield as an A/D converter. It is assumed the high gain neurons have continuous unipolar activation functions.
- i. Write the A/D conversion error E_c by defining every term in E_c
- ii. Write the truncated energy funtion E
- iii. Find the parameters of two neuron network
- iv. Write the state equations and the output equations for this network (choose $C_0=C_1=1F$, $g_0=g_1=1v$ and choose inverting neurons.)
- v. If the analog inputs are given as x = 0, x = 1, x = 1.5 then find the corresponding stationary points of the network and decide on the type of the stationary points (i.e maxima, minima or saddle points?)
- vi. Decide whether the truncated energy function has unconstrained minima
- vii. If this truncated energy function has no unconstrained minima then can the stationary points be reached by the network starting from some initial voltages on the capacitors?
- 5- The linear associator has to associate the following pairs of vectors;

$$S^{(1)} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} \end{bmatrix}^{t} \longrightarrow f^{(1)} = \begin{bmatrix} 0 & 1 & 1 \end{bmatrix}^{t}$$

$$S^{(2)} = \begin{bmatrix} \frac{1}{2} & \frac{-5}{6} & \frac{1}{6} & \frac{1}{6} \end{bmatrix}^{t} \longrightarrow f^{(2)} = \begin{bmatrix} 1 & 0 & -1 \end{bmatrix}^{t}$$

$$S^{(3)} = \begin{bmatrix} \frac{1}{2} & \frac{1}{6} & \frac{1}{6} & \frac{-5}{6} \end{bmatrix}^{t} \longrightarrow f^{(3)} = \begin{bmatrix} 1 & 0 & 0 \end{bmatrix}^{t}$$

- i. Verify that the vectors $S^{(i)}$, i = 1,2,3 are orthonormal,
- ii. Find the memory weight matrix
- iii. Veriy the association performed by the network for the stimuli $S^{\left(1\right)}$
- iv. Distort $S^{(1)}$ as $S^{(1)}+\Delta^{(1)}$ by choosing $\Delta^{(1)}$ orthogonal to $S^{(1)}$ and then find the corresponding response
- 6- The following vectors need to be stored in a recurrent autoassociative memory;

$$S^{(1)} = \begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix}^t, S^{(2)} = \begin{bmatrix} 1 & -1 & -1 & -1 & 1 \end{bmatrix}^t, S^{(3)} = \begin{bmatrix} -1 & 1 & 1 & 1 \end{bmatrix}^t$$

- i. Use the storage algorithm to calculate the weight matrix
- ii. Evaluate the energy function
- iii. Apply the input vector $V^{(0)} = \begin{bmatrix} 1 & -1 & -1 & 1 \end{bmatrix}^t$ and allow for asynchronous update sequence in retrieval algorithm in ascending node order starting at node 1 while evaluating energy values in each updating mode.
- iv. Discuss the convergence of this sequence and if it exists find its limit, determine the stored memory.