**Batch: A2 Roll No.: 16010121045**

**Experiment No. 9**

**Grade: AA / AB / BB / BC / CC / CD /DD**

**Signature of the Staff In-charge with date**

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| **Title: :** Case study on anyone Special Neural Networks |

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**Aim :** To understand the special neural network and their applications

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**Expected Outcome of Experiment:**

**CO3 :** Understand perceptron’s and counter propagation networks **\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_**

**Books/ Journals/ Websites referred:**

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**Discuss the algorithm and application of following special networks**

1. **Boltzmann Machine**:

**Algorithm**: A Boltzmann Machine is a type of stochastic recurrent neural network that consists of a network of binary units or neurons. It uses a stochastic approach for learning and making predictions. The key components and algorithms associated with Boltzmann Machines are:

1. **Units (Neurons)**: Each unit in a Boltzmann Machine can be in one of two states - on or off. These states are typically represented as binary values.
2. **Energy Function**: The network assigns an energy value to each possible state of the units. This energy function is used to measure how well the network is performing. The goal is to minimize this energy.
3. **Learning Algorithm**: Boltzmann Machines use a learning algorithm known as Contrastive Divergence. It involves updating the connections (weights) between units to minimize the energy function. This learning process is typically unsupervised, and the network learns to capture patterns in the data.
4. **Stochastic Update**: The units in a Boltzmann Machine are updated stochastically. During learning, they are updated in parallel, and during inference or prediction, they are updated sequentially using a probabilistic rule based on the current state and the energy function.

**Applications**: Boltzmann Machines have found applications in various areas, including:

1. **Collaborative Filtering**: They can be used in recommendation systems to suggest products, movies, or content to users based on their historical preferences.
2. **Dimensionality Reduction**: Boltzmann Machines can be used for unsupervised feature learning and dimensionality reduction in high-dimensional datasets.
3. **Restricted Boltzmann Machines (RBM)**: RBMs, a variant of Boltzmann Machines, are used in deep learning and feature learning for tasks like image recognition, topic modeling, and more.
4. **Generating New Data**: They can generate new data samples that are similar to the training data, making them useful for data augmentation.
5. **Support Vector Machine (SVM)**:

**Algorithm**: A Support Vector Machine is a supervised machine learning algorithm used for classification and regression tasks. The algorithm's main objective is to find the optimal hyperplane that maximizes the margin between different classes in the data. Here are the key components and algorithms associated with SVMs:

1. **Linear Separability**: SVMs work well when data is linearly separable, meaning it can be divided into two classes by a straight line or hyperplane.
2. **Margin Maximization**: The SVM algorithm aims to find the hyperplane that maximizes the margin between the nearest data points (support vectors) of different classes. This is done by solving an optimization problem.
3. **Kernel Trick**: SVMs can be extended to handle non-linear data by using kernel functions. Kernel functions map the data into a higher-dimensional space, where it might become linearly separable. Common kernel functions include the polynomial kernel and radial basis function (RBF) kernel.
4. **Regularization**: SVMs also include regularization parameters to control overfitting, and they can handle soft-margin cases where the data is not perfectly separable.

**Applications**: Support Vector Machines have a wide range of applications in both classification and regression problems:

1. **Image Classification**: SVMs can be used for tasks like object recognition and image classification.
2. **Text Classification**: They are applied in spam email detection, sentiment analysis, and text categorization.
3. **Bioinformatics**: SVMs are used for protein structure prediction, gene classification, and disease classification based on genetic data.
4. **Finance**: They can be used in financial markets for tasks like stock price prediction and credit scoring.
5. **Anomaly Detection**: SVMs are effective in anomaly detection for fraud detection, network security, and quality control.
6. **Regression**: SVMs can be employed for regression tasks to predict continuous values, such as housing price prediction.

These are two powerful machine learning techniques with distinct applications, and their choice depends on the specific problem and the nature of the data.

**Date: \_\_\_\_\_\_\_\_ Signature of faculty in-charge**