### **UNIT 1 – Fundamentals of Deep Learning**

## 1. What is Artificial Intelligence? List its main branches.

**Answer:** Artificial Intelligence (AI) is the simulation of human intelligence in machines. Branches include:

- Machine Learning
- Natural Language Processing
- Computer Vision
- Robotics
- Expert Systems

## 2. Explain the evolution of Machine Learning.

**Answer:** ML evolved from:

- Probabilistic Modeling (using probabilities),
- Early Neural Networks (Perceptrons),
- Kernel Methods (SVMs),
- Ensemble Models (Random Forests),
- to Deep Learning (neural networks with many layers).

## 3. What is Probabilistic Modeling? Give an example.

**Answer:** It uses probability to predict outcomes. Example: Naive Bayes for spam email classification.

# 4. What are Perceptrons?

**Answer:** Early neural models with a single-layer that adjust weights based on inputs to learn binary classification.

### 5. What are Kernel Methods?

**Answer:** They transform data into higher dimensions using kernels (like RBF), useful in SVM for non-linear classification.

## 6. Compare Decision Trees, Random Forests, and Gradient Boosting.

- Decision Tree: single model
- Random Forest: multiple trees (bagging)
- Gradient Boosting: sequential tree learning (boosting)

## 7. Name four branches of ML with examples.

#### Answer:

- Supervised (e.g., regression)
- Unsupervised (e.g., clustering)
- Reinforcement (e.g., game AI)
- Semi-supervised (e.g., limited labeled data)

## 8. Explain Supervised Learning.

**Answer:** Input-output mapping using labeled data. Example: spam detection. Includes diagram with input  $\rightarrow$  model  $\rightarrow$  output.

### 9. How do we evaluate ML models?

#### Answer:

- Accuracy = (TP + TN) / Total
- Precision = TP / (TP + FP)
- **Recall** = TP / (TP + FN)
- **F1-Score** = 2 \* (Precision \* Recall) / (Precision + Recall)

## 10. What is Overfitting and Underfitting?

### Answer:

- Overfitting: model memorizes training data (high variance)
- Underfitting: model fails to learn patterns (high bias)
- Solutions: Regularization, cross-validation

## UNIT 2 – Introducing Deep Learning

## 1. Define Deep Learning.

**Answer:** A subset of ML that uses deep neural networks to model complex patterns in large datasets.

## 2. Biological vs Machine Vision?

- Biological: Human eye + brain
- Machine: Cameras + neural networks
  Example: Face detection by human vs. using CNN

### 3. Human vs Machine Language Understanding.

#### **Answer:**

- Human: contextual, emotional
- Machine: uses embeddings, RNNs, transformers

#### 4. What is an ANN?

**Answer:** Artificial Neural Network consists of input, hidden, and output layers. Diagram: nodes with connections.

## 5. Single vs Multi-layer Neural Network?

#### Answer:

- Single: learns simple patterns
- Multi-layer (deep): learns complex, hierarchical patterns

## 6. Explain training of deep neural networks.

### Answer:

Forward pass → loss → backpropagation → weight update via gradient descent

# 7. Challenges in deep learning training?

**Answer:** Overfitting, vanishing gradients, computational cost, large data requirements

## 8. Techniques to improve DL models?

**Answer:** Dropout, Batch Normalization, Data Augmentation

## 9. Importance of Activation Functions?

**Answer:** Adds non-linearity.

Compare:

• ReLU: fast, sparse activation

• Sigmoid: saturates

• Tanh: zero-centered

# 10. Optimization algorithms in DL?

### Answer:

• **SGD:** updates weights using gradients

• Adam: combines momentum + RMSprop

#### UNIT 3 – Neural Networks and Tools

## 1. Define Neural Network with diagram.

**Answer:** Mimics brain neurons. Diagram: input  $\rightarrow$  hidden  $\rightarrow$  output layers

## 2. Layers in NN?

#### Answer:

- Input Layer
- Hidden Layer(s)
- Output Layer

Examples: Dense, Convolution, Recurrent

## 3. Activation Functions?

Answer: Introduce non-linearity. Examples: ReLU, Sigmoid, Tanh

# 4. Compare Keras, TensorFlow, Theano, CNTK.

### **Answer:**

Keras: high-level

• TensorFlow: flexible, Google-backed

• Theano: older, research

• CNTK: Microsoft, less used now

# 5. Setting up DL workstation?

### Answer:

- Install Python, libraries (Keras/TensorFlow)
- Set up GPU drivers
- Use Jupyter or IDE

# 6. What is binary classification?

**Answer:** Classify into 2 classes.

Example: Positive/Negative sentiment in IMDB reviews

# 7. Preprocessing IMDB reviews?

- Tokenization
- Padding
- Vectorization
- Label encoding

#### 8. Define multiclass classification.

**Answer:** More than two classes.

Example: Classifying news articles into topics

## 9. Classifying newswires using DL?

#### Answer:

• Dataset: Reuters

• Architecture: Embedding → Dense → Softmax

## 10. Compiling and training a model in Keras?

#### Answer:

- model.compile() → define loss, optimizer
- model.fit() → train
- model.evaluate() → test

#### UNIT 4 – CNN and RNN

# 1. What is a CNN? Main components?

### **Answer:**

- Convolution layer
- Pooling layer
- Fully connected layer
  Used in image classification.

## 2. Representation learning in CNN?

**Answer:** CNNs automatically learn filters to extract meaningful features.

# 3. Convolution layer working?

Answer: Applies filters across image patches to detect patterns. Diagram: sliding filter.

# 4. Multichannel convolution?

**Answer:** Applies filters across all color channels (RGB). Each filter spans all channels.

## 5. Define RNN. Difference from FFNN?

- RNN: has memory via feedback loops
- FFNN: no memory, input → output directly

## 6. RNN working step-by-step?

#### **Answer:**

- Inputs processed sequentially
- Hidden state updated at each step

# 7. Simple RNN in Python?

import torch.nn as nn

model = nn.RNN(input\_size=10, hidden\_size=20, num\_layers=1)

## 8. What are PyTorch Tensors?

Answer: Multi-dimensional arrays for GPU/CPU. Operations: reshape, add, matmul.

## 9. Building CNN in PyTorch?

### Answer:

- Define model
- Use loss = nn.CrossEntropyLoss()
- Optimizer: torch.optim.Adam()

# 10. NumPy vs PyTorch Tensors?

#### Answer:

• NumPy: CPU only

• PyTorch: GPU support, auto-diff

## UNIT 5 – Applications and Research Models

## 1. What is Machine Vision? Example?

**Answer:** Ability of machines to see and interpret visual data. Example: Object detection in self-driving cars.

# 2. NLP Workflow in DL?

### **Answer:**

Tokenization → Embedding → RNN/Transformer → Output (e.g., sentiment)

## 3. What is GAN? Architecture?

- Generator creates fake data
- Discriminator checks real vs fake

• Trained together in a loop

Diagram: Generator → Fake data → Discriminator

## 4. Types of Deep Reinforcement Learning?

#### Answer:

- DQN
- Policy Gradient
- Actor-Critic

Difference: learns through environment feedback

## 5. Agent-Environment interaction?

**Answer:** Agent takes action → Environment gives reward → Agent updates strategy

#### 6. What is an Autoencoder?

**Answer:** Neural network for unsupervised learning. Encoder compresses → Decoder

reconstructs

### 7. How do Boltzmann Machines work?

#### Answer:

- Models energy of system
- Learns by lowering energy of data-like configurations

### 8. What is RBM?

**Answer:** A type of Boltzmann Machine with restricted connections. Faster and easier to train.

### 9. Structure of DBN?

**Answer:** Stack of RBMs. Each layer learns features and passes them up.

# 10. Compare Autoencoders, RBMs, and DBNs.

Mo	del	Structure	Learning	Use Case
Aut	toencoder	Encoder–Decoder	Backpropagation	Dimensionality Reduction
RBI	М	Bipartite graph	Energy Minimization	Feature learning
DBI	N	Stack of RBMs	Layer-wise training	Pre-training deep nets