

DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING



CS19P18- DEEP LEARNING CONCEPTS LABORATORY LAB MANUAL

FINAL YEAR

SEVENTH SEMESTER

2025-2026

ODD SEMESTER

LIST OF EXPERIMENTS

- 1. Create a neural network to recognize handwritten digits using MNIST dataset
- 2. Build a Convolutional Neural Network with Keras/TensorFlow
- 3. Image Classification on CIFAR-10 Dataset using Convolutional Neural Networks
- 4. Transfer learning with CNN and Visualization
- 5. Build a Recurrent Neural Network using Keras/Tensorflow
- 6. Sentiment Classification of Text using Recurrent Neural Network (RNN)
- 7. Build autoencoders with Keras/TensorFlow
- 8. Build GAN with Keras/TensorFlow
- 9. Perform object detection with YOLO3
- 10. Mini Project CNN based or RNN based applications

RAJALAKSHMI ENGINEERING COLLEGE DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING CS19P18- DEEP LEARNING CONCEPTS LABORATORY

LAB PLAN

Sl.No.	Name of the Experiment	Hours		
		Planned		
1	Create a neural network to recognize handwritten	2		
	digits using MNIST dataset			
2	Build a Convolutional Neural Network with	2		
	Keras/TensorFlow			
3	Image Classification on CIFAR-10 Dataset using	2		
	CNN			
4	Transfer learning with CNN and Visualization	2		
5	Build a Recurrent Neural Network using	2		
	Keras/Tensorflow			
6	Sentiment Classification of Text using RNN	2		
7	Build autoencoders with Keras/TensorFlow	2		
8	Perform object detection with YOLO3	2		
0	Perform object detection with 4 OLO3	2		
9	Build GAN with Keras/TensorFlow	2		
	Build OAIV with Kelas/ Telisoff IOW	2		
10	Mini Project – CNN or RNN based applications	8		

HARDWARE AND SOFTWARE REQUIREMENTS

Hardware Requirements	Core i5 and above/M1 Chip with minimum 8 GB RAM and 512 GB HDD
Software Requirements	Windows 10/Apple MacOS, Tensorflow, Keras, Numpy, Pandas and Scikit-learn

Course Outcomes (COs)

Course Name: Deep Learning Concepts Course Code: CS19P18

Outcome 1	Understand the fundamentals of deep learning based on optimizations and backpropagation and machine learning.
Outcome 2	Train neural network models that converge well without overfitting.
Outcome 3	Learn how to improve the deep learning model performance using error analysis, regularization, hyper parameter tuning.
Outcome 4	Build networks to perform sentiment analysis and work on real-time time series data.
Outcome 5	Analyse different supervised, unsupervised, and reinforcement deep learning models and their applications in real world scenarios; Build, train, test and evaluate neural networks for different applications and data types.

CO-PO -PSO matrices of course

PO/PSO															
	PO 1	PO 2	PO 3	PO 4	PO 5	PO 6	PO 7	PO 8	PO 9	PO 10	PO 11	PO 12	PSO 1	PSO 2	PSO 3
CO															
CS19P18.1	3	2	2	-	1	-	-	-	-	-	1	1	2	1	1
CS19P18.2	2	2	2	-	2	-	-	-	-	-	1	2	3	2	2
CS19P18.3	3	3	1	3	2	-	-	-	-	-	1	2	2	2	2
CS19P18.4	2	1	3	-	2	1	1	1	-	1	2	3	3	3	3
CS19P18.5	3	1	1	3	2	2	1	1	1	2	3	3	2	3	3
Average	2.6	1.8	1.8	3.0	1.8	1.5	1.0	1.0	1.0	1.5	1.6	2.2	2.4	2.2	2.2

Note: Enter correlation levels 1,2 or 3 as defined below:

1: Slight (Low) 2: Moderate (Medium) 3: Substantial (High) If there is no correlation, put "-"

INSTALLATION AND CONFIGURATION OF TENSORFLOW

Aim:

To install and configure TensorFlow in anaconda environment in Windows 10.

Procedure:

- 1. Download Anaconda Navigator and install.
- 2. Open Anaconda prompt
- 3. Create a new environment dlc with python 3.7 using the following command: conda create -n dlc python=3.7
- 4. Activate newly created environment dlc using the following command: conda activate dlc
- 5. In dlc prompt, install tensorflow using the following command: pip install tensorflow
- 6. Next install Tensorflow-datasets using the following command: pip install tensorflow-datasets
- 7. Install scikit-learn package using the following command: pip install scikit-learn
- 8. Install pandas package using the following command: pip install pandas
- 9. Lastly, install jupyter notebook pip install jupyter notebook
- 10. Open jupyter notebook by typing the following in dlc prompt: jupyter notebook
- 11. Click create new and then choose python 3 (ipykernel)
- 12. Give the name to the file
- 13. Type the code and click Run button to execute (eg. Type import tensorflow and then run)

- 1. https://docs.anaconda.com/free/anaconda/applications/tensorflow/
- $2. \underline{https://conda.io/projects/conda/en/latest/user-guide/tasks/manage-environments.\underline{html} \#activating-an-environment$

EX NO: 1 CREATE A NEURAL NETWORK TO RECOGNIZE HANDWRITTEN DIGITS USING MNIST DATASET

Aim:

To build a handwritten digit's recognition with MNIST dataset.

Procedure:

- 1. Download and load the MNIST dataset.
- 2. Perform analysis and preprocessing of the dataset.
- 3. Build a simple neural network model using Keras/TensorFlow.
- 4. Compile and fit the model.
- 5. Perform prediction with the test dataset.
- 6. Calculate performance metrics.

- 1. https://www.analyticsvidhya.com/blog/2022/07/handwritten-digit-recognition-using-tensorflow/
- 2. https://www.milindsoorya.com/blog/handwritten-digits-classification

EX NO:2 BUILD A CONVOLUTIONAL NEURAL NETWORK USING KERAS/TENSORFLOW

Aim:

To implement a Convolutional Neural Network (CNN) using Keras/TensorFlow to recognize and classify handwritten digits from the MNIST dataset with high accuracy.

Procedure:

- 1. Import required libraries (TensorFlow/Keras, NumPy, etc.).
- 2. Load the MNIST dataset from Keras.
- 3. Normalize and reshape the image data.
- 4. Convert labels to one-hot encoded vectors.
- 5. Build a CNN model with Conv2D, MaxPooling, Flatten, and Dense layers.
- 6. Compile the model using categorical crossentropy and Adam optimizer.
- 7. Train the model on training data.
- 8. Evaluate the model on test data.
- 9. Display accuracy and predictions.

- 1. https://towardsdatascience.com/build-your-first-cnn-with-tensorflow-a9d7394eaa2e
- 2. https://www.analyticsvidhya.com/blog/2021/06/building-a-convolutional-neural-network-using-tensorflow-keras/

EX NO: 3 IMAGE CLASSIFICATION ON CIFAR-10 DATASET USING CNN

Aim:

To build a Convolutional Neural Network (CNN) model for classifying images from the CIFAR-10 dataset into one of the ten categories such as airplanes, cars, birds, cats, etc.

Procedure:

- 1. Download and load the CIFAR-10 dataset using Keras/TensorFlow.
- 2. Visualize and analyze sample images from the dataset.
- 3, Preprocess the data:
 - Normalize the pixel values (divide by 255)
 - Convert class labels to one-hot encoded format
- 4. Build a CNN model using Keras/TensorFlow:
 - Include convolutional, pooling, flatten, and dense layers.
- 5. Compile the model with suitable loss function and optimizer.
- 6. Train the model using training data and validate using test data.
- 7. Evaluate the model using accuracy and loss on test dataset.
- 8. Perform predictions on new/unseen CIFAR-10 images.
- 9 Visualize prediction results with sample images and predicted labels.

- 1. https://www.analyticsvidhya.com/blog/2021/01/image-classification-using-convolutional-neural-networks-a-step-by-step-guide/
- 2. https://www.geeksforgeeks.org/deep-learning/cifar-10-image-classification-in-tensorflow/

Ex No: 4 TRANSFER LEARNING WITH CNN AND VISUALIZATION

Aim:

To build a convolutional neural network with transfer learning and perform visualization

Procedure:

- 1. Download and load the dataset.
- 2. Perform analysis and preprocessing of the dataset.
- 3. Build a simple neural network model using Keras/TensorFlow.
- 4. Compile and fit the model.
- 5. Perform prediction with the test dataset.
- 6. Calculate performance metrics.

- $1. \underline{https://medium.com/analytics-vidhya/car-brand-classification-using-vgg 16-transfer-learning-f219a0f09765}$
- 2. https://www.kaggle.com/code/kasmithh/transfer-learning-using-keras-vgg-16

EX NO: 5 BUILD A RECURRENT NEURAL NETWORK (RNN) USING KERAS/TENSORFLOW

Aim:

To build a recurrent neural network with Keras/TensorFlow.

Procedure:

- 1. Download and load the dataset.
- 2. Perform analysis and preprocessing of the dataset.
- 3. Build a simple neural network model using Keras/TensorFlow.
- 4. Compile and fit the model.
- 5. Perform prediction with the test dataset.
- 6. Calculate performance metrics.

- 1. https://machinelearningmastery.com/understanding-simple-recurrent-neural-networks-in-keras/
- 2. https://victorzhou.com/blog/keras-rnn-tutorial/

EX NO: 6 SENTIMENT CLASSIFICATION OF TEXT USING RNN

Aim:

To implement a Recurrent Neural Network (RNN) using Keras/TensorFlow for classifying the sentiment of text data (e.g., movie reviews) as positive or negative.

Procedure:

- 1. Import necessary libraries.
- 2. Load and preprocess the text dataset (e.g., IMDb).
- 3. Pad sequences and prepare labels.
- 4. Build an RNN model with Embedding and SimpleRNN layers.
- 5. Compile the model with loss and optimizer.
- 6. Train the model on training data.
- 7. Evaluate the model on test data.
- 8. Predict sentiment for new inputs

- $1. \underline{https://medium.com/@muhammadluay45/sentiment-analysis-using-recurrent-neural-network-rnn-long-short-term-memory-lstm-and-38d6e670173\underline{f}}$
- 2. https://www.ijert.org/text-classification-using-rnn

Ex No: 7 BUILD AUTOENCODERS WITH KERAS/TENSORFLOW

Aim:

To build autoencoders with Keras/TensorFlow.

Procedure:

- 1. Download and load the dataset.
- 2. Perform analysis and preprocessing of the dataset.
- 3. Build a simple neural network model using Keras/TensorFlow.
- 4. Compile and fit the model.
- 5. Perform prediction with the test dataset.
- 6. Calculate performance metrics.

- 1. https://blog.keras.io/building-autoencoders-in-keras.html
- 2. https://towardsdatascience.com/how-to-make-an-autoencoder-2f2d99cd5103

Ex No: 8 OBJECT DETECTION WITH YOLO3

Aim:

To build an object detection model with YOLO3 using Keras/TensorFlow.

Procedure:

- 1. Download and load the dataset.
- 2. Perform analysis and preprocessing of the dataset.
- 3. Build a simple neural network model using Keras/TensorFlow.
- 4. Compile and fit the model.
- 5. Perform prediction with the test dataset.
- 6. Calculate performance metrics.

- 1. https://machinelearningmastery.com/how-to-perform-object-detection-with-yolov3-in-keras/
- 2. https://www.kaggle.com/code/roobansappani/yolo-v3-object-detection-using-keras

Ex No: 9 BUILD GENERATIVE ADVERSARIAL NEURAL NETWORK

Aim:

To build a generative adversarial neural network using Keras/TensorFlow.

Procedure:

- 1. Download and load the dataset.
- 2. Perform analysis and preprocessing of the dataset.
- 3. Build a simple neural network model using Keras/TensorFlow.
- 4. Compile and fit the model.
- 5. Perform prediction with the test dataset.
- 6. Calculate performance metrics.

- 1. https://pyimagesearch.com/2020/11/16/gans-with-keras-and-tensorflow/
- $2. \underline{https://www.analyticsvidhya.com/blog/2021/06/a-detailed-explanation-of-gan-with-\underline{implementation-using-tensorflow-and-keras/}$

Ex No: 10 MINI PROJECT – CNN OR RNN BASED APPLICATION

Aim:

To develop an application that is based on convolutional neural network or recurrent neural network in Keras/TensorFlow.

Instructions:

- 1. Student has to choose one of the projects listed in the below section with a team of maximum two members.
- 2. Apart from these topics, if a team has innovative ideas to do implementation, then they can discuss with their faculty in-charge and get approval to do the same.
- 3. After implementation, a Mini Project report needs to be prepared and signed by HoD and the faculty in-charge.

Mini Project Titles;

- 1. Plant Disease Detection using Deep Learning
- 2. Fake News Detection using Deep Learning
- 3. Breast Cancer Detection using Deep Learning
- 4. Chatbot using Recurrent Neural Network
- 5. Drowsy Driver Detection using Deep Learning
- 6. A Review of Liver Patient Analysis Methods using Deep Learning
- 7. Deep Learning based Thyroid Disease Classification.
- 8. Music Genre Classification System
- 9. Dog Breed Identification using Deep Learning
- 10. Human Face Detection using Deep Learning
- 11. Automated Attendance monitoring using Deep Learning
- 12.Skin Cancer Detection using Deep Learning.

LAB VIVA QUESTIONS

- 1. What is Deep Learning, and how does it differ from traditional machine learning?
- 2. Explain the concept of neural networks and their role in Deep Learning.
- 3. What are the main components of a typical neural network?
- 4. Describe the backpropagation algorithm and its importance in training neural networks.
- 5. How do you choose the appropriate activation function for a neural network?
- 6. Discuss the vanishing gradient problem and its impact on Deep Learning.
- 7. What are some common regularization techniques used in Deep Learning, and how do they prevent overfitting?
- 8. Explain the concept of convolutional neural networks (CNNs) and their applications.
- 9. How does pooling (e.g., max pooling) work in CNNs, and what is its purpose?
- 10. What is data augmentation, and why is it used in CNNs?
- 11. Discuss the challenges and solutions when working with small datasets in Deep Learning.
- 12. Describe the architecture and advantages of recurrent neural networks (RNNs).
- 13. Explain the concept of Long Short-Term Memory (LSTM) cells in RNNs.
- 14. How do you handle the vanishing gradient problem in RNNs?
- 15. What is attention mechanism, and how does it improve the performance of sequence-to-sequence models?
- 16. Discuss the concept of transfer learning and its applications in Deep Learning.
- 17. How can you fine-tune a pre-trained neural network for a specific task?
- 18. Explain the differences between supervised, unsupervised, and reinforcement learning in the context of Deep Learning.
- 19. What are generative adversarial networks (GANs), and how do they work?
- 20. Describe the main components of a GAN architecture (generator and discriminator).
- 21. How can GANs be used for image synthesis and style transfer?
- 22. Discuss the challenges and potential solutions for training GANs.
- 23. What is the concept of autoencoders, and how are they used in unsupervised learning?
- 24. Explain the process of dimensionality reduction using autoencoders.
- 25. How can you use autoencoders for denoising or anomaly detection?
- 26. Discuss the concept of reinforcement learning and its use in training agents to perform tasks.
- 27. Explain the role of the reward function in reinforcement learning algorithms.
- 28. What is Q-learning, and how does it work in reinforcement learning?

- 29. How can you handle the exploration-exploitation trade-off in reinforcement learning?
- 30. Describe the challenges and approaches to dealing with high-dimensional action spaces in reinforcement learning.
- 31. Explain the concept of policy gradients and their advantages in certain reinforcement learning scenarios.
- 32. Discuss the concept of natural language processing (NLP) and its relation to Deep Learning.
- 33. How are recurrent neural networks used in natural language processing tasks like language modeling?
- 34. What is word embedding, and how does it improve the representation of words in NLP models?
- 35. Explain the architecture and applications of transformer models in NLP.
- 36. How does the attention mechanism work in transformer-based models like BERT?
- 37. Discuss the challenges of training large-scale language models and potential solutions.
- 38. Explain the concept of word2vec and its applications in NLP.
- 39. What are the differences between CBOW (Continuous Bag of Words) and Skip-gram word2vec models?
- 40. Describe the process of training a word2vec model.
- 41. How can word embeddings be visualized and evaluated?
- 42. Discuss the concept of auto-regressive models in natural language processing.
- 43. What is beam search, and how does it improve the output generation in sequence-to-sequence models?
- 44. Explain the concept of self-attention and its use in transformer-based models for NLP.
- 45. How can you apply transfer learning to pre-trained language models like GPT-3?
- 46. Discuss the challenges and solutions when working with noisy or unstructured text data in NLP.
- 47. Explain the concept of style transfer in NLP and its applications.
- 48. How can you use Deep Learning models for sentiment analysis on text data?
- 49. Discuss the potential ethical considerations and biases in Deep Learning models for NLP.
- 50. Describe the process of fine-tuning a pre-trained NLP model for a specific language-related task.