

Course Code	PCS21G01J	Course Name	DEEP LEARNING FOR DATA SCIENCE	Course Category	G	Generic Elective Course	L	T	P	C
							3	0	2	4

Pre-requisite Courses	Nil	Co-requisite Courses	Nil	Progressive Courses	Nil
Course Offering Department	Computer Science	Data Book / Codes/Standards			

Course Learning Rationale (CLR):		The purpose of learning this course is to,		
CLR-1 :	Develop knowledge of Neural Network			
CLR-2 :	Understand and analyze deep learning			
CLR-3 :	Perform Optimization Techniques			
CLR-4 :	Implement Deep Learning models			
CLR-5 :	Get Familiar with Keras library			
CLR-6 :	Implement Deep Q-Learning			

Course Learning Outcomes (CLO):		At the end of this course, learners will be able to:		
CLO-1 :	Build a Perceptron model			
CLO-2 :	Build Neural Network model using BP algorithm			
CLO-3 :	Using Pytorch to build a prediction model			
CLO-4 :	Fine Tuning the Deep Learning models for performance optimization			
CLO-5 :	Build, Compile, Test, and evaluate model in Keras			
CLO-6 :	Work with Generative Adversarial Networks			

Learning		
1	2	3
Level of Thinking (Bloom)	Expected Proficiency (%)	Expected Attainment (%)

Program Learning Outcomes (PLO)														
1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
Fundamental Knowledge	Application of Concepts	Link with Related Disciplines	Procedural Knowledge	Skills in Specialization	Ability to Utilize Knowledge	Skills in Modeling	Analyze, Interpret Data	Investigative Skills	Problem Solving Skills	Communication Skills	Analytical Skills	PSO 1	PSO 2	PSO 3
H	H	H	H	H	H	-								
L	H	H	H	H	H	-								
L	H	H	H	H	H	-								
L	H	H	H	H	H	-								
L	H	H	H	H	H	-								

Duration (hour)		15	15	15	15	15
S-1	SLO-1	Introduction to Neural Network	Fine Tuning NN models ANN Processing Components	Fine Tuning NN Models What is Fine Tuning?	Keras and DL Overview of TFMA	Interactive Applications of DL
	SLO-2	Fundamentals	Learning and Training in ANN	Regularization	Practical Consideration for DL	Machine Vision – CNN
S-2	SLO-1	Biological NN Vs ANN	Cluster analysis in ANN	What is Vector Quantization?	DL parameters	Pooling Layers
	SLO-2	ANN Architecture	NN Building blocks	The Encoder-Decoder Model	Data Loading and Preprocessing	Lenet-5 in Keras
S-3	SLO-1	Computational Models in NN Neurons Interconnection	Perceptron to Deep NN Model and Hyper parameters	The Generalized Lloyd Algorithm Relation between SOM and noisy encoder-decoder	Data Preprocessing with Keras Keras Layers	Alexnet and vggnet in keras Natural Language processing
	SLO-2	Threshold Functions Activation functions & ANN	Classification with NN Deep Learning Frameworks	Voronoi Tessellation LVQ Introduction	Training Models with fit() Monitoring Performance Metrics	Creating word embeddings with word2vec Natural Language Classification with familiar networks
S-4-5	SLO-1	Laboratory 1: Implement a Feed Forward Neural Network with Back propagation training algorithm for realizing XOR problem	Laboratory 4: Build a NN model using PyTorch	Laboratory 7: Implement LVQ Network for Pattern Classification	Laboratory 10: Build a model for Credit Card Fraudulence Detection	Laboratory 13: Build a CNN model for Image Classification
	SLO-2					
S-6	SLO-1	Implementing Neural Networks Building Neural Networks Models	NN Categorization	The LVQ Algorithm	Checkpointing	Generative Adversarial Networks

Duration (hour)	15	15	15	15	15
	SLO-2	Use case of ANN	NN Computational Model	The LVQ2 Algorithm	Debugging the model with eager execution
	SLO-1	Perceptrons	NN Building Components	Hebbian Learning	Speed Up process with multiple GPUs
S-7	SLO-2	Single Layer Perceptron Model	Evolutionary Algorithm & Gradient Descent	Hebbian Learning Rule	Multiple GPU and distributed trainings
	SLO-1	Least Mean Square Algorithm	Object Image Classification	Competitive Learning	Transfer Learning
		Learning Curves	Learning rates and Optimization	Optimizing NN	Image classification
S-8		Learning Rates	Optimizing Speed	Debugging NN	Keras Metrics
	SLO-2	Perceptron	Dense Network Tuning using Hyper parameters	Learning rate optimization	Jupyter notebooks
S 9-10	SLO-1	Laboratory 2: Implement a Perceptron in Python	Laboratory 5: Implement ANN Training in Python for MNIST Digit Classification problem	Laboratory 8: Using Keras, perform rate adaption schedule.	Laboratory 11: Work on a text classification problem with Keras API
	SLO-2				Laboratory 14: Design and build a Game environment
S-11	SLO-1	Multilayer Perceptron	Linear Model with Estimators	Optimizing Networks	Dataset for NN
	SLO-2	The XOR Problem	NN for Predictions	Rate adaption schedule	Exploring the Dataset
	SLO-1	Back Propagation Algorithm, Heuristics for improving BP algorithm	Optimization approaches for prediction, NN algorithms	Scaling, Scaling methods	Preparing the dataset, Visualizing the dataset
S-12	SLO-2	Radial Basis Function Networks, Interpolation	Data preparation for NN, ANN Training in python	Batch Normalization, Mini Batch Normalization	Compiling the model, Training the NN
S-13	SLO-1	Regularization	Training Samples	Internal Covariate Shift	Testing the NN
	SLO-2	Learning Strategies	Overfitting and Underfitting	Implement Gradient learning	Evaluate the model
S 14-15	SLO-1	Laboratory 3: Implement a Feed Forward Neural Network with Back propagation training algorithm for realizing Straight line e.g. $y = 2x + 3$	Laboratory 6: Perform Hyper parameter tuning in the ANN model implemented in Lab 5	Laboratory 9: Implement Batch Normalization and gauge its performance	Laboratory 12: Build a DL model for diabetes classification problem
	SLO-2				Laboratory 15: Build and Train the Deep Q Neural Network

Learning Resources	1. Deep Learning with Python, By Francois Chollet, December 2017 2. Deep Learning Illustrated: A Visual, Interactive Guide to Artificial Intelligence, By Jon Krohn, Grant Beyleveld and Aglaé Bassens, September 2019	3. Hugo Larochelle's Video Lectures on Deep Learning 4. Introduction to Deep Learning by Sandro Skansi, Springer, 2018 5. Deep Learning with TensorFlow 2 and Keras - Second Edition, By Antonio Gulli, Amita Kapoor and Sujit Pal, December 2019
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Learning Assessment											
Bloom's Level of Thinking		Continous Learning Assessment(50% Weightage)								Final Examination (50% weightage)	
		CLA – 1 (10%)		CLA – 2 (10%)		CLA – 3 (20%)		CLA – 4# (10%)			
		Theory	Practice	Theory	Practice	Theory	Practice	Theory	Practice	Theory	Practice
Level 1	Remember	20%	20%	15%	15%	15%	15%	15%	15%	15%	15%
	Understand										
Level 2	Apply	20%	20%	20%	20%	20%	20%	20%	20%	20%	20%
	Analyze										
Level 3	Evaluate	10%	10%	15%	15%	15%	15%	15%	15%	15%	15%
	Create										
	Total	100 %		100 %		100 %		100 %		100%	

CLA – 4 can be from any combination of these: Assignments, Seminars, Tech Talks, Mini-Projects, Case-Studies, Self-Study, MOOCs, Certifications, Conf. Paper etc.,

Course Designers		
Experts from Industry	Experts from Higher Technical Institutions	Internal Experts
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