## SEMESTER - IV

Course Code	UDS21401J	Cours Name		L	DEEP LEA	IRNI	VG FOR	ENTERF	PRISE	7		Course Categor		;		P	rofes	sior	nal C	ore	Cou	rse		5	<b>L</b>	<b>T</b>	<b>P</b> 2	<b>C</b> 5		
Pre-re	equisite Courses	Nil				C	o-requisi	ite Courses	s N	Vil		7		6	Pro	gress	sive (	Cour	ses	Nil										
Course Of	fering Departme	ent	Co	omputer A	pplications	3		*		Data Boo	k / Cod	es/Star	dard	ls N	il															
Course Le	arning Rational	e (CLR):	Th	he purpos	e of learnir	ng this	course	is to,				Learnin	g	É			Р	rogra	am Le	earni	ing (	Outco	omes	s (PL	0)					
CLR-1:	To make the p										ер	1	2	3		1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
CLR-2:	To make the pa and ways to in										roache	3,						7												
CLR-3:	To make the p works just as h	articipa numans	nts unde do using	erstand the	ne methods etworks	s of te	eaching n	nachines i	n per	forming	197	1							4		ı									
CLR-4:	To build intellighealthcare, rediverse data se	tail, ene														3														
CLR-5 :	To provide the concepts of ne error function,	urons, w	eight, b	oias etc alo	ong with the								cy (%)	nt (%)		ledge	cepts	Disciplines	dge	ion	Knowledge		Jata		Skills	Skills			ior	
CLR-6:	To bring the le then performs for a given ind	research	n, desigi	n, develop	oment, and	delive	ers an en	nd-to-end o	deep l	earning	solution	0	Proficien	Proficiency		Fundamental Knowledge	on of Cor	Related	al Knowledge	Specialization	Utilize Kn	Modeling	Interpret	Investigative Skills	Solving St	on	Skills		nal Behavior	Learning
ex.				1	1							f	ted	ted	ted		zatio	vith	dural		to	.⊑	ze,		me (	nun	tica	Skills	ssio	ong
Course Le	earning Outcome	es (CLO	): A	At the end	of this cou	ırse, le	earners v	vill be able	to:			evel	Expected	Expected		Fund	Application	Link v	Proce	Skills in	Ability	Skills	Analy	Inves	Problem	Communicati	Analytical	ICT S	Professional	Life
CLO-1 :	Have a strong have a strong					The second second		MARKET THE PARTY OF THE PARTY O	arning	includin	g ability	to 2	100000	80	L	Н	Н	Н	Н	Н	Н	Н	Н	Н	М	М	Н	Н	Н	Н
CLO-2 :	Have a strong techniques for							The second secon				3	85	80		Н	Н	Н	Н	Н	Н	Н	Н	Н	М	М	Н	Н	Н	Н
CLO-3 :	Use all the the for the problen	_		s and know	wledge in a	applyin	ng the rig	ht set of d	eep le	earning t	echniqu	es 3	85	80		Н	Н	Н	Н	Н	Н	Н	Н	Н	М	М	Н	Н	Н	Н
CLO-4 :	Gain hands-or and technique				7.77						ity of to	ols 3	85	80		Н	Н	Н	Н	Н	Н	Н	Н	Н	М	М	Н	Н	Н	Н
CLO-5 :	Get Hands-on network that fa	1000		•				46 14 HOUR HER				1	85	80		Н	Н	Н	Н	Н	Н	Н	Н	Н	М	М	Н	Н	Н	Н
CLO-6:	Able build a fu networks.							-				777	85	80																

Note: All our curriculum, study materials, assignments, quizzes, lab works, and learning resources are personalized and dynamically generated using machine learning models based on the learner's learning ability. Users can review our learning curriculum only through our intelligent learning management platform (iLMSP), and our learning resources and lab infrastructures are available only in the digital form on our cloud infrastructures.

Duration (hour)		18	18	18	18	18		
C 1	SLO-1	Unit 1: Deep Learning - Deep Dive	Backpropagation	Cost Function	Improving Activation  Maximization with an expert	Eigen Vectors, Eigen Value, Single Value Decomposition		
5-1	S-1 SLO-2 Optimization for Deep Learning		Gradient Descent	Gradient Descent	Performing Activation  Maximization in a code space	Unit 13: Generative Adversarial Networks (GAN)		
SLO-1		Aggregated Residual Transformations for Deep Neural Networks	Unit 4: Deep Learning in Real World Applications	Stochastic Gradient Descent	Explaining DNN Decisions	Generative Adversarial Models Overview, Discriminative vs Generative Modelling, Examples of Generative models		
	SLO-2	Spatial Transformer Networks	Deep learning in Healthcare	Learning Rate	Backward Propagation Techniques	Generative Adversarial Networks Overview, The Generator Model, GAN's and CNN's		
6.3	SLO-1	End-to-end Optimized Image Compression  Deep learning in Retail		Batches, Epochs and Iteration	Unit 9: Deep Neural Net optimization, Tuning	Conditional GAN's, Why Generative Adversarial Networks, Generative Adversarial Networks Training, Loss Functions		
S-3	SLO-2	Generative Adversarial Nets	Deep learning in Energy	Unit 7: Deep Neural Networks and Tools	Optimizers overview, Gradient Descent, Stochastic Gradient Descent (SGD), Mini Batch Stochastic Gradient Descent (MB-SGD), SGD with momentum	Unit 14: Backpropagation, Regularization and Batch Normalization		
S-4	SLO-1	Improved Techniques for Training GANs	Deep learning in Oil & Gas	Deep Neural Network overview	Nesterov Accelerated Gradient (NAG), Adaptive Gradient (AdaGrad)	Back Propagation Overview, Working of Back Propagation algorithm, Need of Back Propagation algorithm, Types of Back Propagation algorithms		
	SLO-2	Unit 2: Deep Learning Approaches	Deep learning in Automobile	Difference between neural network and deep neural network	Tuning the layers, Hyperparameter Tuning	Feed Forward Networks Overview, Batch Normalization Overview, Working of Batch Normalization		
&	SLO-1	Lah 1 ·	Lab 4 :	Lab 7:	Lab 10:	Lab 13:		
S-6 S-7	SLO-1	Unit 5: Challenges of Deen		Deep Learning Neural Network overview	learning rate, Momentum β, for RMSprop, etc, Mini-batch size, Number of hidden layers, learning rate decay, Regularization λ	Need of Batch Normalization algorithm, Regularization overview, How does Regularization reduce overfitting		

	SLO-2	Supervised Learning	Data Issues	Deep Convolutional Neural Network overview	Unit 10: Convolutional Neural Network	Types of Regularization techniques  ✓ L2 and L1 regularization ✓ Dropout ✓ Data augmentation ✓ Early stopping		
	SLO-1 Unsupervised Learning		Overfitting in neural networks	Improving accuracy of the neural networks	Convolution , ReLU layer, Pooling, Padding, Flattening	Unit 15: Backpropagation, Regularization and Batch Normalization		
S-8	SLO-2	How to select a Deep Learning Algorithm	Hyperparameter optimization	The problem of explainability	Full Conversion Layer, Softmax, Cross-Entropy	Overview of Gradient descent optimization, Overview of Adagrad Gradient descent algorithm		
S-9	SLO-1 Deep Learning Workflow and applications		High Performance Hardware		Unit 11: Recurrent Neural Network	Gradient descent with Adagrad  ✓ Two-Dimensional Test Problem  ✓ Gradient Descent Optimization With AdaGrad ✓ Visualization of AdaGrad		
	SLO-2	Challenges and Vision for the future	Neural network is a Black Box	Learned features	RNN intuition, Vanishing Gradient Problem, Tackling Vanishing Gradient Problem	Overview of Adadelta Gradient descent algorithm, Gradient descent with Adadelta		
S-	SLO-1	Analysis of Deep Learning applications	Lack of Flexibility	Feature visualization	Exploding Gradient Problem, Tackling Exploding Gradient Problem	Overview of RMSProp Gradient descent algorithm, Gradient descent with RMSProp		
10	SLO-2	Unit 3: Deep Learning Techniques	Multitasking	Feature Visualization through Optimization	Long Short-Term Memory, Applications of Recurrent Neural Networks	Overview of Adam Gradient descent algorithm, Gradient descent with Adam		
S- 11 & S- 12	SLO-2	Lab 2 :	Lab 2 :		Lab 11:	Lab 14:		
S- 13	SLO-1	Classic Neural Networks	Classic Neural Networks Deep Learning Security		Unit 12: Auto Encoders and dimensionality reduction in networks	Unit 16: Deep Learning Hands On Lab Work 2- Build, Test and Deploy ML Models (Health - 3)		
	SLO-2	Convolutional Neural Networks	onvolutional Neural Networks  Unit 6: Artificial Neural Networks		Autoencoders overview	Patient Segmentation		
S- 14	SLO-1	Recurrent Neural Networks (RNNs)		letworks		Problem statement, Problem type		

					✓ Under complete Autoencoder ✓ Variational Autoencoder ✓ LSTM Autoencoders ✓ Hyperparameters of	
	SLO-2	Generative Adversarial Networks	Weight	Network Dissection Algorithm	Autoencoders  Applications of Autoencoders  Dimensionality reduction Anomaly detection Image denoising Image compression Image generation	Data engineering, Data pipeline
S-	SLO-1	Self-Organizing Maps	Bias	Experiments	Dimensionality Reduction with PCA	Model selection, Model engineering
15	SLO-2	Boltzmann Machines	Activation Function	Advantages of Feature visualization	The Curse of Dimensionality	Mode outcome, analysis, and optimization
S-	SLO-1	Deep Reinforcement Learning	Forward Propagation	Disadvantages of Feature visualization	Principal component analysis	Model pipeline
16	SLO-2	Autoencoders	Backward Propagation	Activation Maximization	Eigen Value Decomposition	Data visualization, User interface
S- 17 & S- 18	SLO-1 SLO-2	Lab 3:	Lab 6:	Lab 9:	Lab 12:	Lab 15:

Learning	1.	Machine Learning at Enterprise Scale by Piero Cinquegrana, Matheen Raza Released	2.	Deep Learning for Business Managers Artificial Intelligence Prithwis
Resources		July 2019, Publisher(s): O'Reilly Media, Inc.	100	Mukerjee

Learning	Assessment	ja.			11/6				-			
	D		Continuous Learning Assessment (50% weightage)									
	Bloom's	CLA -	1 (10%)	CLA -	2 (10%)	CLA-	3 (20%)	CLA -	4 (10%) #	(50% weightage)		
	Level of Thinking	Theory	Practice	Theory	Practice	Theory	Practice	Theory	Practice	Theory	Practice	
Level 1	Remember	200/	150/	200/	150/	200/	150/	200/	150/	200/	150/	
	Understand	20%	15%	20%	15%	20%	15%	20%	15%	20%	15%	
Level 2	Apply	200/	200/	20%	20%	20%	200/	20%	20%	200/	20%	
Level 2	Analyze	20%	20%	20%			20%	20%	20%	20%	20%	
Level 3	Evaluate	10%	15%	10%	15%	10%	15%	10%	15%	10%	15%	
	Create	10%	13%	1076	15%	10 %	15%	1076	15%	10%	13%	
Total		10	0 %	10	0 %	10	0 %	10	0 %	10	0 %	

<sup>#</sup> 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		
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