Course Code	PAD21202J	Course Name	BUILD	ING MACHINE	LEARNIN	G PIPELINE	•	ours tego		С	7		Pro	fess	iona	ıl Co	re (	Cours	ses			L 3	T 0	P 4	C 5
Pre-requisite Courses Nil Co-requisite Courses Nil														Prog	ress	ive C	ours	es	Nil						
Course Offe	Course Offering Department Data Science Data Bo							ok / C	Code	s/Sta	ndar	ds	Nil												
Course Learning Rationale (CLR):  The purpose of learning this course is to:						Le	arn	ing				Pro	ogra	m L	earı	ning	Out	com	es (l	PLC	<b>)</b> )		~		
<b>CLR-1</b> :	Simpler proce	esses to u	pdate exis	ting models, Less	time spent	to reproduce n	nodels	1	2	3		1	2	3	4	5 6	5 7	8	9	10	11	12	13	14	15
CLR-2:	Help detect p	otential b	iases in th	e datasets or in th	e trained mo	odels		Le	Fy	Ex			I	i		Α	b				32				
<b>CLR-3</b> :	Free up devel	lopment t	ime for da	ta scientists and i	ncrease their	r job satisfacti	ion.		pe	pe			Ap	- II	r S	k ili	3.86	An		Pr	Co		I	Pr ,	
CLR-4:	Automated machine learning pipelines will free up from maintaining existing models.				of	cte	cte		nd am	pli v	h c	c i	ls y	Sk	aly s ze,	In	ob	m m	An	(	of	Lif e			
<b>CLR-5</b> :	Publishing th	e Trained	l Model as	a Web Service for	or Inference			Th		At		ent		Re u	ır S	p U	t in	Int	sti	m	un	aly tic	r i	0	Lo
CLR-6:	Validating a	Recomme	endation S	ystem				ın ki	ofi	tai		al	- C	<sub>A</sub> a	5.5		i M	er	gat	So	ıca		N .	iai .	ng Le
Course Le	earning Outco	omes	At the en	d of this course, l	earners will	be able to:		ng (B lo o m)	y	me nt (%		Kn ow led ge	nc s	ci oli			n eli w ng d	pr et Da ta	Sk ills	lvi ng Sk ills	n Sk	1115	llis l	vi '	ar ni ng
CLO-1:	Recognise a omanagement		-	an and understand g concept,	d the purpor	se of a data		3	80	70		L	Н		Н	L -	-	-	L	L	-	Н	-	-	-
CLO-2:	Understand E Vector Norm		Data Scie	nce, Linear Alge	bra, Vector	Scalar Multipl	lication -	3	85	75		M	Н	L	M	L -	-	-	М	L	-	Н	-	-	-
CLO-3:	Understand matrix operations				3	75	70		M	H	M .	Н	L -		25	M	L		Н	12	-	2			
CLO-4:	0-4: Determinants – Orthogonal matrices Gaussian distribution – Binomial distribution				3	85	80		M	H	M	H .	L -		-	M	L	-	Н	-	-	-			
<b>CLO-5</b> :	O-5: Loading ,Scaling and encoding the data				3	85	75		H	H	M .	H .	L -	. 2	2	M	L	200	H			2			
CLO-6: Find principal component using Principal Component Analysis and Normalizing a dataset				3	80	70		L	Н	-	Н	L -	-	-	L	L	-	Н	-	-	-				
Duration (hour)		21		21			21						21							2	1				

		Claud took poloning and Data						
S-1	SLO-1	Cloud technologies and Data governance, designing a data governance process	Linear Algebra Basics	Matrix arithmetic , working with matrix, From Scalars and Vectors	Work with Vectors	Loading and exploring a dataset		
	SLO-2	managing a Data governance strategy, monitoring a data governance strategy	Linear algebra for Machine learning	Shapes and indexing	Basis and projection of vectors	the binarizer(), the minmaxscaler()		
S-2	. 71   1   -	maintaining a Data governance strategy	Linear Transformations, Intuition	Matrix operations- Addition and Scalar Multiplication	work with - Matrix multiplication	the standard scalar		
0-2		Data access governance		Transposition, Matrix Decomposition	Inverse matrix	the normalizer, the maxabsscaler()		
	SLO-1	risk and Data safety compliance	Classes of space- scalar	Matrix and PCA-covariant matrix		label encoding, One-Hot encoding		
S-3	5LU-2	governance and its relationship with big data	Vector and its types	Eigen value,	linear Transformations	Loading and analyzing a dataset		
S 04	SLO-1	I AD Data collection	I AP : Vector addition	LAB : Matrix	I AD : Creating a			
- S 07	SLO-2	LAB :Data collection	LAB : Vector addition , vector multiplication	Transformation in	LAB : Creating a Recommendation Engine	LAB :Recommending Items Based on Other Items		
S-8	SLU-I	why big data requires governance	Vector space , Subspaces	Eigen vector calculation	Gaussian elimination	building and evaluating a Linear		
5-0	SLO-2	why is Big Data different?	types of Vector space	sparse matrix	Gaussian elimination-Example problem	Regression mode		
S-9	OLU-I	Cloud technologies and Data governance	Operation on vectors-Addition	Tensor Arithmetic	Determinants	scaling and encoding the data		
3-9	SLU-Z	designing a Data governance process	Subtraction	Hadamard product and Tensors-	Determinants	Analyzing the effects of pre- processing		
S-10	SLU-1	maintaining a data governance strategy	multiplications	Singular-Value Decomposition	Orthogonal matrices	Standardizing continuous data		
3-10	910-2	Data access governance, Data access patterns	multiplications	Probability basics and propositions	Orthogonal matrices	Loading a dataset,		
S 11	SLO-1			2 0/23 00 01 (200000)				
s 14	SLO-2	LAB : Manipulating data		LAB :Using KNN describing Similarity neighborhoods	LAB :Recommending Another Item	LAB :Evaluating a Recommendation System		
S-15	SLU-1	data breach prevention – least privilege	Scalar and vector multiplications	random variable	Eigenvectors and Linear Transformations	scaling a dataset		
3-13	SLO-2	system permissions		Central limit theorem	Change of Basis	spotting correlations in a dataset		
S-16	. 31 \ /-	create an AWS user and group, vulnerability assessments	Linear product Vector	parameter estimation	Linear Transformations in Different Bases	Principal Component Analysis		

	OLU-Z	Data classification,, data encryption	Theorems related to linear products	Gaussian distribution	Eigen decomposition	
S17	SLO-1		Vector Norms- Definitions, Examples of Norms	Binomial distribution	Pseudo inverse	Normalizing a dataset
		solutions, logging	Norm Representations	Dinomial distribution		Normalizing a dataset
S 18 - S 21	2102	Lab: matrix addition, Matrix subtraction, Matrix multiplication	3	LAB :Tensor Hadamard product	EXE II III diling itcilio to	LAB : Validating a Recommendation System

	<ol> <li>Data Governance: The Definitive Guide, By Evren Eryurek, Uri Gilad,</li> </ol>
	Valliappa Lakshmanan, Anita Kibunguchy and Jessi Ashdown, March
Resources	2020
	<ol><li>Essential Math for Data Science, By Hadrien Jean, November 2020</li></ol>

- 1. Feature Engineering for Machine Learning, By Alice Zheng and Amanda Casari, 2018.
- 2. Python Feature Engineering Cookbook, By Soledad Galli, January 2020

Learning A	earning Assessment											
	Bloom's		Final Examination									
	Level of	CLA –	1 (10%)	CLA – 2 (10%)		CLA –	3 (20%)	CLA –	4 (10%)#	(50% weightage)		
	Thinking	Theory	Practice	Theory	Practice	Theory	Practice	Theory	Practice	Theory	Practice	
	Remember	20%	2007	15%	15%	15%	150/	15%	15%	15%	150/	
Level 1	Understand		20%				15%				15%	
I1 0	Apply	2007	200/	200/	200/	200/	200/	200/	200/	200/	200/	
Level 2	Analyze	20%	20%	20%	20%	20%	20%	20%	20%	20%	20%	
Level 3	Evaluate	10%	100/	150/	150/	150/	150/	150/	150/	150/	150/	
Level 3	Create		10%	15%	15%	15%	15%	15%	15%	15%	15%	
	Total	100 % 10		00 % 100 %			10	0 %	×2			

Course Designers								
Experts from Industry	Experts from Higher Technical Institutions	Internal Experts						
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Mr.M. Hemachandar, Tech Lead, Wipro Limited, Chennai	Dr. Vincent, Associate Professor, VIT							