Cour		AD21D01T	Course Name		IIA IA ENGINEERING AND GOVERNANCE			ours tego	12.40.40	D			Disc	ipline	e Sp	ecifi	E EI	ectiv	e			L .	Г Р) 0	-
Pr	Pre-requisite Courses Nil Co-requisite Courses Nil							Progi	ress	ive Cou	rses	N	il											
Course	ourse Offering Department Computer Applications Data Book / Codes/Standards					Nil																		
Course	Course Learning Rationale (CLR): The purpose of learning this course is to:					Le	earni	ng				Pr	ogra	m Le	arni	ng O	utco	omes	s (PLC))				
CLR-1	: To I	earn the cond	cepts of	Big data				1	2	3] [1	2	3	4	5	6	7	8	9	10	11 12	2 13	14	15
CLR-2	: To i	mpart in-dept	th knowle	edge of data la	akes								Lin			Ab								
CLR-3	: Una	lerstand the p	rinciples	of Data ware	ehouse			Le	I E X	Ex	Fu	ı Ar	k wit	Pr	Ski	ilit	1	An .		Pr (Co		Pr	
CLR-4	: Bas	ic knowledge	of lake	on AWS				vel of	nec	pe cte		pli	h	OC	le l		ki C	aly II	nv (obl	n Ar	000	ofe	Life
CLR-5	: Bas	ic knowledge	of distri	buted compu	ting using Spark			Thi	tea	ч	er al	tio	t Re	ur	n	l 1#i [11		ze, e nt li		em	nu al	y IC	on	Lon
CLR-6	: Des	ign principles	of Resi	ient distribute	ed datasets					Att ain	al Kı	U		6.41			lo S	nt iq er t		So ir Ivi ia			al Be	50
Course	urse Learning Outcomes (CLO): At the end of this course, learners will be able to:		(BI oo m)	y (%)	me nt (%)	le	v no d ep	I DII	led	OII	ow led ge	1	et S Da II ta		Ski S	Ski IIs	ki IIs	ha vio r	ning					
CLO-1	: Hav	e a thorough	Underst	anding of Big	data			3	80	70	ŀ	H	М	-	-	-	-	-	Н	Н		N	l H	Н
CLO-2	35 (3	erstand the o	concepts	of Data ware	house			1	_	_	1		~	Н	Н	-	М		-	Н	- -	N		Н
CLO-3	: Rea	l time applica	ations of l	Data lake des	sign principles			3	75	70	ŀ	l H	М	Н	Н	-	М	-	Н	Н		N	8050	Н
CLO-4	: Dep	loyment kno	wledge o	of AWS				3	85	80	ŀ	H	Н	-	-	-	-	-	Н	М		N	Н	Н
CLO-5	CLO-5 : Design and implementation knowledge of Spark data frames				3	85	75	H	l N	И	М	М	М	М	-	Н	Н	- N	1 N	Н	Н			
CLO-6	CLO-6 : Real time application of accumulator param			3	80	70	ŀ	Н	М	-	-	-	-	-	Н	Н		N	I H	Н				
	Duration (hour) 12 12			12			12 12																	
S-1	S-1 SLO-1 DataArchitecture: Data and Data lifecycle databases and it's types BigData :Introduction to big data			DataLakeArchitecture	e: Da	ıta sil	IOC	Data LakeonAWS: Data lakes and Data warehouses Distributed Computing usingSpark: PySpark are basics introduction to specific to the second se																

7		Henrity relationship diadram i	building systems to scale with data	Data lakes	Data lake selection criteria	Resilient distributed datasets(RDD), Spark data frames		
S-2		implementing NoSQL with AWS, create, NoSQL DB with python	building systems to scale with	Data lakes	Data lake and data	Spark architecture		
3-2	ISLU-Z	create SQL DB with python, big data	data	characteristics of Data lakes	Democratization	working with RDD		
S-3	SLO-1	reading data from csv	a quick overview of Hadoop	Data lake architecture	Data lake design principles	creating data frames for RDD – SQL context		
	SLU-2	importance of volume	a quion ovoiviou oi i ladoop	Data lake are intectare	AWS Data lake architecture	map() function of RDD		
	SLO-1	the importance of variety, the importance of velocity		Data warehouse	Implement AWS data store	access content of data frame		
S-4	SLO-2	the importance of veracity	Map-Reduce overview	Data Lakes	Data lake for on-premise and multi-cloud	data frame in spark and Pandas, performance improvements in Spark		
S-5	SLO-1	the relationship between the four VS	Map-Reduce overview	Data streams	data processing frameworks for data lake	broadcast variables and accumulators – loading data into a data frame		
88.300 9	SLO-2	Variety and Data structure			real-time big data architectures	Sampling the contents of a data frame		
S-6		Validity and Volatility	man nhaco	Data streams	Data lake reference architecture	grouping and aggregations – visualizing data in a data frame		
3-0	SLO-2	finding balance in the four VS use cases	map phase	Data streams	data ingestion and file formats	trimming and cleaning data		
S-7	SLO-1	LOVING VALUE FROM THE TOUR V.	map phase of Map-Reduce	migrate data to AWS	ingestion using Sqoop	user-defined functions and Data frames – combining filters –		
9	SLO-2	Data driven organizations	map phase of wap-reduce	inigrate data to Avvo	Data processing strategies	aggregations – and sorting – using broadcastvariables		
	SLO-1	decision making			deriving value from data lakes	using accumulators – exporting		
S-8	SLO-2	distributed systems	Shuffle Phase	migrate data to AWS	data life cycle – and glacier	data frame Contents – Custom accumulators		
	SLO-1	batch vs in- memory processing	Shuffle Phase		create role for AWS glue service	Join operations – the Spark catalyst optimizer		
S-9	SLO-2	tools for data management		data lakes on AWS	upload data to explore the glue web console	Introduction To Spark SQL – Preparing Data For Analysis – Running SQL Queries		

S-10	Control of the Contro	understanding ETL	roducophaco	data lakes on AWS	manually create glue table	Inferred And Explicit Schemas – Windowing In Spark	
3-10	new management	ETL with Talend open studio	reducephase	uata lakes on Avvo	query data lake using amazon Athena	applying – Window Functions – PySpark Basics – sparkconf	
	131 V J= 1	ETL pipeline in Python, Al and machine learning		Working with data lakes	configure and run glue crawlers	Spark context – Spark files – RDD – Storage level – Broadcast	
S-11		data modelling	reducephase		access data in crawled tables	accumulator – accumulatorparam – marshalserializer – Pickle serializer – Status tracker	
	SLO-1	data partitioning/onginooring/			Crawl CSV files, merge data	sparkjobinfo – sparkstageinfo – profiler – basic profiler	
S-12	SLO-2	reporting	Difference between Shuffle and Reduce Phase	Data Lakes on AWS	Same schema file manipulation	evaluatorpyspark.ml.tuning Module – PySpark SQL functions – PySpark SQL data Types	

Ÿ .	Text Book:	Reference Book:			
Looming	1. Data Architecture: A Primer for the Data Scientist, 2nd Edition, By	 Concise Guide to Databases by Peter Lake and Paul Crowther, Springer, 2013 			
Learning	H.Inmon, Daniel Linstedt and Mary Levins, April 2019 2. The Enterprise Big, Data Lake, By Alex Gorelik, March 2019				
Resources	2. Data Architecture, By Charles Tupper, May 2011	3. Apache Spark with Python - Big Data with PySpark and Spark, By Pedro Magalhães			
		Bernardo, Tao W and James Lee, May 2018			

Learning A	Learning Assessment														
	B		Final Examination												
Level	Bloom's Level of Thinking	CLA - 1 (10%)		CLA - 2 (10%)		CLA –	3 (20%)	CLA - 4 (10%)#		(50% weightage)					
	Level of Tilliking	Theory	Practice	Theory	Practice	Theory	Practice	Theory	Practice	Theory	Practice				
Lovel 1	Remember	400/		400/		400/		400/		400/					
Level 1	Understand	40%		40%	-	40%	5	40%	ā	40%	ā				
Level 2	Apply	30%	200/	200/	200/	200/		30%		30%		30%		30%	
Level 2	Analyze		-	30%	-	30%	•	30%	-	30%	-				
evel 3	Evaluate	30%	524	200/		30%	152	30%	8	30%	88				
	Create	30%		30%	·	30%	7	30%		30%	5				
Total		100) %	100) %	100) %	100	%	100	0 %				

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