	urse ode	UDS21D06T	Course Name	SOCIAL MEDIA	AND TEXT ANALY	TICS	Cou	rse (Cate	gory	'	D		Di	iscip	oline	Spe	ecifi	ic El	ecti	ve		L 4	T 0	P 0	C 4
F	re-requ	uisite Courses	Nil		Co-requisite Co	ourses	Nil	Progressive Courses Nil																		
Cours	e Offer	ing Department	Com	puter Applications			Data Bo	ok / (Code	es/St	andar	ds	Nil													
Course Learning Rationale (CLR): The purpose of learning this course is to:						Learning Program Learning Outcomes (PLO)																				
CLR-1: To leverage the power of the R eco-system to extract, process, analyze, visualize and model social media data							1	2	3		1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	
CLR-2: Visualize and analyze data from social media platforms to understand and model complex relationships using various concepts and techniques CLR-3: Understand the fundamentals of text mining							(mc	(%)	(%)		ge	ts	iplines			edge		ta						9)		
CLR-	CLR-4: Utilize text for prediction techniques CLR-5: Understand the relevance between information retrieval and text mining CLR-6: Analyze different case studies related to text mining							Thinking (Bloom)	roficiency	Proficiency Attainment			of Concepts	elated Disciplines	Knowledge	Specialization	Itilize Knowledge	Modeling	aInterpret Data	e Skills	olving Skills	ation Skills	Skills		I Behavior	earning
Course Learning Outcomes At the end of this course, learners will be able to:						Level of Thi	Expected	Expected /		Fundament		Link with Re	Procedural	Skills in Spe	Ability to Ut	Skills in Mo	Analyze, alr	Investigative	Problem So	Communica	Analytical S	ICT Skills	ession	Life Long Le		
CLO-				dia analytics and R		77. 7	40.7	3	80			100	Н	-	Н	L	-	-	-	L	L	-	Н	-	-	-
CLO-				dia channels such a	as Twitter & Flickr			3	85			М	Н	L	М	L	-	-	-	M	L	-	Н	7	-	7
CLO-		quire knowle <mark>dge o</mark>					1	3	75	_	-	М		М	Н	L	-	-	-	M	L	=	Н	-		1.7
CLO-		rform prediction fr		aluate it				3	85			1000000	2.2	M	Н	L	-	-	-	M	L	-	Н	-	-	-
CLO-		rform document n						3	85		+ ⊢	Н	Н	М	Н	L	-	-	-	M	L	-	Н	-	-	-
CLO-	6 : <i>Un</i>	derstand how text	mining is impl	emented		- 41		3	80	70		L	Н	-	Н	L	-	-	-	L	L	-	Н	-	-	-
155	Duration (hour) 12 12 12				12	12 12																				
S-1	SLO-1 Getting Started with R and Social Visualizing data Overview of Text M				Mining Using T			ing Text for Prediction					Finding Structure in a Document													
77500000	SLO-2	Understanding So	ocial Media	Managing packages What's Si Mining?		Special Ab	The second secon			Recognizing that Documents Fit a Pattern				it	t											
120000000000000000000000000000000000000	SLO-1 Advantages and Significance of Social Media Data analytics - Analytics w		nalytics workflow	Structured or Unstructured Data			How Many Documents Are Enough?					Clustering Documents by Similarity														
S-2	SLO-2	Disadvantages ar Social Media	Machine learning techniques		t Different from ers?			Document Classification					Similarity of Composite Documents													
		learning Be So			What Types of Problem Be Solved?			blems Can Learning			ing to Predict from Text				_ l	Mor	ane	Clus	torin	10						
SLO-2 A typical social media anal workflow		edia analytics	Text analytics Docume		ment Classification			Similarity and Nearest-Neighbor Methods				or N	k-Means Clustering													

S-4	0.454 (0.414 (4.424 (4.4	Data access, Data processing and normalization, Data analysis and Insights	Understanding Twitter, APIs	Information Retrieval	Document Similarity	Hierarchical Clustering		
	SLO-2	Opportunities and Challenges	Registering an application	Clustering and Organizing Documents	Decision Rules			
S-5	SLO-1	Getting started with R	Connecting to Twitter using R	Information Extraction	Decision Trees	The EM Algerithm		
3-3	SLO-2	Environment setup	Extracting sample Tweets	Prediction and Evaluation	Scoring by Probabilities	The EM Algorithm		
	SLO-1	Data types		From Textual Information to Numerical Vectors	Linear Scoring Methods	What Do a Cluster's Labels Mean?		
S-6	SLO-2	Data structures-Vectors	Trend analysis	Collecting Documents	Evaluation of Performance - Estimating Current and Future Performance	Applications, Evaluation of Performance		
C 7	SLO-1	Arrays	Sentiment analysis	Document Standardization	Getting the Most from a Learning Method	Case Study: Market Intelligence		
S-7	SLO-2	Matrices	Key concepts of sentiment analysis –Subjectivity, Sentiment polarity	Tokenization	Errors and Pitfalls in Big Data Evaluation	from the Web		
	SLO-1	Lists	Opinion summarization	Lemmatization-Inflectional Stemming	Information Retrieval and Text Mining	Case Study: Lightweight Document Matching for Digital		
S-8	SLO-2	Data Frames	Features	Stemming to a Root	Is Information Retrieval a Form of Text Mining?	Libraries		
S-9	SLO-1	Functions - Built-in functions	Sentiment analysis in R	Vector Generation for Prediction	Key Word Search	Minin <mark>g Social M</mark> edia		
	SLO-2	User-defined functions		Multiword Features	Nearest-Neighbor Methods			
S-10		Controlling code flow - Looping constructs	Follower graph analysis	Labels for the Right Answers, Feature Selection by Attribute Ranking	Measuring Similarity -Shared Word Count	E-mail Filtering		
	SLO-2	Conditional constructs	Flickr Data Analysis	Sentence Boundary Determination	Word Count and Bonus, Cosine Similarity			
S-11	SLO-1	Advanced operations	Accessing Flickr's data	Part-of-Speech Tagging	Web-Based Document Search - Link Analysis	Emerging Directions		
Parks Rate	SLO-2	apply, lapply	Understanding Flickr data	Word Sense Disambiguation	Document Matching	Summarization		
<i>*</i>	SLO-1	sapply,tapply	Understanding interestingness – similarities	Phrase Recognition, Named Entity Recognition, Parsing	Inverted Lists	Active Learning		
S-12	1	mapply	Are your photos interesting? - Preparing the data -Building the classifier	Learning with Unlabeled Data				

Sholom M. Weiss, Nitin Indurkhya, Tong Zhang, (2015), "Fundamentals of Predictive Text Mining", Second Edition, Springer London.

Learning Ass	sessment											
	Disamia	Continuous Learning Assessment (50% weightage)								Final Examination		
	Bloom's Level of	CLA - 1 (10%)		CLA - 2 (10%)		CLA -	3 (20%)	CLA – 4	4 (10%)#	(50% weightage)		
	Thinking	Theory	Practice	Theory	Practice	Theory	Practice	Theory	Practice	Theory	Practice	
Level 1	Remember Understand	40%	-	40%	CAL	40%	77 -	40%	-	40%	-	
Level 2	Apply Analyze	40%	-	40%	7.3	40%	1111	40%	-	40%	H	
Level 3	Evaluate Create	20%	-4-	20%		20%	-	20%	-	20%	•	
	Total	100	0 %	10	0 %	10	0 %	10	0 %	100	0 %	

Course Designers							
Experts from Industry	Experts from Higher Technical Institutions	Internal Experts					
Mr.Jothi, Periyasamy, Chi <mark>ef Al Arch</mark> itect DeepSphere Al, CA, USA	Dr. S. Gopinathan, Associate Professor, University of Madras, Chennai	Mrs.S.Chandrakala, SRMIST					
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