

Scheme – 2020

Department of Computer Science & Engineering

G. Pulla Reddy Engineering College (Autonomous): Kurnool

Accredited by NBA of AICTE and NAAC of UGC

Affiliated to JNTUA, Anantapuramu

Scheme and Syllabus for

Honors in

COMPUTER SCIENCE & TECHNOLOGY

(With Effect from the Batch Admitted in 2020-21)

Honors in Computer Science & Technology (CST) Scheme of instruction and examination

(Effective from 2020-2021)

S	Semester	Course Title	Credits	instru (per	me of uction iods/ ek)	Schem	e of Examination		
No.				L	Р	End Exam	Internal Assessment	Total 100	
1	IV	Foundations of AI & Machine Learning	4	4	0	60	40	100	
2	V	Python for Data Science	4	3	2	60	40	100	
3	VI	Social Data Analytics	4	3	2	60	40	100	
4	VII	Natural Language Processing	4	3	2	60	40	100	
5		MOOC – 1	2	0	0			100	
6		MOOC – 2 / Mini Project	2	0	0			100	
		Total	20						

	MOOCS - 1
1	Scalable Data Science
2	Information Security
3	Cyber Security
4	Internet of things
	MOOCS – 2 / Mini Project
1	Computer Vision
2	Virtual Reality
3	Database security
4	Fuzzy Logic & Neural Networks

FOUNDATIONS OF AI & MACHINE LEARNING (FAIM)

Course	er : Honors						Sc	heme : 2020
Code	Category	Ho	urs/W	eek	Credits	Maxi	mum Mar	ks
HCS01	Н	L	Т	Р	С	Continuous Internal Assessment	End Exam	TOTAL
		4	0	-	4	40	60	100
Sessional F	Exam Duration	n:1½	2 Hrs			End I	Exam Dur	ation: 3 Hrs
Course Ou	tcomes :At the	e end o	f the co	ourse th	ne student v	will be able to		
	erstand the bas		<u> </u>		<u> </u>			
	oret the importa					<u> </u>		
					•	ibution in ML an		
	· .				0 0	ms and their appl	ications.	
CO5: Desc	ribe the found	ations c	ot Al a	nd then	r applicatio	ns.		
				U	I – TIN			
						, Solving System near Mappings, A		
	ees, Emear me	iepenae	filee, D		NT – II	itear mappings, r	time space	
<i>Probability</i> Probabilitie	es, Sum Rule	tions:C	Constru duct I	UN action		bility Space, Di 5 Theorem, Su		
Independen	ice, Gaussians	Distrib	ution.	UN	IT – IV			
Application Regression,		Associ	iations	hat is N , Class	Machine Le	arning? Example Types of Classif		ine Learning
Learning					1	d Learning Exa	mples, Re	-
Learning				UN	NIT – V	d Learning Exa	mples, Re	-
-	on to AI :Wha	atIsAI?	,TheF		NIT – V		-	-
Introductio Intelligent/		andEnv	vironm	oundat ents,Go	NIT – V ionsofArti: podBehavic	ficialIntelligence our:TheConcepto	2	einforcement
Introductio Intelligent/	Agents:Agents nents,And the	andEnv	vironm	oundat ents,Go	NIT – V ionsofArti: podBehavic	ficialIntelligence our:TheConcepto	2	einforcement

University Press, 2020.

- 2. Introduction to Machine Learning, EthemAlpaydin, MIT Press, Second Edition, 2010.
- 3. StuartRusselland
 - PeterNorvig, "ArtificialIntelligence: AModernApproach", ThirdEdition, 2010. PearsonEduc ation.

Reference Books:

- 1. Introduction to the Theory of Statistics , A. M. Mood, F.A. Graybill and D.C. Boes, McGraw Hill Education, 2017.
- 2. A First Course in Probability, S.M Rose, Prentice Hall, 2012.
- 3. Introduction to Machine Learning with Python, Andreas C. Müller & Sarah Guido, OReily publications, 2016.

4. ArtificialIntelligence,ElaineRichie,KevinKnight, 2008,3rdEdition,TMH.

Web References:

1.https://www.tutorialspoint.com/machine_learning/machine_learning_tutorial.pdf 2.https://www.w3schools.com/python/python_ml_getting_started.asp

Question Paper Pattern:

Sessional Exam:

The question paper for sessional examination shall be for 25 marks, covering half of the syllabus for first sessional and remaining half for second sessional exam. The question paper shall consist of three sections with Two Questions (EITHER/ OR Type) in each section. The student shall answer one question from each section.

End Exam:

The question paper for End examination shall be for 60 marks. The Question paper shall contain Five Units with Two Questions (Either or Type) from each unit. Each of these questions may contain sub question and the student should answer any one question from each unit. Each Question carries 12 marks..

PYTHON FOR DATA SCIENCE (PDS)

			SBS				S	
Course Code	Category	Н	ours/W	Veek	Credits	Max	imum Mar	ks
HCS02	Н	L	Т	Р	С	Continuous Internal Assessment	End Exam	TOTAL
		3		2	4	40	60	100
	xam Duration :						d Exam Du	ration: 3 Hr
	tcomes : At the e							
		s using	the co	re conce	epts like Lis	ts, Dictionaries, so	ets, tuple, fu	inctions and
regular expr								
CO2: Demo	onstrate various r	nathem	atical	operation	ns on arrays	using NumPy		
CO3: Analy	yze and manipula	te Data	using	Pandas				
CO4: Creat	ing static and int	eractive	e visua	lizations	s using Matp	olotlib.		
CO5: Enum	nerate machine le	arning	algorit	hms, De	escribe the C	lassification and	Clustering	
				TI	I – TIV			
	gular expressions			-	•	ions, Modules ar		
Objects, Re Introduction Functions, A Comparison	gular expressions on to NumPy: T Aggregations: M as, Masks, and E	The Bas	sics of , and l	UN NumPy Everythi	IIT – II y Arrays, C ng in Betwo	Computation on Neen, Computation Sorting Arrays, S	JumPy Arra on Arrays:	ys: Universa Broadcasting
Objects, Rep Introduction Functions, A	gular expressions on to NumPy: T Aggregations: M as, Masks, and E	The Bas	sics of , and l	UN NumPy Everythi c, Fancy	IIT – II y Arrays, C ng in Betwo	omputation on N een, Computation	JumPy Arra on Arrays:	ys: Universa Broadcasting
Objects, Rey Introduction Functions, A Comparison Structured A Data Manij on Data in Append, Co	gular expressions on to NumPy: 7 Aggregations: Masks, and E Arrays pulation with Pa Pandas, Handlin	The Bas in, Max Boolean andas: ng Miss ts: Mer	sics of c, and I Logic Introdu- sing D ge and	UN Everythi , Fancy UN ucing Pa vata, Hie I Join, A mance I	NIT – II y Arrays, C ng in Betwo Indexing, S IT – III andas Object prarchical In Aggregation Pandas	omputation on N een, Computation	JumPy Arra on Arrays: tructured D	ys: Universa Broadcasting ata: NumPy's
Objects, Rey Introduction Functions, A Comparison Structured A Data Manij on Data in Append, Co	gular expressions on to NumPy: 7 Aggregations: Masks, and E hs, Masks, and E Arrays pulation with P Pandas, Handlin public Datase	The Bas in, Max Boolean andas: ng Miss ts: Mer	sics of c, and I Logic Introdu- sing D ge and	UN Everythi , Fancy UN ucing Pa vata, Hie I Join, A mance I	NIT – II y Arrays, C ng in Betwo Indexing, S IT – III andas Object prarchical In Aggregation	Computation on Neen, Computation Sorting Arrays, S ts, Data Indexing	JumPy Arra on Arrays: tructured D	ys: Universa Broadcasting ata: NumPy' on, Operating
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Objects, Rey Introduction Functions, A Comparison Structured A Data Manijon Data in Append, Co Vectorized S Visualization Plots, Visua Plot Legend	gular expressions on to NumPy: 7 Aggregations: Masks, and E as, Masks, and E Arrays pulation with Pandas, Handlin pmbining Datase String Operations on with Matplot alizing Errors, D ds, Customizing	The Bas in, Max Boolean andas: ng Miss ts: Mer s, High clib: Tw ensity a Colort	sics of a, and I Logic Introdu- sing D ge and -Perfor vo Inte and Co pars, N	UN Everythi Everythi Fancy UN ucing Pa vata, Hie Join, A mance I UN erfaces for ontour P Aultiple nd Style	NIT – II y Arrays, C ng in Betwe Indexing, S IT – III andas Object erarchical Ir Aggregation Pandas IT – IV or the Price lots, Histog Subplots, 7	Computation on Neen, Computation Sorting Arrays, S ts, Data Indexing adexing, Combini and Grouping Pl of One, Simple I rams, Binnings, a Text and Annota	NumPy Arra on Arrays: tructured D and Selecting Datasets anets Data, Line Plots, S and Density tion, Custo	ays: Universa Broadcasting ata: NumPy' on, Operatin s: Concat an Pivot Tables Simple Scatte , Customizin mizing Ticks

Text Books:

- 1. Python Basics: With Illustrations from the Financial Market, Vivek Krishnamoorthy, Jay Parmar, Mario Pisa Pena, AQuantInsti® Publication, 2020
- 2. Python Data Science Handbook:Essential Tools for Working with Data,JakeVanderPlas, O'reilly publications, 2016

Reference Books:

- 1. Python® for Programmers, Paul Deitel, Harvey Deitel, Pearson Education, Inc, 2019
- 2. Data Science & Big Data Analytics: Discovering, Analyzing, Visualizing and Presenting Data,David Dietrich, Barry Heller, Beibei Yang, Published by John Wiley & Sons, Inc, 2015

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	List of Experiments
1.	Data Types: Operations on Number data type, sets and tuples.
2.	Operators: Example programs using all operators.
3.	Conditional Statements: Usage of if-else -if statement, break, continue and pass.
4.	Loop Statements: Armstrong number, Palindrome.
5.	Loop Statements: Bubble sort using lists.
6.	Functions: Calculator program.
7.	Functions: Programs on anonymous functions.

8. NumPy: Arithmetic Operations on Arrays.

SOCIAL DATA ANALYTICS (SDA)

Course							S	cheme : 2020
Code	Category	Но	ours/W	eek	Credits	Max	imum Mar	ks
HCS03	Н	L	Т	Р	С	Continuous Internal Assessment	End Exam	TOTAL
		3		2	4	40	60	100
	xam Duration :						d Exam Du	ration: 3 Hrs
	tcomes : At the e							
	-	al medi	a lands	cape, ke	ey figures a	nd the methods co	onnecting the	e popular
social netwo								
		-				hodology, text min	-	lues :
						on customized cle		0
		-	-			ement using the c		of traffic and
		-				d technologies on		
		al topic	s from	Interne	et Forums a	and users interest	s from Pin	terest through
network ana								
CO5: Write	and execute Pyt	hon rea	dy to r	un scrip	ts and Sparl	k jobs on AWS		
				TT	I – TIN			
-	sualizing the data	a.					-	, , , ,
0			U .	puts. Ba	0	aning: APIs in a g techniques, Mon		ntroduction to
authentication social data	on techniques, Pa	arsing A	API out	puts. Ba	NIT – II	g techniques, Mon	ngoDB to sto	ntroduction to ore and access
authentications social data	on techniques, Paranti de la constructiva de la construcción de la con	arsing A	API out	puts. Ba UN , and E	NIT – II motions on	•	ngoDB to sto book brand	ntroduction to ore and access page, Project
authentication social data Uncovering planning, An Analysing	on techniques, Pa 5 Brand Activit nalysis, Keyword Twitter Using	arsing <i>A</i> y, Popu ds, Nou Sentime	API out	puts. Ba UN , and E ses, Dete alysis a	ASIC cleaning AIT – II motions on ecting trend and Entity	g techniques, Mon	book brand ncovering e	page, Project motions.
authentication social data Uncovering planning, An Analysing ' analysis: Cu	on techniques, Pa 5 Brand Activit nalysis, Keyword Twitter Using	arsing <i>A</i> y, Popu ds, Nou Sentime	API out	puts. Ba UN , and E ses, Deto alysis a Nameo	ASIC cleaning AIT – II motions on ecting trend and Entity	g techniques, Mon Facebook: Face s in time series, U Recognition: Ge	book brand ncovering e	ntroduction to ore and access page, Project motions. ata, Sentimen
authentication social data Uncovering planning, An Analysing 7 analysis: Cu analysis: Campaigns	on techniques, Pa g Brand Activit nalysis, Keyword Twitter Using ustomized sentin	y, Popu ds, Nou Sentime nent ar	API out Ilarity, n phras ent An nalysis, ion Ar	puts. Ba UN , and E ges, Deta alysis a Nameo UN nalytics	IT – II motions on ecting trend and Entity d entity rec IT – III on YouTu	g techniques, Mon Facebook: Face s in time series, U Recognition: Ge	book brand ncovering e etting the da ning NER a	ntroduction to ore and access page, Projec motions. ata, Sentimen and sentimen
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Scraping and Extracting Conversational Topics on Internet Forums: Getting the data, Data pull and pre-processing, Data analysis.

Demystifying Pinterest through Network Analysis of Users Interests: Getting the data, Data pull and pre-processing, Data analysis

UNIT – V

Social Data Analytics at Scale – Spark and Amazon Web Services: Different scaling methods and platforms, Topic models at scale, Spark on the Cloud – Amazon Elastic MapReduce

Text Books:

1. Python Social Media Analytics, Siddhartha Chatterjee, Michal Krystyanczuk, Packt Publishing Ltd, 2017

Reference Books:

1. Mastering Social Media Mining with Python, Marco Bonzanini, Packt Publishing Ltd, 2016

2. Thoughtful Data Science, David Taieb, Packt Publishing, June 2018.

Question Paper Pattern:

Sessional Exam:

The question paper for sessional examination shall be for 25 marks, covering half of the syllabus for first sessional and remaining half for second sessional exam. The question paper shall consist of three sections with Two Questions (EITHER/ OR Type) in each section. The student shall answer one question from each section.

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List of Experiments
1. Extract Data from Twitter Using Python and Twitter APIs
2. Extract Data from Face book Using Python and Face book APIs
3. Analyze Twitter Data using Sentiment Analysis and Entity Recognition.
 Analyze the YouTube data by 'Sentiment by weekday', 'Comments by weekday' after pre- processing
5. Extract Data from Twitter Using Python and Twitter APIs

NATURAL LANGUAGE PROCESSING (NLP)

Course Code		ST/C		alı	Cuadita	M		cheme: 2020
	Course Category	HOU	rs/We	eк	Credits	INI 8	aximum Ma	rks
HCS04	H	L	Т	Р	С	Continuous Internal Assessment	End Exam	Total
		3	0	2	4	40	60	100
Sessional Exam	Duration: 1 ¹ / ₂	Hrs				En	d Exam Dur	ation: 3 Hr
		0.1						
	omes: At the end							
	stand the pre proc	_		-		guage processin	g	
			•			al data		
	lassification and text tokenization							
	other topics relat							
	other topics rela		Inatura	I Lang	guage Floce	ssing		
				UN	IT– I			
	Natural Langua	0		0	•	-		
	okenization,Sten					ord removal ,R	are word ren	noval, Spell
correction.POS 7	Fagging and Nar	ned Ei	ntity R	_				
				UNI	T– II			
NLP Applicatio	n : Building your	first l	NLP A	pplicat	tion Other	NLP Applicatio	ns-Machine	Translation
Vector Space Mo	<u>odel, Probabilisti</u>	c Mod	lel.				Jiis-iviaeiiiiie	Translation,
Vector Space Mo	<u>odel, Probabilisti</u>	c Mod	lel.		T– III			Translation,
				UNI	T– III			
Fext Classificat	ion: Machine Le c Regression,SV	arning	g, Sam	UNI pling,	T− III Naïve Baye	es, Decision Tre	ees, Stochast	
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Fokenizing Tex Fokenizing Tex sentences into w Filtering stop wo emmas and sync Feature Extract classifier, Entrop Fopics in Brief:	ion: Machine Le c Regression,SV xt and WordNo vords, Tokenizin ords in a tokeniz onyms in WordN tion: BOW, TF-I oy classifier.	arning M, Ra et Ba g sent ed ser et. DF Fe	g, Sam ndom f sics: I sences ntence, eature f	UNI pling, forest UNI ntrodu using Look Extrac	T– III Naïve Baye algorithm ,T T– IV action, Toka regular exp ing up Syns tion. Trainin	enizing text in ressions, Train ressions, Train rests for a word	ees, Stochast K Means. nto sentence ning a senter in WordNet yes classifier	ic gradient s, Tokenizin ace Tokenize t, Looking t , Decision tr
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Text Classificat Descent, Logistic Tokenizing Tex sentences into w Filtering stop wo emmas and sync Feature Extract classifier, Entrop Topics in Brief: Translation, Que TextBooks : 1. Natural Lar	ion: Machine Le c Regression,SV xt and WordNo vords, Tokenizin ords in a tokeniz onyms in WordN tion: BOW, TF-I oy classifier.	arning M, Ra et Ba g sent ed ser et. DF Fe matio Syste g: Pyt	g , Sam ndom 1 sics: I sences ntence, eature 1 on Retri-	UNI pling, forest i UNI introdu using Look Extrac UNI ieval,	T– III Naïve Baye algorithm ,T T– IV Iction, Toke regular exp ing up Syns tion. Trainin IT– V Topic Mode	es, Decision Tra 'ext clustering- enizing text in pressions, Train sets for a word ng a Naive Bay eling, Text Sum	ees, Stochast K Means. nto sentence ning a senter in WordNet yes classifier	ic gradient s, Tokenizi ace Tokeniz t, Looking , Decision tr
 Fext Classificat Descent, Logistic Fokenizing Texternation Feature Stop were emmas and synce Feature Extract classifier, Entrop Fopics in Brief: Translation, Que TextBooks : Natural Lan Nitin Hard Practical Na NLP Syste 	ion: Machine Le c Regression,SVI xt and WordNo vords, Tokenizin ords in a tokeniz onyms in WordN tion: BOW, TF-I oy classifier. Search and Infor stion-Answering	arning M, Ra et Ba g sent ed ser et. DF Fe Trmation Syste g: Pyt 016. Proces	g , Sam ndom 1 sics: I sences ntence, eature 1 on Retri- ms hon an ssing: 2	UNI pling, forest i UNI introdu using Look Extrac UNI ieval, i ieval, i	T– III Naïve Baye algorithm ,T T– IV Iction, Tok regular exp ing up Syns tion. Trainin IT– V Topic Mode	enizing text in oressions, Trains ets for a word ng a Naive Bay eling, Text Sum Chopra, Jacob I Guide to Build	ees, Stochast K Means. nto sentence ning a senter in WordNet yes classifier. marization, I Perkins, and ling Real-Wo	ic gradient s, Tokenizi ace Tokeniz t, Looking , Decision tr Machine

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List of Experiments

- 1. Perform Text pre processing using NLP techniques
- 2. Program using POS Tagging and Named Entity Recognition
- 3. Program using Machine Translation.
- 4. Perform text classification using Machine Learning classifiers
- 5. Apply Feature Extraction techniques on textual dataset