NOIDA INSTITUTE OF ENGINEERING AND TECHNOLOGY, GREATER NOIDA (An Autonomous Institute)



Affiliated to

DR. A.P.J. ABDUL KALAM TECHNICAL UNIVERSITY UTTAR PRADESH, LUCKNOW



Evaluation Scheme & Syllabus For

Bachelor of Technology Computer Science and Engineering (Artificial Intelligence) Fourth Year

(Effective from the Session: 2024-25)

NOIDA INSTITUTE OF ENGG. & TECHNOLOGY, GREATER NOIDA, GAUTAM BUDDH NAGAR (AN AUTONOMOUS INSTITUTE)

Bachelor of Technology Computer Science and Engineering (Artificial Intelligence) <u>EVALUATION SCHEME</u> SEMESTER-VII

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S.	Subject	Subject Name	Type of Subject]	Perio	ods		Evalua	tion Schem	e	Seme	ster	Total	Credit
No.	Codes	Subject Name	-91-0-2003000	L	Т	Р	СТ	TA	TOTAL	PS	TE	PE		
			WEEKS COMPULSOR	Y IN	NDU	CTIC	ON PRO	GRAM	[
1	ACSML0702	Deep Learning	Mandatory	3	0	0	30	20	50		100		150	3
2		Departmental Elective-V	Departmental Elective	3	0	0	30	20	50		100		150	3
3		Open Elective-II	Open Elective	3	0	0	30	20	50		100		150	3
4		Open Elective-III	Open Elective	3	0	0	30	20	50		100		150	3
5	ACSML0752	Deep Learning Lab	Mandatory	0	0	2				25		25	50	1
6	ACSE0759	Internship Assessment-III	Mandatory	0	0	2				50			50	1
		*Massive Open Online Courses	*MOOC											
		(For B.Tech. Hons. Degree)	*MOOCs											
		GRAND TOTAL											700	14

* List of Recommended MOOCs (Massive Open Online Courses) for Final Year B. Tech Students (Semester-VII)

S. No.	Subject Code	Course Name (Block chain)	University / Industry Partner Name	No of HOURS	Credits
1.	AMC0278	Natural Language Processing using Python	Infosys Wingspan (Infosys Springboard)	15h 45m	1
2.	AMC0279	Spring Boot and Angular-React Stack -DevOps Tools and Capstone Project	Infosys Wingspan (Infosys Springboard)	107h 50m	4

PLEASE NOTE:-

• Internship (3-4 weeks) shall be conducted during summer break after semester-VI and will be assessed during Semester-VII

Abbreviation Used:

L: Lecture, T: Tutorial, P: Practical, CT: Class Test, TA: Teacher Assessment, PS: Practical Sessional, TE: Theory End Semester Exam., PE: Practical End Semester Exam, CE: Core Elective, OE: Open Elective, DE: Departmental Elective, CA: Compulsory Audit, MOOCs: Massive Open Online Courses.

	List of D	epartmental Electiv	/es	1	
Subject Codes	Subject Name	Type of Subject	Bucket Name	Branch	Semester
ACSE0712	RPA Implementation	Departmental Elective-V	CRM-RPA	AI	7
ACSAI0712	Natural Language Processing	Departmental Elective-V	Data Analytics	AI	7
ACSE0713	Web Development using MERN Stack with DevOps	Departmental Elective-V	Full Stack Development	AI	7
ACSE0711	Game Programming	Departmental Elective-V	Mobility Management	AI	7

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Bachelor of Technology Computer Science and Engineering (Artificial Intelligence) <u>EVALUATION SCHEME</u> SEMESTER-VIII

S. No	Subject		Type of Subject		Perio	ds		Evalu	ation Sch	eme	End	Semest		
	Codes	Subject Name	Type of Subject	L	Т	Р	СТ	TA	TOTAL	PS	ТЕ	PE	Total	Credit
1		Open Elective-IV	Open Elective	2	0	0	30	20	50		100		150	2
2	ACSE0859/ ACSE0858	Capstone Project/Industrial Internship	Mandatory	0	0	20				200		300	500	10
		*Massive Open Online Courses (For B.Tech. Hons. Degree)	*MOOCs											
		TOTAL											650	12

* List of Recommended MOOCs (Massive Open Online Courses) for Final Year B. Tech Students (Semester-VIII)

S.no	Subject Code	Course Name	University/Industry Partner Name	No. of Hours	Credit
1	AMC0280	Natural Laungage Processing for developers	Infosys Wingspan (Infosys Springboard)	13 h 37 m	1
2	AMC0281	Angular Developer Program	Infosys Wingspan (Infosys Springboard)	48h 36m	4

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Bachelor of Technology Computer Science and Engineering (Artificial Intelligence)

AICTE Guidelines in Model Curriculum:

A student will be eligible to get Under Graduate degree with Honours only, if he/she completes the additional MOOCs courses such as Coursera certifications, or any other online courses recommended by the Institute (Equivalent to 20 credits). During Complete B.Tech. Program Guidelines for credit calculations are as follows.

- 1. For 6 to 12 HOURS =0.5 Credit
- 2. For 13 to18 =1 Credit
- 3. For 19 to 24 =1.5 Credit
- 4. For 25 to 30 =2 Credit
- 5. For 31 to 35 =2.5 Credit
- 6. For 36 to 41 =3 Credit
- 7. For 42 to 47 =3.5 Credit
- 8. For 48 and above =4 Credit

For registration to MOOCs Courses, the students shall follow Coursera registration details as per the assigned login and password by the Institute these courses may be cleared during the B. Tech degree program (as per the list provided). After successful completion of these MOOCs courses, the students shall provide their successful completion status/certificates to the Controller of Examination (COE) of the Institute through their coordinators/Mentors only.

The students shall be awarded Honours Degree as per following criterion.

- i. If he / she secures 7.50 as above CGPA.
- ii. Passed each subject of that degree program in the single attempt without any grace.
- iii. Successful completion of MOOCs based 20 credits.

B. TECH FOURTH YEAR

Course code	ACSML0702 L	T P	Credits
Course title	DEEP LEARNING 3	0 0	3
	e: To be able to learn unsupervised techniques and provide continuous in various datasets with more reliable and concise analysis results.	nproven	nent in accuracy
Pre-requisites:	Python, Basic Modeling Concepts.		
	Course Contents / Syllabus		1
UNIT-I	INTRODUCTION		8 HOURS
Precision, Recall – Grid search, ray Artificial Neura functions, Neura networks. Variou	Regression - MAE, MSE, RMSE, R Squared, Adjusted R Squared, p , F1, Other topics, K-Fold Cross-validation, RoC curve, Hyper-Parame ndom search, Introduction to Deep Learning. I Network: Neuron, Nerve structure and synapse, Artificial Neuron and network architecture: Single layer and Multilayer feed-forward network is learning techniques; Perception and Convergence rule, Hebb Learnin ptron, Gradient descent, and the Delta rule, Multilayer networks, Deriv Algorithm.	eter Tuni Id its mo orks, recu Ig. Perce	ng Introduction del, activation urrent eptron's,
UNIT-II	CONVOLUTION NEURAL NETWORK		8 HOURS
neural net, Explo	er vision? Why Convolutions (CNN)? Introduction to CNN, Training oring the design space for convolutional nets, Pooling layer motivatio yered application, Understanding and visualizing a CNN, Transfer la sification, Text classification, Image classification and hyper-parameter	on in CN earning	N, Designing a and fine-tuning
UNIT-III	DETECTION & RECOGNITION		8 HOURS
	Detection, Strided Convolutions, Networks in Networks and 1x1 (tion, Object Detection, and YOLO Algorithm.	Convolu	tions, Inception
UNIT-IV	RECURRENT NEURAL NETWORKS		8 HOURS
Different types of	e models? Recurrent Neural Network Model, Notation, Back-propagat of RNNs, Language model and sequence generation, Sampling nove NNs, Gated Recurrent Unit (GRU), Long Short-Term Memory (LST)	el sequer	nces, Vanishing
UNIT-V	AUTOENCODERS IN DEEP LEARNING		8 HOURS
	nd unsupervised learning, Stacked auto-encoders and semi-supervised Dropout and Batch normalization.	learning	,
Course outco	me: After completion of this course students will be able to		
CO 1	Analyze the ANN model and understand the ways of accuracy measured	ement.	K4

CO 2	Develop a convolutional neural network for multi-class classification in images.	K6
CO 3	Apply a Deep Learning algorithm to detect and recognize an object.	К3
CO 4	Apply RNNs to Time Series Forecasting, NLP, Text and Image Classification.	К3
CO 5	Apply Lower-dimensional representation over higher-dimensional data for dimensionality reduction and capture the important features of an object.	К3
Textbooks:		
1. Zurada	and Jacek M, "Introduction to Artificial Neural Systems", West Publishing (780534954604	Company, 1992,
	C. M. Neural Networks for Pattern Recognition. Oxford University Press. 199	95.
	laykin, "Neural Networks and Learning Machines" Third Edition	
	arning", I Goodfellow, Y Bengio and A Courville, 1st Edition 2016	
	tion to Machine Learning with Python ", by Andreas C. Müller, Sarah Guido	
	p Learning with Python by François Chollet 1st Edition	
Reference Bo	oks:	
1. Aston Zha 0.17.4	ng, Zachary C. Lipton, Mu Li, and Alexander J. Smola "Dive into Deep Learn	ing", Release
2. Artificial Intelligen	Intelligence: A Modern Approach. Prentice Hall Series in Arti Russell, S. and nce. 2003.	l Norvig, N. Arti
	tube/ Faculty Video Link:	
Unit 1	 (371) Lec-1 Introduction to Artificial Neural Networks - YouTube (3) Deep Learning(CS7015): Lec 8.1 Bias and Variance - YouTube (3) Mod-10 Lec-39 Assessing Learnt classifiers; Cross Validation; - YouTu (3) Lec-1 Introduction to Artificial Neural Networks - YouTube (3) Lec-2 Artificial Neuron Model and Linear Regression - YouTube (3) Evaluation and Cross-Validation - YouTube 	<u>be</u>
Unit 2	 (3) Lecture 1 Introduction to Convolutional Neural Networks for Visua YouTube (3) Lecture 2 Image Classification - YouTube (3) Lecture 3 Loss Functions and Optimization - YouTube (3) Hyperparameter optimization - YouTube (3) Deep Learning(CS7015): Lec 11.3 Convolutional Neural Networks - YouTube 	
Unit 3	(3) C4W3L09 YOLO Algorithm - YouTube (3) Edge Detection - YouTube (3) Neural Networks - Networks in Networks and 1x1 Convolutions - YouT	
Unit 4	 (3) Backpropagation in CNNs - YouTube (3) Deep RNNs and Bi- RNNs - YouTube (3) Deep Learning(CS7015): Lec 13.4 The problem of Exploding and Vanis YouTube (3) Deep Learning(CS7015): Lec 14.2 Long Short Term Memory(LST Recurrent Units(GRUs) - YouTube 	hing Gradients -
Unit 5	(3) Deep Learning(CS7015): Lec 7.1 Introduction to Autoncoders - YouTu (3) Deep Learning(CS7015): Lec 9.5 Batch Normalization - YouTube (3) Deep Learning(CS7015): Lec 7.3 Regularization in autoencoders YouTube	

	B. TECH FOURTH YEAR	
Course code	ACSML0752 LT P	Credit
Course title	DEEP LEARNING LAB0 0 2	1
	Suggested list of Experiments	
Sr. No.	Name of Experiment	CO
1	Write a program Print Dimensions of dataset	CO1
2.	Write a program to Calculate of Accuracy Values.	CO1
3.	Write a program to Build an Artificial Neural Network by implementing the	CO1
	Backpropagation algorithm and test the same using appropriate data sets.	
4.	Write a program to Compose Matrix Shape and Tensor Shape.	CO2
5.	Write a program to showing Accessing and Manipulation of tensors.	CO2
6.	Write a program to understand the mechanism of practically training a binary classifier.	CO2
7.	Implement with a program showing Access and manipulation of tensors.	CO2
8.	Write a program to show Regression Data Sampling.	CO2
9.	Write a program to Combat Overfitting.	CO1
10.	Write a program Print Dimensions of dataset.	CO2
11.	Write a program to Calculate of Accuracy Values.	CO2
12.	Write a program to Build an Artificial Neural Network by implementing the	CO1
	Backpropagation algorithm and test the same using appropriate data sets.	
13.	Write a program to build a simple autoencoder based on a fully-connected layer	CO3
	in Keras.	
14.	Implement Long Short-Term Memory Networks using sample data.	CO1
15.	Write a program showing Automatic Image Captioning with KerasFacial	CO3
	Recognition.	
Lab Course Ou	utcome: After completion of this course students will be able to	
C01	Develop python programs to work on Data sets and Implement Artificial	K6
	Neural Network Techniques.	
CO2	Explore different types of tensors and perform exploratory data analysis on	K4
	different data sets.	
CO3	Apply Automatic Image Captioning with KerasFacial Recognition.	K3

	B. TECH FOURTH YEAR	
Course code	ACSE0712 LTP	Credits
Course title	RPA IMPLEMENTATION3 0 0	3
	ve: This course is designed to give a thorough understanding and practical skills in offware robots for Robotic Process Automation (RPA).	developing
Pre-requisites:	Basic Knowledge of C Programming	
	Course Contents / Syllabus	
UNIT-I	DATA MANIPULATION	8 HOURS
Manipulation, G	Data Manipulation, Scalar variables, collections and Tables, Text Manipula athering and Assembling Data Recording and Advanced UI Interaction; Recording Ir op Recording, Web Recording, Input/output Methods, Screen Scraping, Data Scrapin ques. SELECTORS	troduction,
Selectors, Defin RPA Challenge, based automatio	ing and Assessing Selectors, Customization, Debugging, Dynamic Selectors, Partia Image, Text & Advanced Citrix Automation, Introduction to Image & Text Automat n, Keyboard based automation, Information Retrieval, Advanced Citrix Automation sing tab for Images Starting Apps. DATA TABLES AND AUTOMATION	l Selectors, ion, Image-
	les & PDF, Data Tables in RPA, Excel and Data Table Basics Data Manipulatio	
Email Automati	from PDF, extracting a single piece of data, Anchors, Using anchors in PDF. on: Email Automation, Incoming Email automation, Sending Email automation.	
UNIT-IV	DEDUCCING AND EXCEPTION HANDI INC	
	DEBUGGING AND EXCEPTION HANDLING	8 HOURS
	s, Strategies for solving issues, Catching errors.	8 HOURS
Orchestrator: Te UNIT-V	s, Strategies for solving issues, Catching errors. nants, Authentication, Users, Roles, Robots, Environments, Queues & Transactions, ROBOTIC FRAMEWORK cemplate, Re-Framework template works, Use Re-Framework to automate your owr	Schedules. 8 HOURS
Orchestrator: Te UNIT-V Re-Framework .NET Classes ar	s, Strategies for solving issues, Catching errors. nants, Authentication, Users, Roles, Robots, Environments, Queues & Transactions, ROBOTIC FRAMEWORK cemplate, Re-Framework template works, Use Re-Framework to automate your owr	Schedules. 8 HOURS
Orchestrator: Te UNIT-V Re-Framework .NET Classes ar	s, Strategies for solving issues, Catching errors. nants, Authentication, Users, Roles, Robots, Environments, Queues & Transactions, ROBOTIC FRAMEWORK remplate, Re-Framework template works, Use Re-Framework to automate your owr ad Objects.	Schedules. 8 HOURS
Orchestrator: Te UNIT-V Re-Framework .NET Classes ar Course outcom	s, Strategies for solving issues, Catching errors. nants, Authentication, Users, Roles, Robots, Environments, Queues & Transactions, ROBOTIC FRAMEWORK remplate, Re-Framework template works, Use Re-Framework to automate your owr id Objects. e: After completion of this course students will be able to: Apply basic concepts and methods from design engineering to explore creative	Schedules. 8 HOURS
Orchestrator: Te UNIT-V Re-Framework .NET Classes ar Course outcom	s, Strategies for solving issues, Catching errors. nants, Authentication, Users, Roles, Robots, Environments, Queues & Transactions, ROBOTIC FRAMEWORK template, Re-Framework template works, Use Re-Framework to automate your owr d Objects. e: After completion of this course students will be able to: Apply basic concepts and methods from design engineering to explore creative solutions of real-world problems. Learn Robotic Process Automation, and massive career opportunity in this field. Implement the knowledge of RPA tools, functions in various industries and	Schedules. 8 HOURS n processes. K3
Orchestrator: Te UNIT-V Re-Framework to .NET Classes ar Course outcom CO 1 CO 2	s, Strategies for solving issues, Catching errors. nants, Authentication, Users, Roles, Robots, Environments, Queues & Transactions, ROBOTIC FRAMEWORK remplate, Re-Framework template works, Use Re-Framework to automate your own ad Objects. e: After completion of this course students will be able to: Apply basic concepts and methods from design engineering to explore creative solutions of real-world problems. Learn Robotic Process Automation, and massive career opportunity in this field.	Schedules. 8 HOURS n processes. K3 K2

Textbooks:

- 1) Vaibhav Jain, "Crisper Learning: For UiPath", Latest Edition, Independently Published, 2018.
- 2) Alok Mani Tripathi, "Learning Robotics Process Automation", Latest Edition, Packt Publishing ltd, Birmingham. March 2018

Reference Books/E-Books:

- 1) Kelly Wibbenmeyer, "The Simple Implementation Guide to Robotic Process Automation (RPA)", Latest Edition, iUniverse Press.
- 2) https://www.uipath.com/hubfs/ebook-its-time-to-automate.pdf

Links:

https://www.youtube.com/watch?v=6QoCG6YIPVo&list=PL41Y-9S9wmyJarNN2KnB4XudpT1yE1kVd

https://www.youtube.com/watch?v=YOHFgrOvPTM&list=PL41Y-9S9wmyLvF6Ou0oPhg6MrFWSw7sn4

https://www.youtube.com/watch?v=QMBuyLMjOhM&list=PL41Y-9S9wmyIYX6kciM8DboVYymsv2y6K

https://www.youtube.com/watch?v=KE9raKNTkfI&list=PL41Y-9S9wmyLeXL1DY9j-XepNb_vg9N8t

https://www.youtube.com/watch?v=2rjr8QhD9oc&list=PL41Y-9S9wmyJi2zmWY77yPZrdVI7ab3Ja

Course code	ACSAI0712 LTP	Credits
Course title	NATURAL LANGUAGE PROCESSING3 0 0	3
	e: The course aims to provide an understanding of the foundational concepts and is is on providing application-based knowledge.	d techniques
Pre-requisites: Learning.	Programming Skills, Data Structures, Algorithms, Probability and Statis	tics, Machine
	Course Contents / Syllabus	
UNIT-I	OVERVIEW OF NATURAL LANGUAGE PROCESSING	8 HOURS
Definition, Appli	cations and emerging trends in NLP, Challenges. Ambiguity.	
NLP tasks using	NLTK: Tokenization, stemming, lemmatization, stop-word removal, POS tag	ging. Parsing
	cognition, coreference resolution.	
UNIT-II	REGULAR EXPRESSIONS	8 HOURS
-	ng: Using Python - Convert to lower case, handle email-id, HTML tags, URLs, alization of data (contractions, standardize) etc.	emojis, repea
• •	pora, and linguistic resources, Linguistic foundations: Morphology, syntax, s	semantics and
pragmatics, Lang	uage models: Unigram, Bigram, N-grams.	
pragmatics, Lang UNIT-III	uage models: Unigram, Bigram, N-grams. TEXT ANALYSIS AND SIMILARITY	8 HOURS
pragmatics, Lang UNIT-III	uage models: Unigram, Bigram, N-grams.	8 HOURS
pragmatics, Lang UNIT-III Text Vectorizatio	uage models: Unigram, Bigram, N-grams. TEXT ANALYSIS AND SIMILARITY	8 HOURS
oragmatics, Lang U NIT-III Text Vectorizatio Fextual Similarit	uage models: Unigram, Bigram, N-grams. TEXT ANALYSIS AND SIMILARITY on: Bag-of-Words model and vector space models, Term Presence, Term Frequence	8 HOURS
Dragmatics, Lang UNIT-III Fext Vectorization Fextual Similarity UNIT-IV Fext classification	TEXT ANALYSIS AND SIMILARITY on: Bag-of-Words model and vector space models, Term Presence, Term Frequency: Cosine similarity, Word Mover's distance, Word embeddings: Word2Vec, Gl TEXT CLASSIFICATION & NLP APPLICATIONS on: Implement of applications of NLP using text classification- Sentiment Application	8 HOURS ncy, TF-IDF oVe. 8 HOURS
Dragmatics, Lang UNIT-III Text Vectorization Textual Similarity UNIT-IV Text classification	TEXT ANALYSIS AND SIMILARITY on: Bag-of-Words model and vector space models, Term Presence, Term Frequency: Cosine similarity, Word Mover's distance, Word embeddings: Word2Vec, Gl TEXT CLASSIFICATION & NLP APPLICATIONS on: Implement of applications of NLP using text classification- Sentiment Application	8 HOURS ncy, TF-IDF oVe. 8 HOURS
Dragmatics, Lang UNIT-III Text Vectorization Textual Similarity UNIT-IV Text classification modelling, Spam	uage models: Unigram, Bigram, N-grams. TEXT ANALYSIS AND SIMILARITY on: Bag-of-Words model and vector space models, Term Presence, Term Frequency: Cosine similarity, Word Mover's distance, Word embeddings: Word2Vec, Gl TEXT CLASSIFICATION & NLP APPLICATIONS on: Implement of applications of NLP using text classification- Sentiment Andetection.	8 HOURS ncy, TF-IDF oVe. 8 HOURS nalysis, Topio
pragmatics, Lang UNIT-III Text Vectorizatio Textual Similarit UNIT-IV Text classificatio modelling, Spam High Level NLP	TEXT ANALYSIS AND SIMILARITY on: Bag-of-Words model and vector space models, Term Presence, Term Frequency: Cosine similarity, Word Mover's distance, Word embeddings: Word2Vec, Gl TEXT CLASSIFICATION & NLP APPLICATIONS on: Implement of applications of NLP using text classification- Sentiment Application	8 HOURS ncy, TF-IDF oVe. 8 HOURS nalysis, Topio
Dragmatics, Lang UNIT-III Fext Vectorization Fextual Similarit UNIT-IV Fext classification nodelling, Spam High Level NLP Dialog systems, o	mage models: Unigram, Bigram, N-grams. TEXT ANALYSIS AND SIMILARITY on: Bag-of-Words model and vector space models, Term Presence, Term Frequency: Cosine similarity, Word Mover's distance, Word embeddings: Word2Vec, Gl TEXT CLASSIFICATION & NLP APPLICATIONS on: Implement of applications of NLP using text classification- Sentiment Andetection. applications: Machine translation: Rule-based and statistical approaches, Text sconversational agents and chatbots.	8 HOURS ncy, TF-IDF oVe. 8 HOURS nalysis, Topic
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Dragmatics, Lang UNIT-III Fext Vectorization Fextual Similarity UNIT-IV Fext classification modelling, Spam High Level NLP Dialog systems, on UNIT-V Sequential data,	uage models: Unigram, Bigram, N-grams. TEXT ANALYSIS AND SIMILARITY on: Bag-of-Words model and vector space models, Term Presence, Term Frequency: Cosine similarity, Word Mover's distance, Word embeddings: Word2Vec, Gl TEXT CLASSIFICATION & NLP APPLICATIONS on: Implement of applications of NLP using text classification- Sentiment Ardetection. applications: Machine translation: Rule-based and statistical approaches, Text sconversational agents and chatbots. ADVANCED NLP TECHNIQUES Introduction to sequence models - RNN and LSTM, Attention Mechanism,	8 HOURS ncy, TF-IDF oVe. 8 HOURS nalysis, Topic summarization 8 HOURS Transformer
Dragmatics, Lang UNIT-III Fext Vectorization Fextual Similarity UNIT-IV Fext classification modelling, Spam High Level NLP Dialog systems, on UNIT-V Sequential data,	mage models: Unigram, Bigram, N-grams. TEXT ANALYSIS AND SIMILARITY on: Bag-of-Words model and vector space models, Term Presence, Term Frequency: Cosine similarity, Word Mover's distance, Word embeddings: Word2Vec, Gl TEXT CLASSIFICATION & NLP APPLICATIONS on: Implement of applications of NLP using text classification- Sentiment Andetection. applications: Machine translation: Rule-based and statistical approaches, Text sconversational agents and chatbots. ADVANCED NLP TECHNIQUES	8 HOURS ncy, TF-IDF oVe. 8 HOURS nalysis, Topi summarization 8 HOURS Transformer
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Dragmatics, Lang UNIT-III Fext Vectorization Fextual Similarity UNIT-IV Fext classification nodelling, Spam High Level NLP Dialog systems, of UNIT-V Sequential data, Fransformer-base Course outcome	uage models: Unigram, Bigram, N-grams. TEXT ANALYSIS AND SIMILARITY on: Bag-of-Words model and vector space models, Term Presence, Term Frequency: Cosine similarity, Word Mover's distance, Word embeddings: Word2Vec, Gl TEXT CLASSIFICATION & NLP APPLICATIONS on: Implement of applications of NLP using text classification- Sentiment And detection. applications: Machine translation: Rule-based and statistical approaches, Text s conversational agents and chatbots. ADVANCED NLP TECHNIQUES Introduction to sequence models - RNN and LSTM, Attention Mechanism, ed models: BERT, GPT, T5, Introduction to Hugging Face Transformers, Case s e: After completion of this course students will be able to:	8 HOURS ncy, TF-IDF oVe. 8 HOURS nalysis, Topic summarization 8 HOURS Transformentudies.
Dragmatics, Lang UNIT-III Fext Vectorization Fextual Similarity UNIT-IV Fext classification nodelling, Spam High Level NLP Dialog systems, of UNIT-V Sequential data, Fransformer-base Course outcome	uage models: Unigram, Bigram, N-grams. TEXT ANALYSIS AND SIMILARITY on: Bag-of-Words model and vector space models, Term Presence, Term Frequency: Cosine similarity, Word Mover's distance, Word embeddings: Word2Vec, Gl TEXT CLASSIFICATION & NLP APPLICATIONS on: Implement of applications of NLP using text classification- Sentiment Ardetection. applications: Machine translation: Rule-based and statistical approaches, Text sconversational agents and chatbots. ADVANCED NLP TECHNIQUES Introduction to sequence models - RNN and LSTM, Attention Mechanism, ed models: BERT, GPT, T5, Introduction to Hugging Face Transformers, Case s :: After completion of this course students will be able to: Appreciate the emerging trends and challenges in NLP and perform the basic	8 HOURS ncy, TF-IDF oVe. 8 HOURS nalysis, Topic summarization 8 HOURS Transformer
Dragmatics, Lang UNIT-III Fext Vectorization Fextual Similarity UNIT-IV Fext classification nodelling, Spam High Level NLP Dialog systems, of UNIT-V Sequential data, Fransformer-base Course outcome CO 1	TEXT ANALYSIS AND SIMILARITY on: Bag-of-Words model and vector space models, Term Presence, Term Frequency: Cosine similarity, Word Mover's distance, Word embeddings: Word2Vec, Gl TEXT CLASSIFICATION & NLP APPLICATIONS on: Implement of applications of NLP using text classification- Sentiment Audetection. applications: Machine translation: Rule-based and statistical approaches, Text sconversational agents and chatbots. ADVANCED NLP TECHNIQUES Introduction to sequence models - RNN and LSTM, Attention Mechanism, ed models: BERT, GPT, T5, Introduction to Hugging Face Transformers, Case s :: After completion of this course students will be able to: Appreciate the emerging trends and challenges in NLP and perform the basic NLP tasks using some NLP library.	8 HOURS ncy, TF-IDF oVe. 8 HOURS nalysis, Topi summarization 8 HOURS Transformentudies.
pragmatics, Lang UNIT-III Text Vectorization Textual Similarity UNIT-IV Text classification modelling, Spam High Level NLP Dialog systems, or UNIT-V Sequential data, Transformer-base	uage models: Unigram, Bigram, N-grams. TEXT ANALYSIS AND SIMILARITY on: Bag-of-Words model and vector space models, Term Presence, Term Frequency: Cosine similarity, Word Mover's distance, Word embeddings: Word2Vec, Gl TEXT CLASSIFICATION & NLP APPLICATIONS on: Implement of applications of NLP using text classification- Sentiment Ardetection. applications: Machine translation: Rule-based and statistical approaches, Text sconversational agents and chatbots. ADVANCED NLP TECHNIQUES Introduction to sequence models - RNN and LSTM, Attention Mechanism, ed models: BERT, GPT, T5, Introduction to Hugging Face Transformers, Case s :: After completion of this course students will be able to: Appreciate the emerging trends and challenges in NLP and perform the basic	8 HOURS ncy, TF-IDF oVe. 8 HOURS nalysis, Topi summarizatio 8 HOURS Transformentudies. K2

CO4	Implement NLP techniques to design real-world NLP applications	K6
CO 5	Apply advanced techniques like sequential modelling and attention	K6
	mechanism to develop NLP applications	

Textbooks:

1)Daniel Jurafsky, James H. Martin, "Speech and Language Processing", Second Edition, Pearson Education, 2009 ISBN 0131873210.

2)James Allen, Natural Language Understanding, 2nd edition, 1995 Pearson Education ISBN 13: 9780805303346.

3)Akshar Bharti, Vineet Chaitanya and Rajeev Sangal, NLP: A Paninian Perspective,1st edition1995, Prentice ISSBN 9788120309210

Reference Books:

1)Christopher D.Manning and Hinrich Schutze,, "Foundations of Statistical Natural Language Processing", MIT Press, 1999 Second Edition, ISBN No. 0-262-13360-1.

2)T. Winograd, Language as a Cognitive Process, 1st edition, 1983 Addison- Wesley ISBN 020108-571-2 3)L.M. Ivansca, S. C. Shapiro, Natural Language Processing and Knowledge Representation, 2nd edition, 2000 AAAI Press ISBN-13: 978-0262590211

Links:

1) https://realpython.com/nltk-nlp-python/

2) https://www.coursera.org/lecture/python-text-mining/basic-nlp-tasks-with-nltk-KD8uN

3) https://www.coursera.org/lecture/nlp-sequence-models/learning-word-embeddings-APM5s

4) https://www.coursera.org/projects/regular-expressions-in-python

5) https://www.coursera.org/learn/python-text-mining/lecture/sVe8B/regular-expressions

	B.TECH FOURTHYEAR	
Subjec	t Code: ACSE0713	LT P
Subjec	t Couc. Meshov 15	300
Subjec	t Name:Web Development using MERN Stack	Credits
-	with DevOps	3
Course	Objective: This course focuses on how to design and build static as well as dynamic	web
pages an	d interactive web applications. Students can understand how to put them together to tack application.	
Pre- req	uisites: Student should have the knowledge of HTML, CSS and ES6	
	Course Contents/Syllabus	
	Introduction to React JS:	
Unit-1	Overview of frameworks, NPM commands, React App, Project Directory Structure,	8
	React Component Basic, Understanding JSX, Props and State, Stateless and Stateful	Hours
	Components, Component life cycle, Hooks, react-router vs react-router-dom,	
	Connecting React with mongodB: Google Material UI, AppBar, Material UI's Toolbar, NavBar, Material UI Buttons, SQL	8
Unit-2	and Complex Transactions, Dynamic Schema, create Index (), get Indexes () & drop	Hours
01111-2	Index (), Replication, Statement-based vs. Binary Replication,	IIUuis
	Auto-Sharding and Integrated Caching, Load balancing, Aggregation, scalability.	
	Node js & Express Framework:	
	Introduction, Environment Setup, serving static resources, template engine with vash and	8
Unit-3	jade, Connecting Node.js to Database, Mongoose Module, Creating Rest APIs, Express	Hours
	Framework, MVC Pattern, Routing, Cookies and Sessions, HTTP Interaction, User	
	Authentication	
	Evolution of DevOps:	8
Unit-4	DevOps Principles, DevOps Lifecycle, DevOps Tools, and Benefits of DevOps,	o Hours
UIIIt-4	SDLC (Software Development Life Cycle) models, Lean, ITIL and Agile Methodology, Agile vs DevOps, Process flow of Scrum Methodologies, Project planning, scrum testing,	Hours
	sprint Planning and Release management, Continuous Integration and Delivery pipeline.	
	CI/CD concepts (GitHub, Jenkins, Sonar):	
	GitHub, Introduction to Git, Version control system, Jenkins Introduction, Creating Job	8
Unit-5	in Jenkins, adding plugin in Jenkins, Creating Job with Maven & Git, Integration of	Hours
	Sonar, Dockers, Containers Image: Run, pull, push containers, Container lifecycle,	
	Introduction to Kubernetes.	
Course	• Outcomes –	
<u>CO1</u>	Apply the knowledge of ES6 that are vital to implement react employed and the week	120
CO1	Apply the knowledge of ES6 that are vital to implement react application over the web. Implement and understand the impact of web designing by database connectivity with	K3
CO2	Mongodb.	K3
	Explain, analyze and apply the role of server-side scripting language like Nodejs and	K4
CO3	Express js framework	IX7
	Identify the benefits of DevOps over other software development processes to Gain	K2
CO4	insights into the DevOps environment.	112
COF	Demonstrate popular open-source tools with features and associated terminology used to	K3
CO5	perform Continuous Integration and Continuous Delivery.	_
Textbo	· · · · · · · · · · · · · · · · · · ·	
		on.

2.	Mohan Mehul, "Advanced Web Development with React", 2 nd Edition 2020, BPB Publications.
3.	Dhruti Shah, "Comprehensive guide to learn Node.js", 1 st Edition, 2018 BPB Publications.
	Jennifer Davis, Ryn Daniels, "Effective DevOps: Building, Collaboration, Affinity, and
	Tooling at Scale",1 st Edition, 2016, O'Reilly Media Publication.
5	John Edward Cooper Berg, "DevOps. Building CI/CD Pipelines with Jenkins, Docker
5.	Container, AWS (Amazon Web Services) ECS, JDK 11, Git and Maven 3, Sonar, Nexus", Kindle
	Edition,2019, O'Reilly Media Edition.
Reference Books:	
1.	Anthony Accomazzo, Ari Lerner, and Nate Murray, "Fullstack React: The Complete Guide to ReactJS and Friends", 4th edition, 2020 International Publishing.
2.	David Cho, "Full-Stack React, Type Script, and Node: Build cloud-ready web applications using React 17 with Hooks and GraphQL", 2nd edition, 2017 Packt Publishing Limited.
3.	Richard Haltman & Shubham Vernekar, "Complete node.js: The fast guide: Learn complete backend development with node.js"5th edition, 2017 SMV publication.
4.	Glenn Geenen, Sandro Pasquali, Kevin Faaborg, "Mastering Node.js: Build robust and scalable real-time server-side web applications efficiently" 2nd edition Packt,2017 Publishing Limited.
5.	Greg Lim," Beginning Node.js, Express & MongoDB Development, kindle edition,2019 international publishing.
6.	Daniel Perkins, "ReactJS Master React.js with simple steps, guide and instructions" 3rd
7	edition, 2015 SMV publication. Peter Membrey, David Hows, Eelco Plugge, "MongoDB Basics", 2nd edition ,2018
/.	International Publication.
Links: NPTEL/You Tube/Web Link:	
	S. INF IEL/ IOU TUDE/ WED LINK.
https://	youtu.be/QFaFIcGhPoM?list=PLC3y8-rFHvwgg3vaYJgHGnModB54rxOk3
https://	youtu.be/pKd0Rpw7O48
	youtu.be/TIB_eWDSMt4
	youtu.be/QFaFIcGhPoM
·	youtu.be/Kvb0cHWFkdc
·	youtu.be/pQcV5CMara8
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-	youtu.be/AXIAP83CuK4
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<u>mups.//</u>	youtu.oo/151 pexcell1