

Course Code	PAD21D04T	Course Name	MACHINE LEARNING MODEL MANAGEMENT	Course Category	D	Discipline Specific Elective	L	T	P	C
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Pre-requisite Courses	Nil	Co-requisite Courses	Nil	Progressive Courses	Nil
Course Offering Department	Computer Applications	Data Book / Codes/Standards	Nil		

Course Learning Rationale (CLR):	The purpose of learning this course is to,	Learning	Program Learning Outcomes (PLO)
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CLR-1 :	To learn how to Data architects help companies manage, store and secure their data.	1	2	3	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
CLR-2 :	To learn about SQL and NoSQL databases	Level of Thinking (Bloom)	Expected Proficiency (%)	Expected Attainment (%)	Disciplinary Knowledge	Critical Thinking	Problem Solving	Analytical Reasoning	Research Skills	Team Work	Scientific Reasoning	Reflective Thinking	Self-Directed Learning	Multicultural Competence	Ethical Reasoning	Community Engagement	ICT Skills	Leadership Skills	Lifelong Learning
CLR-3 :	To get a clear understanding about Hadoop				L	H	H	H	H	M	-	H	M	H	-	H	-	-	-
CLR-4 :	To get in-depth knowledge of the Big Data framework using Hadoop				L	H	H	H	H	-	-	M	M	L	-	H	-	-	-
CLR-5 :	To analyze large data sets to find trends, correlations or other insights not visible with smaller data sets or traditional processing methods.				L	H	H	H	H	-	-	M	M	L	-	H	-	-	-
CLR-6 :	To observe various customer related patterns and trends.				L	H	H	H	H	-	-	M	M	L	-	H	-	-	-

Course Learning Outcomes (CLO):	At the end of this course, learners will be able to:																		
CLO-1 :	To gather extensive knowledge in Big Data and its architecture	2	85	80	L	H	H	H	H	M	-	H	M	H	-	H	-	-	-
CLO-2 :	To improve and understand ETL process and relevant ETL tools	3	85	80	L	H	H	H	H	-	-	M	M	L	-	H	-	-	-
CLO-3 :	To differentiate between a batch layer for large volumes of data and a speed layer for real time processing of data streams	3	85	80	L	H	H	H	H	-	-	M	M	L	-	H	-	-	-
CLO-4 :	To understand Distributed systems	3	85	80	L	H	H	H	H	-	-	M	M	L	-	H	-	-	-
CLO-5 :	To understand Data modelling techniques	3	85	80	L	H	H	H	H	-	-	M	M	L	-	H	-	-	-
CLO-6 :	To incorporate data from all sources is key to optimizing the insights gained with Big Data.	3	85	80	L	H	H	H	H	-	-	M	M	L	-	H	-	-	-

Duration (hour)	12	12	12	12	12
S-1	SLO-1	ML Model Evaluation and Selection-Introduction	MLModel Management Introduction	Hyperparameter tuning, versioning	Recommendation Algorithms -Recommendation Evaluation and Validation

S-2	SLO-1	Cross-validation	Creating and saving ML models with scikit-learn	Deployment - Hyperparameter tuning with grid search	Recommendation algorithms design	Describing Recommendation Engines
S-3	SLO-1	evaluating estimator performance	Models for Regression	reproducing study	- User-based-Collaborative-Filtering (UBCF) Algorithm	Types of Recommendation Engines
S-4	SLO-1	Computing cross- validated metrics	Getting started with Regression model example	Machine Learning metrics	Collaborative Filtering based Online Recommendation Systems	Comparing the Types of Recommendation Engines
S-5	SLO-1	Cross validation iterators, A note on shuffling, Cross validation and model selection	Classification management	Machine learning model versioning	Item-based Collaborative Filtering (IBCF)	Collecting and Manipulating Data
S-6	SLO-1	Tuning the hyper-parameters of an estimator, Exhaustive GridSearch-Randomized - Parameter Optimization	Building machine Learning pipelines	Machine learning Model versioning with git and DVC	Rule- real-life sequential recommendation systems	Describing Similarity and Neighbourhoods
S-7	SLO-1	Alternatives to brute force parameter search- Metrics and scoring, quantifying the quality of predictions-	Overview-Pipelines	Model management framework	SVD algorithm for recommender systems	creating a Recommendation Engine
S-8	SLO-1	defining model evaluation rules- Classification metrics	Machine Learning pipeline Tools	Studio ml setup	An incremental algorithm- Incremental Appro SVD.	Recommending Another Item
S-9	SLO-1	Multilabel ranking metrics, Regressionmetrics, Clusteringmetrics-Dummyestimators-Model persistence	Machine Learning Pipeline Techniques and tools -Example	Machine learning model creation	Singular Value Decomposition VS Matrix Factorization in Recommender Systems.	Finding Items to Recommend
S-10	SLO-1	&maintainability I imitations- Validation curves	Machine Learning Pipeline implementation	Machine learning model in production	collaborative filtering algorithm	Recommending Items Based on Other Items -
S-11	SLO-1	plotting scores to evaluate models	Iterative Machine Learning model	Deploying model process	Online Recommendation Systems	Evaluating a Recommendation System -
S-12	SLO-1	Validationcurve- Learningcurve	Comparisons of Different Models	Deployed machine learning model in production	Comparisons	Validating Recommendation System

Learning Resources	1 Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow, 2nd Edition, Aurélien Géron, September 2019	1. Hands-On Recommendation Systems with Python: Start Building Powerful and Personalized, Recommendation Engines with Python, Rounak Banik, July 2018
	2. Building Recommendation Engine, By Suresh K Gorakala, December 2016	

Learning Assessment											
Level	Bloom's Level of Thinking	Continuous Learning Assessment (50% weightage)								Final Examination (50% weightage)	
		CLA – 1 (10%)		CLA – 2 (10%)		CLA – 3 (20%)		CLA – 4 (10%)#			
		Theory	Practice	Theory	Practice	Theory	Practice	Theory	Practice	Theory	Practice
Level 1	Remember	40%	-	40%	-	40%	-	40%	-	40%	-
	Understand										
Level 2	Apply	40%	-	40%	-	40%	-	40%	-	40%	-
	Analyze										
Level 3	Evaluate	20%	-	20%	-	20%	-	20%	-	20%	-
	Create										
	Total	100 %		100 %		100 %		100 %		100 %	

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
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