## Name of the Programme: MCA

Course Code: CSA-612

Title of the Course: MLOps

Number of Credits: 4 (2L+2T+0P) Effective from AY: 2022-23

Prerequisites for the course	Familiarity with linear algebra, probability theory, machine learning , familiarity with python.	
Objectives:	This course is aimed at any one who wishes to explore deep learning	
	from scratch. This course offers a practical hands on exploration of	
	deep learning, avoiding mathematical notation, preferring instead to	
	explain quantitative concepts through programming using python	
	API	
Content:	Unit 1. Introduction to MLOps Rise of the Machine Learning Engineer	3 hours
	and MLOps-What Is MLOps?-DevOps and MLOps-An MLOps	
	Hierarchy of Needs-Implementing DevOps-Configuring-Continuous	
	Integration with GitHub Actions-DataOps and Data Engineering-	
	Platform Automation-MLOps	
	Unit 2. MLOps Foundations-Bash and the Linux Command Line-Cloud	3 hours
	Shell Development Environments-Bash Shell and Commands-List Files	
	Run CommandsFiles and Navigation-Input/Output-Configuration-	
	Writing a Script-Cloud Computing Foundations and Building Blocks-	
	Getting Started with Cloud Computing- minimalistic python revision-	
	Descriptive Statistics and Normal Distributions-Optimization-Machine	
	Learning Key Concepts-Doing Data Science-Build an MLOps Pipeline	
	from Zero	
	Unit 3. MLOps for Containers and Edge Devices Containers-Container	3 hours
	Runtime-Creating a Container Running a Container-Best Practices-	
	Serving a Trained Model Over HTTP-Edge Devices-Coral Azure	
	Percept-TFHub-Porting Over Non-TPU Models-Containers for	
	Managed ML Systems-Containers in Monetizing MLOps-Build Once,	
	Run Many MLOps Workflow	
	Unit 4. Continuous Delivery for Machine Learning Models-Packaging	<mark>3 hours</mark>
	for ML Models-Infrastructure as Code for Continuous Delivery of ML	
	Models-Using Cloud Pipelines-Controlled Rollout of Models-Testing	
	Techniques for Model Deployment	
	Unit 5. AutoML and KaizenML-AutoML-MLOps Industrial Revolution-	3 hours
	Kaizen Versus KaizenML-Feature Stores-Apple's Ecosystem-Apple's	
	AutoML: Create ML-Apple's Core ML Tools orGoogle's AutoML and	
	Edge Computer Vision or Azure's AutoMLor AWS AutoML-Open	
	Source AutoML Solutions-Ludwig-FLAML-Model Explainability	
	Unit 6. Monitoring and Logging-Observability for Cloud MLOps-	<mark>3 hours</mark>
	Introduction to Logging-Logging in Python-Modifying Log Levels-	
	Logging Different Applications-Monitoring and Observability-Basics of	
	Model Monitoring-Monitoring Drift with AWS SageMaker-Monitoring	
	Drift with Azure ML	
	Unit 7. MLOps for AWS-Introduction to AWS-Getting Started with	<mark>3 hours</mark>
	AWS Services-MLOps on AWS-MLOps Cookbook on AWS-CLI Tools-	
	Flask Microservice-AWS Lambda Recipes-AWS Lambda-SAM Local-	
	AWS Lambda-SAM Containerized Deploy-Applying AWS Machine	
	Learning to the Real World	
	Unit 8. Machine Learning Interoperability-Why Interoperability Is	3 hours
	Critical-ONNX: Open Neural Network Exchange-ONNX Model Zoo-	
	Convert PyTorch into ONNX -Convert TensorFlow into ONNX-Deploy	
	ONNX to Azure-Apple Core ML-Edge Integration	
	Unit 9: Building MLOps Command Line Tools and Microservices-	3 hours

	Python Packaging-The Requirements File-Command Line Tools- Creating a Dataset Linter Modularizing a Command Line Tool- Microservices-Creating a Serverless Function-Authenticating to Cloud Functions-Building a Cloud-Based CLI-Machine Learning CLI Workflows <b>Unit 10.</b> Machine Learning Engineering and MLOps Case	3 hours
	StudiesUnlikely Benefits of Ignorance in Building Machine Learning Models-MLOps Projects at Sqor Sports Social Network-Mechanical Turk Data Labeling-Influencer Rank-Athlete intelligence (AI product)- The perfect techniques versus the real world-critical challenges in MLops- Ethical and unintended consequences-lack of operational	
	excellences- focus on prediction accuracy vs the big picture	40 * 2 20
	Assignments to be done during Tutorial Slots:	10 * 3 = 30
	1. Perfect Project Structure – Cookiecutter & readme.so	hours
	<ol> <li>Speed Exploratory Data Analysis to Minutes – Pandas Profiling, SweetViz</li> </ol>	
	3. Track Data Science Projects with CI, CD, CT, CM –Data Version	
	$\frac{1}{2} = \frac{1}{2} = \frac{1}$	
	4. EXPlainable AI / XAI – SHAP, LIVIE, SHAPASH	
	5. Deploy Mil Projects III minutes - Docker, FastAPI	
	<ol> <li>End to End Machine Learning – Milliow</li> <li>Ruilding Production Roady MI, Dipolinos – Model Registry, Feature</li> </ol>	
	Store (Foast ButterFlow)	
	Store (reast, butter row) 8 Big Data using Bython instead of BySpark - DASK	
	<ul> <li>Big Data using Fython, instead of Fyspark – DASK</li> <li>Build a Chathot and Doploy it (open source)</li> </ul>	
	<ol> <li>Build a Chalbot and Deploy it (Open-source)</li> <li>Ease Framowork implementation – Anacho OpenWhick, OpenEase</li> </ol>	
Podogogy:	10. Faas Framework implementation – Apache Openwinsk, OpenFaas	
Pedagogy:		
References/	Main Reading :-	
Readings	Practical MLops – Noah Gift and AlfredoDeza	
	Introduction to MLOps – Noah Gift and AlfredoDeza	
	1 Machine Learning Engineering	
	By Andriy Burkov	
	2.ML Ops: Operationalizing Data Science By David Sweenor, Dev	
	Kannabiran, Thomas Hill, Steven Hillion, Dan Rope and Michael	
	O Connell-O Relliy	
	3. Building Machine Learning Pipelines by Hannes Hapke, Catherine	
	Nelson A Practical MI One by Neeb Cift, Alfrede Deze, O'Deilly	
	4. Flactical Wildes by North Gill, Allfedd Deza. O Kelliy	
	5. Introducing WICOps by Wark Heveli & Dataiku Team 6. Reginning MICops with MI Flow: Deploy Models in AWS SegeMaker	
	Google Cloud, and Microsoft Azura	
	By Sridhar Alla, Suman Kalvan Adari, O'Reilly	
Course	Students will be able to _	
Outcomes	1 handle deployment challenges in ML projects	
	<ol> <li>Advelop technical competence to deploy the ML projects</li> </ol>	
	2. develop teennear competence to deploy the ML projects.	