

Day - 1

• Overview of Databricks and Machine Learning on Databricks

- Introduction to Databricks: Platform overview, key components, and features
- Machine learning on Databricks: Databricks runtime for ML, Spark MLlib, and MLflow
- Setting up the environment for machine learning workflows
- Hands-on Lab: Set up a Databricks workspace for ML experiments.

• Deep Dive into Supervised Learning

- Advanced Regression Techniques: Lasso, Ridge, ElasticNet, and Polynomial Regression
- Classification Algorithms: Logistic Regression, Support Vector Machines (SVM), and K-Nearest Neighbors (KNN)
- Ensemble Learning: Random Forests, Gradient Boosting Machines (GBM), and XGBoost
- Hands-on Lab: Building and tuning classification and regression models with XGBoost.

• Feature Engineering for Machine Learning

- Feature selection techniques: Univariate selection, recursive feature elimination, and principal component analysis (PCA)
- Feature transformation: Normalization, Standardization, Log Transform, and Polynomial Features
- Handling categorical data: One-Hot Encoding, Label Encoding
- Hands-on Lab: Implementing feature engineering on a real dataset (e.g., customer segmentation).

• Hyperparameter Tuning & Model Optimization

- Introduction to Hyperparameter Tuning: GridSearchCV, RandomSearchCV, and Bayesian optimization
- Tuning XGBoost and Random Forests: Best practices for performance optimization
- Techniques for preventing overfitting: Cross-validation, Regularization
- Hands-on Lab: Hyperparameter tuning of classification models (e.g., XGBoost, Random Forest)

- **Clustering and Dimensionality Reduction**

- Introduction to Clustering: K-Means, DBSCAN, Agglomerative Clustering
- Dimensionality Reduction Techniques: Principal Component Analysis (PCA) and t-SNE
- Model evaluation in clustering: Silhouette Score and Inertia
- Hands-on Lab: Implementing K-Means and PCA for dimensionality reduction in a dataset.

- **Advanced Deep Learning Techniques**

- Introduction to Deep Learning: Neural Networks basics
- Convolutional Neural Networks (CNNs): Use cases in image classification
- Recurrent Neural Networks (RNNs) and Long Short-Term Memory (LSTM): Use cases in time-series analysis
- Hands-on Lab: Implementing a basic CNN and RNN for a classification task using Keras in Databricks.

- **Model Evaluation and Interpretation**

- Advanced Model Evaluation Metrics: Precision-Recall curve, ROC-AUC, Log-Loss
- Model Interpretability: SHAP values, LIME for model explainability
- Feature Importance: Understanding and visualizing feature contribution to predictions
- Hands-on Lab: Evaluating and interpreting a classification model using SHAP and LIME.

- **Deploying Machine Learning Models on Databricks**

- Introduction to Model Deployment: Benefits of model serving and deployment at scale
- Deploying models using MLflow: Model packaging and serving REST APIs
- Scaling models using Databricks Model Serving
- Hands-on Lab: Deploying a trained model as a REST API using MLflow and Databricks Model Serving

- **MLflow and Managing Machine Learning Lifecycles**

- Introduction to MLflow: Managing the complete ML lifecycle—tracking, model versioning, and artifact storage
- Experiment Tracking: Logging parameters, metrics, and artifacts
- Model Registry: Registering, versioning, and managing models in MLflow
- Hands-on Lab: Implementing MLflow to track machine learning experiments and register models.

- **AutoML with Databricks**

- Introduction to AutoML: What is AutoML, and how it accelerates model development
- AutoML in Databricks: AutoML for regression and classification tasks
- Hands-on Lab: Running AutoML for a classification problem in Databricks

- **Building End-to-End ML Pipelines in Databricks**

- Overview of an end-to-end ML pipeline: Data preprocessing, model training, hyperparameter tuning, and evaluation
- Automating ML pipelines with Databricks Workflows
- Integrating pipelines with Databricks Jobs and triggering automated workflows
- Hands-on Lab: Building an end-to-end machine learning pipeline for batch processing data.

- **Exam Preparation and Review**

- Exam structure: Review of the certification exam topics and question types
- Time management strategies for the exam
- Hands-on mock exam: 45 minutes of simulated exam questions
- Answer key review and discussions
- Tips for continued learning and certification readiness.