## SUMMARY OF THE PROJECT

## METHODOLOGY AND WORKING

- Database Schema Design (MySQL)
- Tables to Create:
  - Models Table: Stores model-specific details (ID, name, type, framework, etc.).
  - Training Table: Stores training dataset, hyperparameters, duration, etc.
  - Metrics Table: Stores model performance metrics (accuracy, loss, AUC, etc.).
  - Versioning Table: Tracks versions of the model (each time a model is retrained, a new version is logged).
  - Deployment Table: Logs deployment details (date, environment, endpoints, etc.).

- ► Inserting Metadata
- Before Training: Insert model and hyperparameter configurations into the database.
- After Training: Insert performance metrics, training duration, dataset used, and the trained model file path or location.
- On Model Versioning: Update version history and track changes in the model and parameters.
- .Querying Metadata
- Accessing Model Details: Retrieve models based on ID, framework, or type.
- Performance Monitoring: Retrieve and compare historical performance metrics.
- Version History: Track which versions performed better or are deployed in specific environments.
- **Deployment Tracking:** Monitor the production status and rollback history.

- ▶ Best Practices for Metadata Management
- Standardization: Define a standardized format for logging metadata.
- Automation: Automate the logging process during training and deployment.
- Version Control: Ensure that each model version has associated metadata for reproducibility.
- Security: Secure access to metadata for audit purposes and compliance.
- Integration with ML Pipelines
- Logging: Integrate metadata management with your ML pipeline (e.g., during data preprocessing, model training, evaluation).
- Traceability: Ensure that every experiment is logged with all relevant metadata to support debugging, auditing, and collaboration.