



## **Project Initialization and Planning Phase**

| Date          | 26 June 2025   |
|---------------|--|
| Team ID       | NONE   |
| Project Title | Employee Performance Prediction using Machine Learning |

## **Project Proposal (Proposed Solution) report**

The proposal report aims to revolutionize employee performance evaluation using machine learning, enabling data-driven, proactive workforce management. It addresses the limitations of subjective and delayed performance reviews by introducing a predictive system that forecasts productivity based on real work metrics. Key features include a machine learning-based prediction model and a user-friendly Flask web interface for real-time insights

| Project Overview  |  |
|-------------------|--|
| Objective         | The primary objective is to develop a machine learning-based system that predicts employee productivity using historical work-related data, enabling organizations to identify performance trends early and support employees proactively.   |
| Scope             | The project covers the complete machine learning pipeline — from data collection, preprocessing, and exploratory data analysis (EDA), to model training (Linear Regression, Random Forest, XGBoost), evaluation, and deployment. A Flask-based web application is developed to allow real-time predictions, making the solution accessible and user-friendly for management and HR teams.  |
| Problem Statement |  |
| Description       | The current methods of evaluating employee performance are often subjective, relying on periodic reviews and qualitative feedback rather than real-time, data-driven insights. This leads to delayed identification of underperformance, inefficient resource allocation, and missed opportunities for timely employee interventions. As a result, organizations may struggle to optimize workforce efficiency and retain high-performing employees. |
| Impact            | Solving this issue enables organizations to proactively identify performance trends, support employees through targeted training, and make informed decisions about team management. This leads to improved productivity, higher employee engagement, reduced turnover, and better overall operational efficiency.   |
| Proposed Solution |  |
| Approach          | The proposed solution leverages machine learning to predict employee productivity based on historical work data. The approach involves collecting and preprocessing the Garment Employee   |





|              | Productivity dataset, performing exploratory data analysis (EDA), training and evaluating multiple regression models (Linear Regression, Random Forest, XGBoost), and selecting the best-performing model (Random Forest) for deployment. The final model is integrated into a Flask-based web application, enabling real-time predictions through a user-friendly interface.   |
|--------------|---|
| Key Features | -Machine Learning-Powered Prediction: Uses Random Forest algorithm to predict employee productivity with an R² score of ~0.46  Real-Time Web Interface: A Flask-based web app allows managers to input employee data and receive instant performance predictions.  Data-Driven Decision Making: Helps HR and team leads identify performance trends early, enabling proactive interventions like training or resource reallocation.  End-to-End Pipeline: Complete workflow from data preprocessing to model deployment, showcasing a full-stack machine learning implementation. |

## Resource Requirements

| Resource Type           | Description                             | Specification/Allocation                              |
|-------------------------|---|---|
| Hardware                |   |   |
| Computing Resources     | CPU/GPU specifications, number of cores | CPU (No GPU required — model trained on CPU)          |
| Memory                  | RAM specifications                      | 8 GB  |
| Storage                 | Disk space for data, models, and logs   | 1 TB SSD  |
| Software                | *                                       | -10   |
| Frameworks              | Python frameworks                       | Flask   |
| Libraries               | Additional libraries                    | scikit-learn, pandas, numpy,<br>matplotlib, seaborn   |
| Development Environment | IDE                                     | Jupyter Notebook, pycharm                             |
| Data                    | ,                                       |   |
| Data                    | Source, size, format                    | Kaggle dataset, 1197 rows<br>× 15 columns, CSV format |