





Phase-2 Submission Template

Student Name: [Enter Your Name]

Register Number: [Enter Your Register Number]

Institution: [Insert College Name]

Department: [Enter Your Department Name]

Date of Submission: [Insert Date]

Github Repository Link: [Update the project source code to

your Github Repository]

1. Problem Statement

[Clearly articulate the real-world problem being solved, refined from Phase-1

- Revisit and refine the problem based on additional understanding of the dataset.
- Clearly define the type of problem (classification, regression, clustering, etc.).
- Explain why solving this problem matters (impact, relevance, or application area).]

2. Project Objectives

[Update the project goals now that you're entering practical implementation.

• Define the key technical objectives







- Specify what the model aims to achieve (e.g., accuracy, interpretability, real-world applicability).
- Mention if the goal has changed or evolved after data exploration.]

3. Flowchart of the Project Workflow

[Visually represent the entire workflow from start to finish.]

4. Data Description

[Provide a short recap of the dataset and its source.

- Dataset name and origin (e.g., Kaggle, UCI, open APIs).
- Type of data: structured, unstructured, image, text, time-series, etc.
- Number of records and features.
- Static or dynamic dataset.
- Target variable (if supervised learning).]

5. Data Preprocessing

[Perform and document data cleaning and preparation.

- Handle missing values (removal, imputation, etc.).
- Remove or justify duplicate records.







- Detect and treat outliers.
- Convert data types and ensure consistency.
- Encode categorical variables (label encoding, one-hot encoding).
- Normalize or standardize features where required.
- Document and explain each transformation step clearly in code and markdown.]

6. Exploratory Data Analysis (EDA)

[Perform detailed statistical and visual exploration of the data.

- Univariate Analysis:
 - Distribution of features using histograms, boxplots, countplots, etc.
- Bivariate/Multivariate Analysis:
 - o Correlation matrix, pairplots, scatterplots, grouped bar plots, etc.
 - Analysis of relationship between features and the target variable.
- Insights Summary:
 - $\circ \ \ \textit{Highlight patterns, trends, and interesting observations}.$
 - Mention which features may influence the model and why.]







7. Feature Engineering

[Enhance or transform data to improve model performance.

- Create new features based on domain knowledge or EDA insights.
- Combine or split columns (e.g., extracting date parts).
- Use techniques like binning, polynomial features, ratios, etc.
- Apply dimensionality reduction (optional, e.g., PCA).
- Justify each feature added or removed.]

8. Model Building

[Build and compare multiple models to solve the defined problem.

- Select and implement at least 2 machine learning models.
 - o E.g., Logistic Regression, Decision Tree, Random Forest, KNN, etc.
- Justify why these models were selected (based on problem type and data).
- Split data into training and testing sets (with stratification if needed).
- Train models and evaluate initial performance using appropriate metrics.
 - For classification: accuracy, precision, recall, F1-score.
 - For regression: MAE, RMSE, R² score].







9. Visualization of Results & Model Insights

[Use plots and charts to explain model behavior.

- Confusion matrix, ROC curve, feature importance plot, residual plots, etc.
- Include visual comparisons of model performance.
- Interpret top features influencing the outcome.
- Clearly explain what each plot shows and how it supports conclusions.]

10. Tools and Technologies Used

[Mention all tools used in this phase of the project.

- Programming Language: Python or R.
- IDE/Notebook: Google Colab, Jupyter Notebook, VS Code, etc.
- Libraries: pandas, numpy, seaborn, matplotlib, scikit-learn, XGBoost, etc.
- Visualization Tools: Plotly, Tableau, Power BI.]

11. Team Members and Contributions

[List names and responsibilities.

- Clearly mention who worked on:
 - Data cleaning







- \circ EDA
- o Feature engineering
- o Model development
- Documentation and reporting]