# Machine Learning Project Plan

This document outlines the steps for the Machine Learning project plan. Each step is listed in a table with additional columns for Points, Self Grade, and Grader Grade. On UB Learns, upload a completed version of this sheet. The project is due 12/18/2024 @11:59 PM. No late submissions. I will ignore all emails about late submissions. **You MUST keep your apps running for one week, so you can turn it off 12/25/2024 @11:59PM.**

URL to your Jupyter Book: <https://someshbhandarkar.github.io/Loan-Classification-Project/intro.html>

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| --- | --- | --- | --- |
| **Step** | **Points** | **Self Grade** | **Grader Grade** |
| Create a normalized database (3NF). | 20 | 20 |  |
| Write SQL join statement to fetch data from the database and into Pandas DataFrame. | 5 | 5 |  |
| Explore the data to determine if you need to stratify it by some attribute when doing train/test split. Perform the train/test split. | 10 | 10 |  |
| Explore the data using yprofile and correlation matrix. Make observations about features, distributions, capped values, and missing values. Create a list of data cleanup tasks. | 10 | 10 |  |
| Experiment #1: Create a pipeline for preprocessing (StandardScaler, MinMaxScaler, LogTransformation, OneHotEncoding) and Logistic Regression. Log F1-score/(TP,TN,FN,FP) in MLFlow on DagsHub. – Cross validation 3/10 folds. Results—mean/std of CV results and results on the whole training data – add in parameter hyper tuning | 20 | 20 |  |
| Experiment #2: Create a pipeline for preprocessing and use LogisticRegression, RidgeClassifier, RandomForestClassifier, and XGBClassifier. Log results in MLFlow on DagsHub. | 10 | 10 |  |
| Experiment #3: Perform feature engineering and attribute combination. Log results in MLFlow. | 10 | 10 |  |
| Experiment #4: Perform feature selection using Correlation Threshold, Feature Importance, and Variance Threshold. Log results in MLFlow. | 10 | 10 |  |
| Experiment #5: Use PCA for dimensionality reduction on all the features. Create a scree plot to show which components will be selected for classification. Log results in MLFlow. | 10 | 10 |  |
| Experiment #6: Design and execute a custom experiment. Log results in MLFlow. | 10 | 10 |  |
| Experiment #7: Design and execute another custom experiment. Log results in MLFlow. | 10 | 10 |  |
| Create meaningful F1-score plots to compare experiments and determine the best model. | 10 | 10 |  |
| Save the final model using joblib.  Create a FastAPI application to serve the model.  Containerize the FastAPI application using Docker and push to Docker Hub.  Deploy the containerized API to a cloud platform. | 20 | 20 |  |
| Create a Streamlit app to interact with the deployed model for real-time classification. | 15 | 15 |  |
| Create a 12-15 minute video explaining the project. Walk through the Jupyter Notebook explaining your project. Your presentation should be coherent, organized, and you must speak clearly. **-5 points** for not having video with face. Use Zoom to record your video. Maybe you can record to cloud to link your video or you can upload to UB Box. | 30 | 30 |  |
| Create a JupyterBook website with your resume, embedded video, and final code. Include links to MLFlow/DagsHub experiments, Docker Hub container, deployed model, and Streamlit app. | 20 | 20 |  |
| **Grade** | 220 | 220 |  |

Use the following section to inform your grader why you missed certain things or how they were not applicable to your project so your grades can be adjusted.

Observations made during the projects:

1. Need to address the imbalance in the target variable –> the personal loan column is showed as imbalanced, but in real life it is okay because the no. of people having loan can be less than that of the people having loans taken or vice-versa. But for dataset to predict properly we need to use a parameter “stratify=” to balance the imbalanced data. You can see the imbalance in Y-Profile for ‘Personal-Loan’ column. I have handled that properly

2. There were missing values in the start, and they have been filled using raw python.

apart from all this there were no other observations, the data was all good for the rest of the procedures to be followed.

**\*Note: I have used Google cloud run for deployment purposes - Access Transparency is not available for projects that are not part of an organization. To enable Access Transparency for a single project, please contact sales or support. \***

This means that the project that I have created on google cloud cannot be shared / kept public for the above reasons. But my model has been deployed. You can confirm it using the link which say ‘app.run’ in the last.

Check the below photo to know that project has been created on google cloud!

A screenshot of a computer

Description automatically generated