

Fuse Machines
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A Project Proposal on:
Wine Quality Classification

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Chapter 1

Introduction

1.1 Problem Statement

The dataset is related to red variants of the Portuguese "Vinho Verde" wine. The dataset describes the amount of various chemicals present in wine and their effect on its quality. The datasets can be viewed as classification or regression tasks. The classes are ordered and not balanced (e.g. there are much more normal wines than excellent or poor ones). Task is to predict the quality of wine using the given data.

A simple yet challenging project, to anticipate the quality of wine.

The complexity arises due to the fact that the dataset has fewer samples, & is highly imbalanced.

Data Source : <https://archive.ics.uci.edu/ml/datasets/wine+quality>

1.2 About Dataset

The dataset contains the following columns:

1. fixed acidity
2. volatile acidity
3. citric acid
4. residual sugar
5. chlorides
6. free sulfur dioxide
7. total sulfur dioxide
8. density
9. pH
10. sulphates

11. alcohol

12. quality (Targe Variable) : ranges from 0 to 10

The data information is as follow:

```
Data columns (total 13 columns):
#      Column      Non-Null Count  Dtype
---  -
0      fixed acidity    1143 non-null   float64
1      volatile acidity  1143 non-null   float64
2      citric acid       1143 non-null   float64
3      residual sugar    1143 non-null   float64
4      chlorides         1143 non-null   float64
5      free sulfur dioxide 1143 non-null   float64
6      total sulfur dioxide 1143 non-null   float64
7      density           1143 non-null   float64
8      pH               1143 non-null   float64
9      sulphates         1143 non-null   float64
10     alcohol           1143 non-null   float64
11     quality            1143 non-null   int64
12     Id                1143 non-null   int64
dtypes: float64(11), int64(2)
memory usage: 116.2 KB
```

Figure 1.1: Data Information

Description of each column is shown below:

| stats | fixed acidity | volatile acidity | citric acid | residual sugar | chlorides | free sulfur dioxide |
|-------|---------------|------------------|-------------|----------------|-----------|---------------------|
| mean | 8.311 | 0.5313 | 0.264 | 2.532 | 0.0869 | 15.615 |
| std | 1.747 | 0.179 | 0.196 | 1.355 | 0.0477 | 10.250 |
| min | 4.600 | 0.120000 | 0.000 | 0.900 | 0.0120 | 1.00 |
| 25% | 7.100 | 0.39250 | 0.0900 | 1.900 | 0.070 | 7.000 |
| 50% | 7.900 | 0.5200 | 0.2500 | 2.200 | 0.0790 | 13.000 |
| 75% | 9.100 | 0.6400 | 0.4200 | 2.6000 | 0.090 | 21.000 |
| max | 15.900 | 1.580 | 1.00 | 15.50 | 0.6110 | 68.00 |

Table 1.1: Data Description 1

| stats | total sulfur dioxide | density | pH | sulphates | alcohol |
|-------|----------------------|----------|----------|-----------|-----------|
| mean | 45.914698 | 0.996730 | 3.311015 | 0.657708 | 10.442111 |
| std | 32.782130 | 0.001925 | 0.156664 | 0.170399 | 1.082196 |
| min | 6.000000 | 0.990070 | 2.740000 | 0.330000 | 8.400000 |
| 25% | 21.000000 | 0.995570 | 3.205000 | 0.550000 | 9.500000 |
| 50% | 37.000000 | 0.996680 | 3.310000 | 0.620000 | 10.200000 |
| 75% | 61.000000 | 0.997845 | 3.400000 | 0.730000 | 11.100000 |
| max | 289.000000 | 1.003690 | 4.010000 | 2.000000 | 14.900000 |

Table 1.2: Data Description 2

1.3 Dataset Statistics

The count plot of the whole dataset on the basis of quality of wine is shown below.

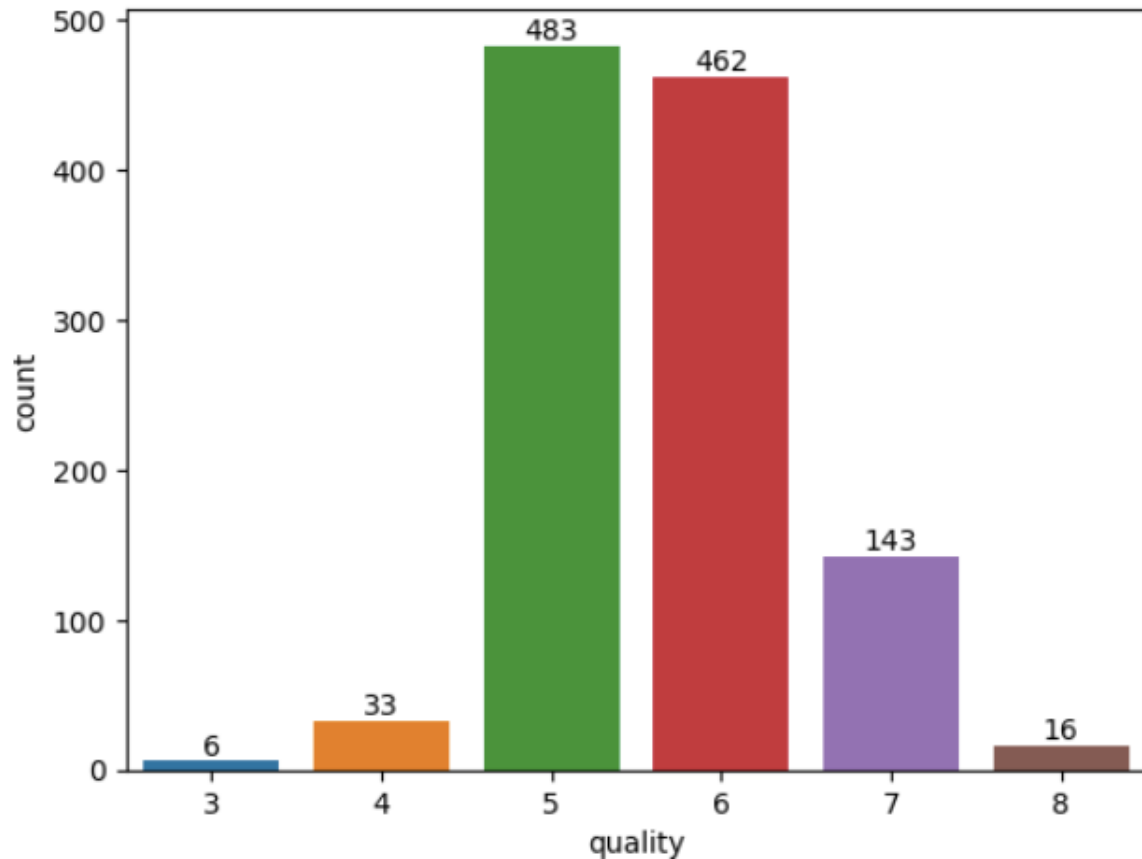


Figure 1.2: Quality Count plot

We can see from above countplot, that the data is highly imbalanced and quality of wine(0, 1, 2, 9, 10) are not present in the dataset.

In order to deal with this imbalanced, at first we are going to merge the quality labels (3 and 4) and (7 and 8) as shown below.

```
def wineQualityTransform(quality):  
    wine_quality_transformation = {3: 0, 4: 0, 5:1, 6:2, 7: 3, 8: 3}  
    return wine_quality_transformation[quality]  
  
data_df["quality"] = data_df["quality"].apply(wineQualityTransform)  
print(data_df.quality.value_counts())
```

Figure 1.3: Code for transformation of quality attribute

The resulting count plot is as shown below:

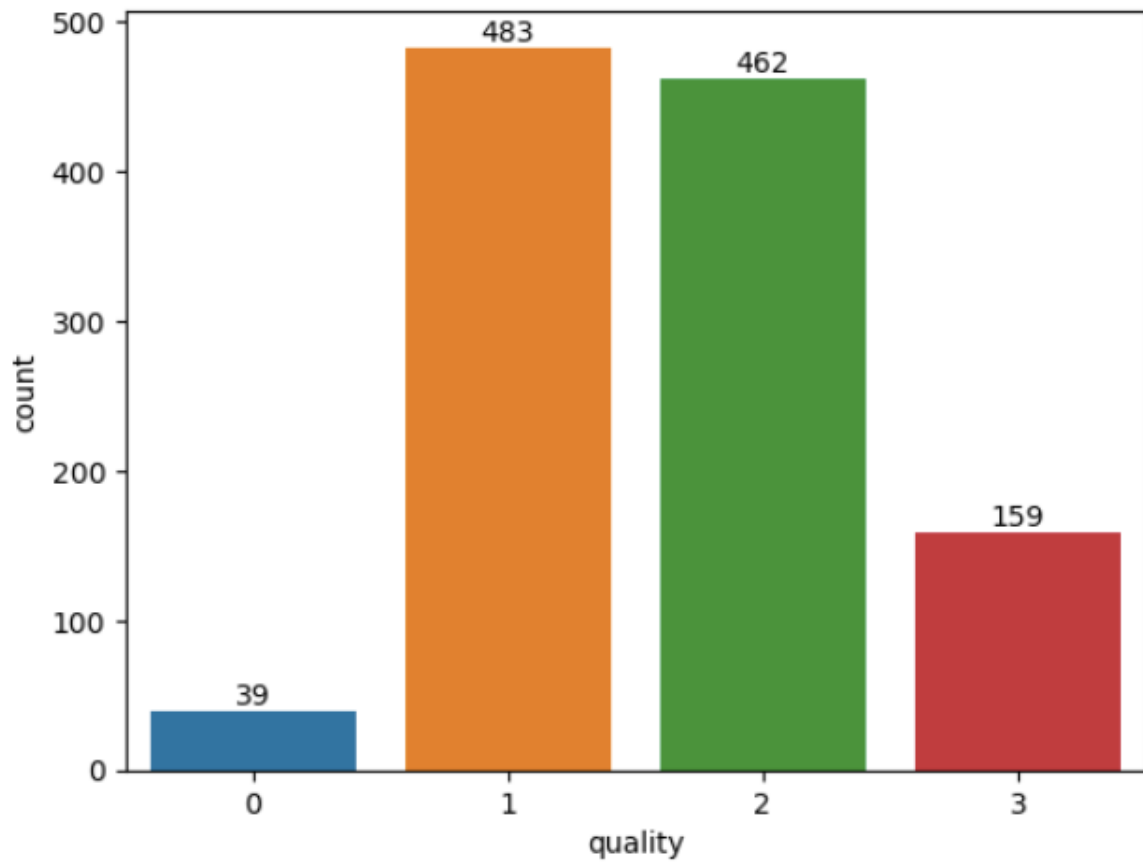


Figure 1.4: Quality Count plot after transformation

1.3.1 Train, validation and test dataset

We are going to use 80%, 20% and 20% of the dataset as training, validation and test data. The countplot are as shown below:

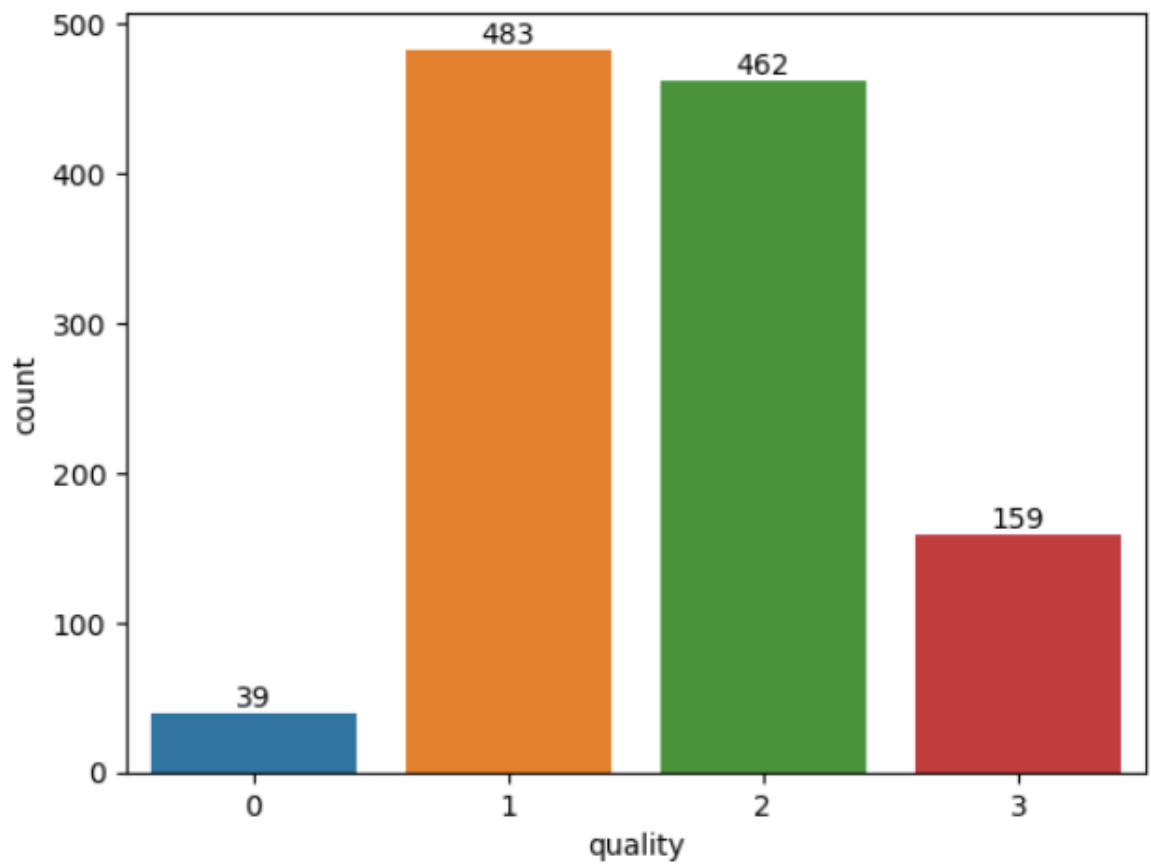


Figure 1.5: Train Quality Count plot after transformation

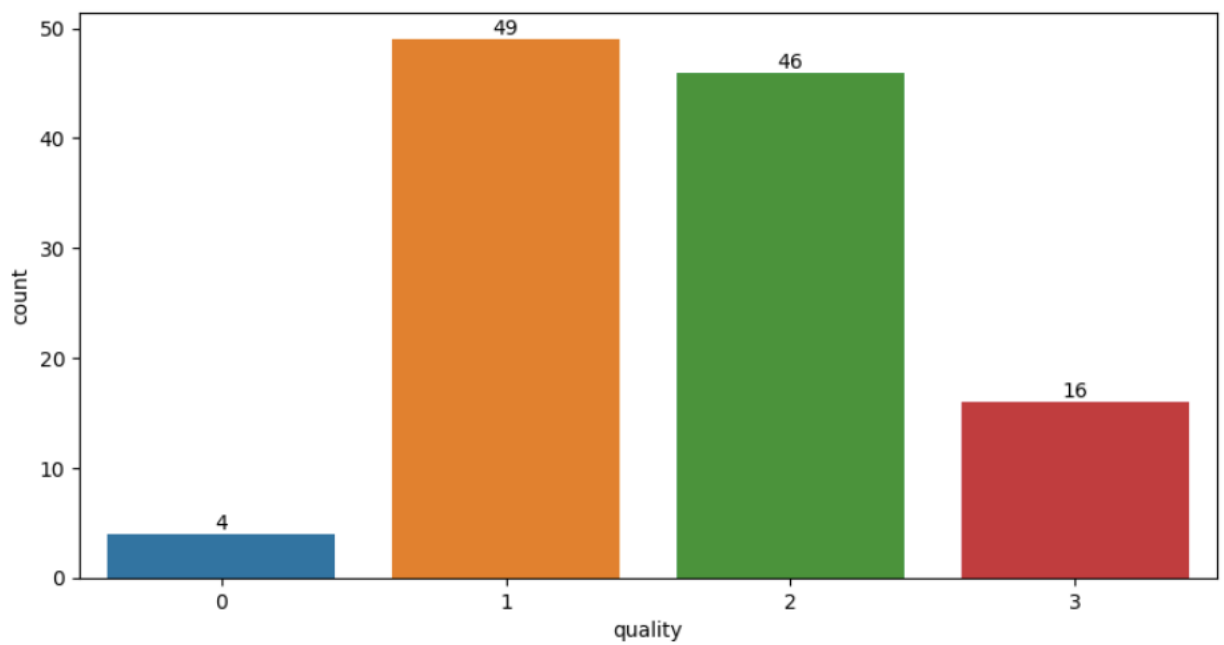


Figure 1.6: Validation Quality Count plot after transformation

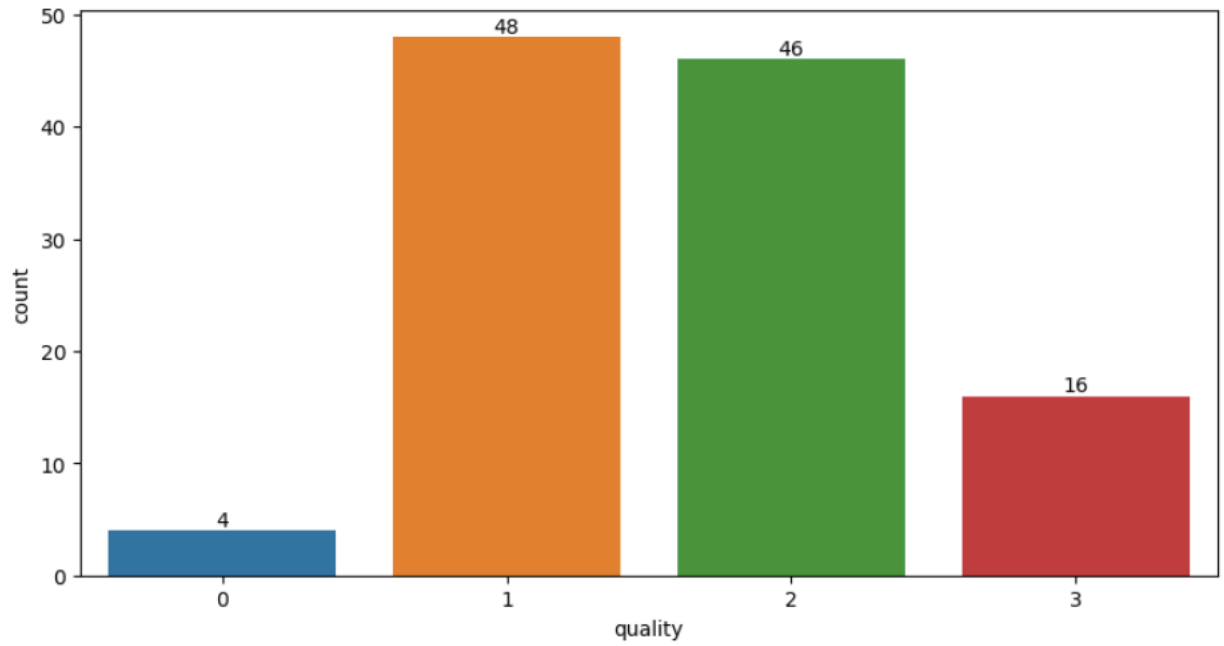


Figure 1.7: Test Quality Count plot after transformation

1.4 Expected Outcome

We expect the output to be the list of probabilities of belonging to a class[0, 1, 2 or 3]. Among these, the class with higher probability is our predicted class.

Chapter 2

Project Management

2.1 Agile Development Methodology

Teams use the agile development methodology to minimize risk (such as bugs, cost overruns, and changing requirements) when adding new functionality. In all agile methods, teams develop the software in iterations that contain mini-increments of the new functionality. There are many different forms of the agile development method, including scrum, crystal, extreme programming (XP), and feature-driven development (FDD).

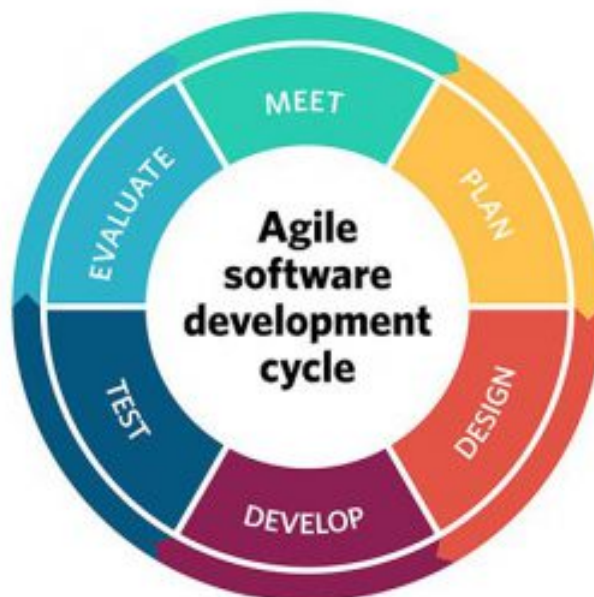


Figure 2.1: Agile Development Methodology

2.2 Trello

We are going to use Trello's default agile board for task management.

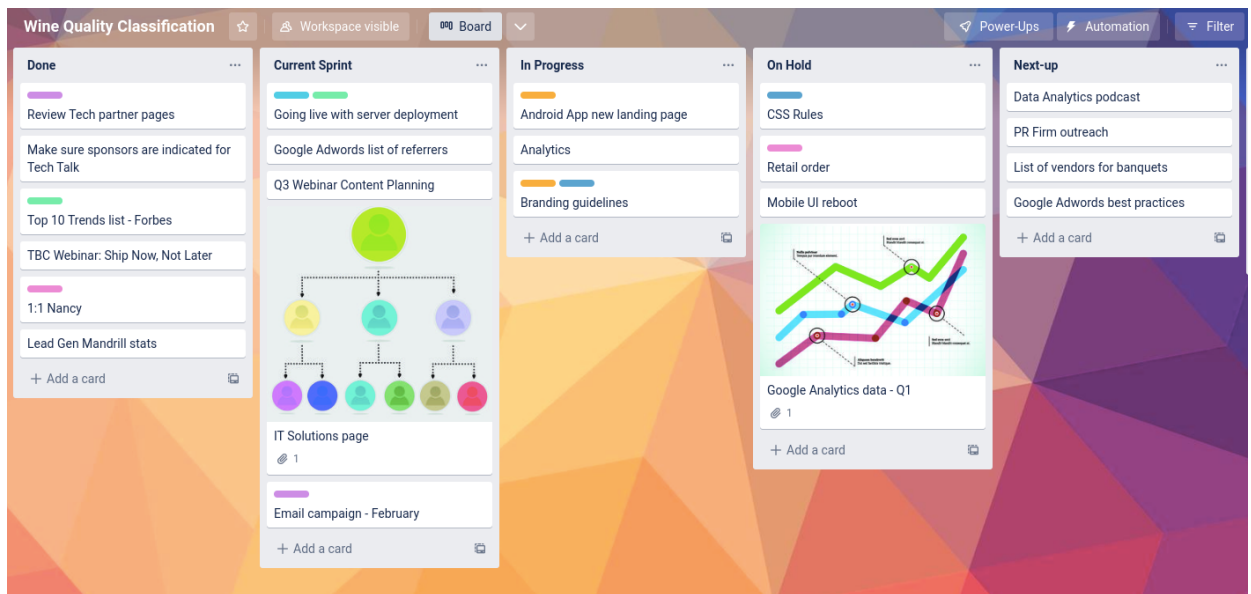


Figure 2.2: Trello

2.3 Discord

We are going to use Discord server for effective communication and team meetings.

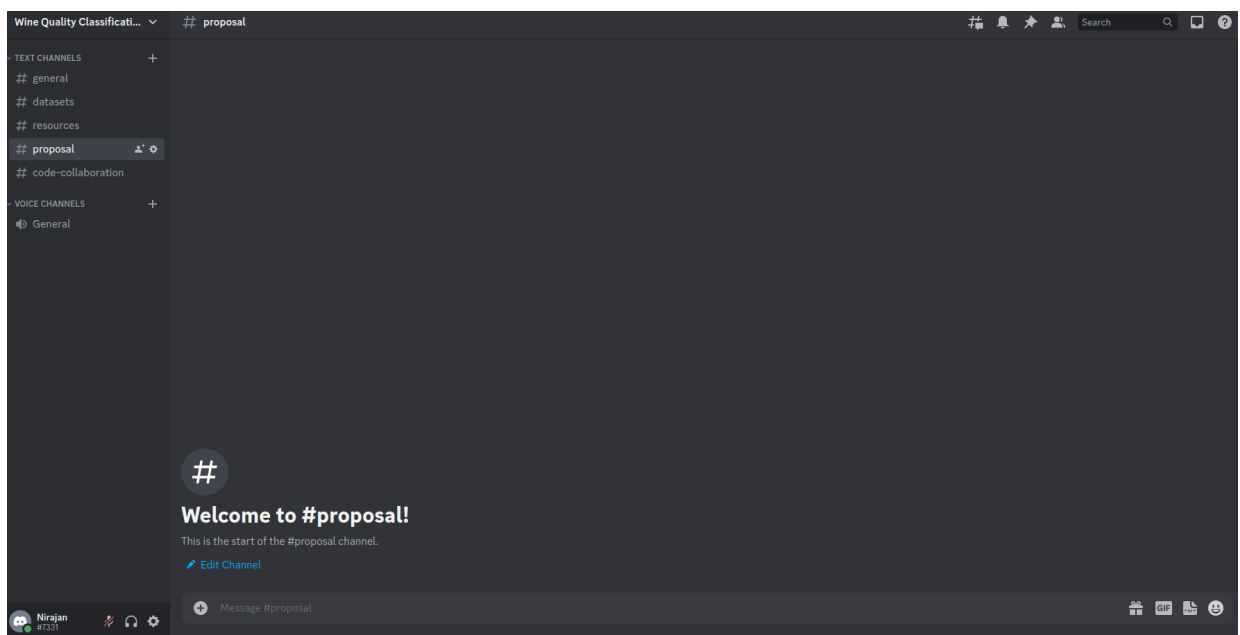


Figure 2.3: Discord

2.4 Tensorboard

Tensorboard will be used for experiment tracking.

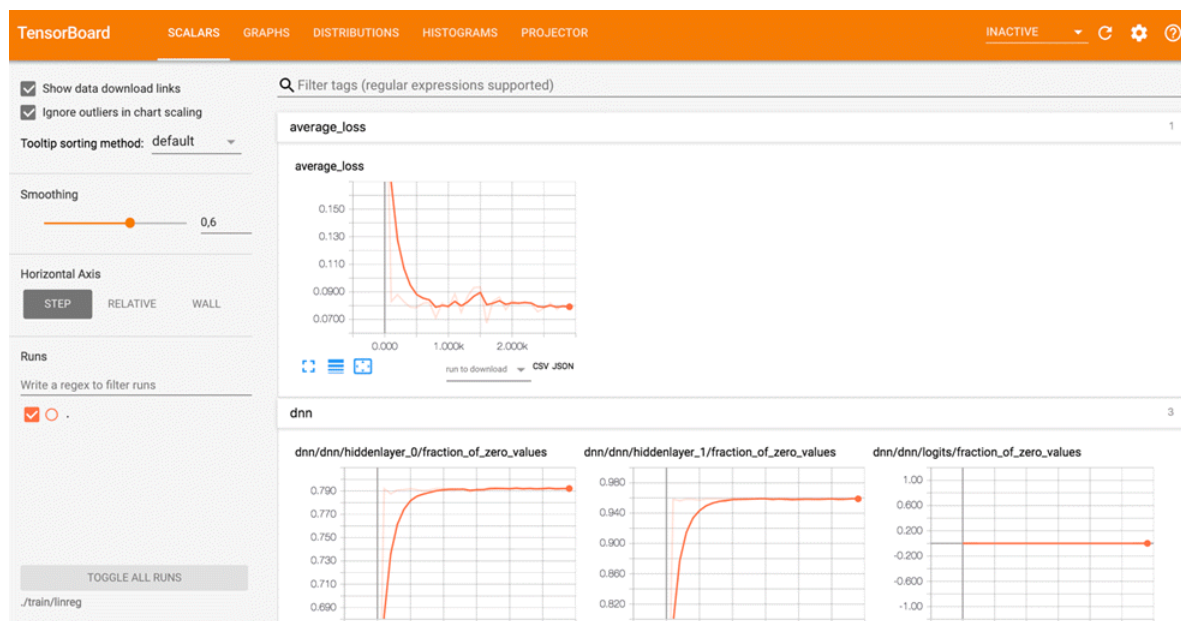


Figure 2.4: Tensorboard

Chapter 3

Model Development and Evaluation

3.1 Model Development

Our data is highly imbalanced as discussed in chapter 1. So, we are going to use different methods like undersampling, oversampling, smote and class weights to handle the imbalanced data.

Since this is a classification task, we can use various machine learning models (e.g : Logistic regression, Decision Trees, Ensemble methods, etc.)

3.2 Model Evaluation

We have highly imbalanced data, so we are going to use macro F1-score as the primary model evaluation metrics. Different other evaluation like roc, auc-score, precision and recall, etc. will also be studied.