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Project Step II

Spotify genre classification Dataset

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| --- | --- | --- | --- |
| Project # | Project Description | Date Submitted | Project Priority |
| 1 |  |  |  |

## Step 1. Project Deliverables

Please list *all project deliverables* listed in the Project Charter and, if necessary, elaborate on them. *Do not list dates*. Add more rows as necessary.

|  |  |
| --- | --- |
| **Deliverable ID#** | **Description** |
| 1 | Project Charter |
| 2 | Ensemble trained model with preliminary results of the training and testing. |
| 3 | Flask and Heroku application. (customer will choose whether to deploy it locally or on the cloud) |
| 4 | Final report of the project in addition to the cloud (heroku) deployed link. |

## Step 2. List of Project Tasks

Please list ***all project tasks*** to be completed, based on the “Deliverables” specified in the Project Charter. *Do not list dates*. Add more rows as necessary. Optional: you may substitute a work breakdown structure (WBS) or mind-map in lieu of Step 2. Please attach WBS or mind-map to document.

|  |  |  |  |
| --- | --- | --- | --- |
| **Task ID#** | **Task to be completed** | **Delivery Date** | **For Deliverable #** |
| 1 | Submit Project Charter | 09/07/2022 | 1 |
| 2 | Ensemble trained model with preliminary results of the training and testing. | 09/30/2022 | 2 |
| 3 | Flask and Heroku application. (customer will choose whether to deploy it locally or on the cloud) | TBD | 3 |
| 4 | Final report of the project in addition to the cloud (heroku) deployed link. | TBD | 4 |

# Introduction

To make effective recommendations, we must understand what constitutes a specific genre. What is the difference between a hip-hop song and a pop song, and how is this discrepancy represented by data? Is it the tempo? Energy? Danceability? Our project will focus on analyzing how one genre is different from one another, how artists from the same genre could differ based on style, and how to accurately classify songs into the proper genre based on the features from the data. This will allow us to generate playlists and recommendations for users based on their preferences of genres and/or artists.

# Analysis of the dataset and Trained Model

## Exploratory Analysis and Visualization

The following graphs show the distribution of selected variables. The dataset does not have any missing values.

Chart, line chart

Description automatically generated

Figure : Distribution of Danceability

Figure 1 shows the distribution of the variable, Danceability. Price is approximately normally distributed. Figure 2 shows a correlation matrix. Graphical user interface

Description automatically generated

Figure 2: Variation of Genres based on Popularity and Danceability

Chart, treemap chart

Description automatically generated

Figure 3: Correlation of Variables

## Baseline Model

Since this is the classification dataset, we have decided to analyze the model using various classiers. For simplicity, all the model code is provided with an attached jupyter notebook, not in the report. We measured the accuracy of the classifers by precision score. Logistic Regression was the best performing model with a precision score of 0.59. A confusion matrix can be seen below which shows a majority of rock songs were currently identified as rock songs.

Diagram

Description automatically generated

Figure 4: Confusion Matrix of Logistic Regression

# **Model** Selection

Obtaining a baseline provided us necessary insight for model selection. Following the preliminary regression analysis, we have decided to evaluate, AdaBoost Classifier, Bagging Classifier, Random Forest Classifier, Decision Tree Classifier, Logisitc Regression,and Gradient Boosting, and K Neighbors as candidate models. Figure 4 shows the model results, and we can see Logisitic Regression and Random Forest classifiers perform the best, since they have the best precision scores. This means that they were better at predicting the genres. Because of this, we will select them as the candidate models for level 0 in the stacking regression. We decided to use Bagging Classifier, and Gradient Boosting Classifier and K Neighbors as the level 1 combiner or metamodel to aggregate the results of the level 0 models.

## Model Performance Evaluation

Below you will see the different precision scores of each classifer. Logistic Regression and Random Forest performed the best at approximately 60%. Decision Trees yielded the lowest performance at 44% precision as can be seen in Figure 6. For this reason, this learning algorithm was omitted from the stacked model. The stacked model when in comparison to the other individual algorithms yielded a precision score of 57%. Although this is performance is similar to those of our other learners, it still does not outperform logistic regression in our case. However following the trend seen in the confusion matrix above, the stacked model also misclassifies many pop songs as rock songs.

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Figure 6: Bar Chart of Precision Scores