

Final Project Progress Report

Report two

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So far, we have finished implementing the basic structure of the program. Our main class is a do while loop, if the user doesn't want to quit, the program will continue by load conversation class. Conversation class is a class that will require commands from user. Based on different commands, the class will redirect the user to different classes. For example:

Menu: 1.Like | 2.Explore | 3.Daily Selection | 4.quit

If user types "1", then the program will direct the user to liked song list, then like class will show user the music list. The like class will also ask the user how to sort the list. The rest of the important classes, such as *explore()* and *dailySelection()*, will implement our random suggesting system and forecast suggesting system.

The explore class reads the Cloud musiclist CSV file and randomly selects a few songs and feeds it to user, based on song's name initial letter sequence. If the user likes any of the songs, we will ask if the user want to add it to his liked music list. If they do, then the program will mark a song with y in the last column. and add it to like list. *need more stuff here*

In the daily selection class, the program will select songs that the user may like based on the forecasting of whether the user will like it (The algorithm will be explained later). *need more stuff here*

Algorithm Picking

For the algorithm for forecasting potential user like song, which is Collaborative filtering, there are two algorithms that we know.

Memory-based

The first one is user grouping method. In this method, the algorithm will first track down people who have similar activities, then recommend songs to each other by their playlists. For example, scores are assigned to different actions:

Repeat song = 5, Share = 4, Like = 3, Play = 2,
Played completely = 1, Skip = -1, Dislike = -5

Then, a person's preference would be a N-Dimensional vector. Here N is the number of songs by default. By using the cosine of the vector angle, which is generated by two vectors, we can know how similar two users can be. The cosine of 0 degree, which means two people are exactly the same, is 1. The cosine of 180 degree, which means two people have the opposite preference, is -1. The equation is:

Assume we have vector \vec{a} and \vec{b} ,

$$\cos(A) = \frac{\vec{a} \cdot \vec{b}}{|\vec{a}| \times |\vec{b}|}$$

This prediction method is very accurate, but there are too many calculations and it is not accurate for a new user.

Model-based

As for the second method, the algorithm will analyse the playlist of a person, then do recommendations. Models such as the latent factor are used in this method. ***I would explain more here***

Of course, in the real world, if we implement algorithm based on song, the second dimension factor will rapidly increase. Plus there will be too many songs. Therefore, it will be more efficient if we calculate based on tags of the song. For example, for “The Sound of Silence” by Simon & Garfunkel, tags can be Folk Rock, 60s, and Movie Track. Due to the time we have is limited, we would not build tags databases, therefore we would analyse by songs. Additionally, we would use the second method to calculate scores because we do not have time to build multiple users.