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# **Recommendation of Songs for Next Generation with Square Euclidean Distance**

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**ABSTRACT:** Recommender systems are widely implemented in E-commerce websites to assist customers in finding the items they need. A recommender system should also be able to provide users with useful information about that might interest them. The ability of promptly responding to changes in user's preference is a valuable asset for such systems. The recommender system presents an innovative recommender system for music data that combines two methodologies; the content based filtering technique and the interactive genetic algorithm. The recommender system provides recommendation by collecting user's profiles and discovers relations between each profile. Today increasing numbers of people are turning to computational recommendersystems. Emerging in response to the technological possibilities and human needs created by the World Wide Web, these systems aim to mediate, support, or automate the everyday process of sharing recommendations. The main goal is to identify challenges and suggest new opportunities.

**KEYWORDS**: Recommender System, User's preference, Music Extraction ,*Interactive* genetic algorithm, Content based filtering

## **I.INTRODUCTION**

When users browse through a web site they are usually looking for items they find interesting. Interest items can consist of a number of things. For example, textual information can be considered as interest items or an index on a certain topic could be the item a user is looking for. Another example, applicable for a web vendor, is to consider purchased products as interest items. Whatever the items consist of, a website can be seen as a collection of these interest items. Recommender systems are widely implemented in E-commerce websites to assist customers in finding the items they need. A recommender system should also be able to provide users with useful information about the item that interest them. The ability of promptly responding to the changes in user's preference is a valuable asset for such systems.

A Recommender system for music data it proposed which assists customers in searching music data and provides result with items resulting in own user preference. This system first extracts unique properties of music like pitch, chord, and tempo from the music file using MARSYAS software. This extracted data is then stored on the database. Each stored property is analyzed using content Based filtering and interactive genetic algorithm. After acquiring records, the system recommends items Appropriate to user's own favorite.

When we need to make a decision we often seek out the opinions of others. This is not only true for individuals but also true for organizations .The goal is to design a recommender system on music domain. The system will be composed of server side components and client side components. The server-side component will manage the database operations and algorithms that produce recommendation results. The client-side components will be graphical interfaces that are integrated into corresponding larger systems.

## 1.1 What is Recommender system?

Recommender systems or recommendation systems are a subclass of information filtering system that seek to predict the 'rating' or 'preference' that a user would give to an item (such as music, books, or movies) or social element (e.g. people or groups) they had not yet considered, using a model built from the characteristics of an item (content-based approaches).



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Recommendation Systems forms a specific type of information filtering technique that attempts to present information items. (like movies, music, books, news, images, etc.)

A Recommender system for music data is proposed which assists customers in searching music data and provides result with items resulting in own user preference. Recommender systems have become extremely common in recent years.[1]



Figure 1: A Recommender System

## **1.2** What is Content based filtering?

A common approach when designing recommender systems is content-based filtering. Content-based filtering methods are based on information about and characteristics of the items that are going to be recommended. In other words, these algorithms try to recommend items that are similar to those that a user liked in the past (or is examining in the present). In particular, various candidate items are compared with items previously rated by the user and the best-matching items are recommended.

>> key issue with content-based filtering is whether the system is able to learn user preferences from user's actions regarding one content source and use them across other content types. When the system is limited to recommending content of the same type as the user is already using, the value from the recommendation system is significantly less than when other content types from other services can be recommended.[3].



Figure 2: Content based Approach

# **II. RELATED WORK**

## 2.1 Recommender Systems:

The main issue of a recommender system is how to recommend items tailored with user's preference from resources. The recommender system also has to recognize and provide this matter, there are two main approaches:

Content based filtering and Collaborative filtering

• In the collaborative filtering approach, the recommender system provides recommendation by collecting user's profiles and discovers relation between each profile. After identifying correction of each profile, the system classifies users having profiles that are similar to the others. The system then recommends items derived from other profiles in the same group.

• Collaborative filtering

The advantage of this approach, thus is that it has a high possibility to recommend items corresponding with user's preference by providing environments in each user can share his or her profile.



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## 2.2 Music Extraction:

The feature extraction is a technique that derives properties from specific data, such as music, document and photo. It also useful for grouping statics resources. In our proposed system, it employ the content based filtering to acquire information from music data. The analysis of item is an essential step of filtering items in the content based filtering approach. We then use a Feature extraction tool (MARSYAS) to analyze the properties of items.

The software framework for research and application development on the audio and music domain, The MARSYAS Software provides complex audio signal analysis, transformations and synthesis.

## 2.3 Interactive genetic algorithm:

Genetic Algorithm (GAs) are stochastic search methods inspired from the mechanism of natural evolution and genetic inheritance. Gas work on a population of candidate solutions; each solution has a fitness value indicating its closeness to the optimal solutions of the problem. The solution having higher fitness values than others are selected and also survive to the next generation. GAs then produce better offspring by the combination of selected solutions. The methods can discover preserve and propagate promising sub-solutions.

# **III. GENETIC ALGORITHM**

## **Introduction of GA:**

The evolutionary process of a GA is a highly simplified and stylized simulation of the biological version. It starts from a population of individuals randomly generated according to some probability distribution, usually uniform and updates this population in steps called generations. Each generation, multiple individuals are randomly selected from the current population based upon some application of fitness, bred using crossover, and modified through mutation to form a new population.

## Genetic algorithm procedure:

- 1. Choose the initial <u>population</u> of <u>individuals</u>
- 2. valuate the <u>fitness</u> of each individual in that population
- 3. Repeat on this <u>generation</u> until termination (time limit, sufficient fitness achieved, etc.):
- 4. Select the best-fit individuals for <u>reproduction</u>
- 5. <u>Breed</u> new individuals through <u>crossover</u> and <u>mutation</u> operations to give birth to <u>offspring</u>
- 6. Evaluate the individual fitness of new individuals
- 7. Replace least-fit population with new individuals

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• Crossover – exchange of genetic material (substrings) denoting rules, structural components, features of a machine learning, search, or optimization problem.

• Selection – the application of the fitness criterion to choose which individuals from a population will go on to reproduce.

- Replication the propagation of individuals from one generation to the next
- Mutation the modification of chromosomes.



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# PHASES OF GENETIC ALGORITHM IN RECOMMENDER SYSTEM:

Figure 3: The process of Interactive GA phase [1]



The following are phases of genetic algorithm are as follows:

**1.** Selection phase – Music features are extracted using MARSYAS software. The items which are rated above threshold value are selected rest other items are ignored.

**2.** Crossover phase – The BLX – alpha crossover algorithm is used since extracted features are real numbers. Hence crossover is performed with this algorithm resulting in new generation.

# 3.

BLX-a:

# Input:

- 1. Select two parents X(t) and Y(t) from a parent pool
- 2. Create two offspring X(t+1) and Y(t+1) as follows:
- 3. for i = 1 to n do
- 4. di = |xi(t) yi(t)|
- 5. Choose a uniform random real number u from
- 6. interval < min(xi(t),yi(t))-adi,
- 7.  $\max(xi(t),yi(t))+adi >$
- 8. xi(t+1)=u
- 9. Choose a uniform random real number u from
- 10. interval
- 11. (xi(t),yi(t))-adi, max(xi(t),yi(t))+adi)
- 12. yi(t+1)=u
- 13. end do
- 14. where: a positive real parameter

## **Output:**

- to provide users with useful information about that might interest them.
- the system recommends items Appropriate to user's own favorite.

**4. Matching phase-** This phase finds the similar items stored in database to the newly Generated music features. Once similarity is found those items are recommended to the User.

This phase uses Euclidean distance between two offspring and distance between Each feature of the two offspring is calculated, resulting value is used to match the records Stored in the database those records are compared with the resulting value which the user has given highest rating to the tracks.



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# IV. DATA FLOW ARCHITECTURE OF SYSTEM



Figure 4: The flow Architecture of the system

## V. THE EXPERIMENT AND RESULTS

The system will display music files from the database based on the ratings given by the user and by applying genetic algorithm. The recommender system should allow any user to give ratings to the music items. The system will provide error messages to help direct the users through various options. Once the user has given rating to the music files these selected files will be stored onto the database. Each user details from profile page are stored directly to the database.

#### 5.1 Implementation

The recommended system incorporates with this system, which is implemented in java Servlet page (jsp), the information gathered in the previous step (feature extraction phase ). After that then build a website provides essential information such as artist name and song title; The test strategy consists of a series of different tests that will fully exercise the system. The primary purpose of these tests is to uncover the systems limitations and measure its full capabilities. Any data item will have certain constraints like, the type of the data, length of the data, unique data and compulsory data and when getting these data from the user, appropriate tests should be done to ensure that the data item confirm to the constraints.

Each time a user evaluates a page of 10 times. The initial page is randomly successive page is constructed based on the user evaluation of the preceding one. In order to help user's evaluation, It provide a function that user can listen to the music tracks which they are not familiar with.

#### 5.2 Experiment Results

The tests will focus on the behavior of the system. User scenarios will be executed against the system as well as screen mapping and error message testing. Overall, the system tests will test the integrated system and verify that it meets the requirements.

Here we can store unlimited songs file in the database and the local server. The songs page displays 10 music files on each page. It is based on Simple and Square Euclidean distance.



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Figure 5: The Experiment Results

The Table shows Square Euclidean Distance

	Pitch	Tempo	Octave	Root	Mode
ltem1	0.0451	0.0334	0.1244	0.123	0.257
ltem2	0.31461	0.458	0.03954	0.038542	0.02584
ltem3	0.04033	0.01133	0.04033	0.74033	0.13933
ltem4	0.186	0.4618	0.0946	0.09761	0.1379
ltem5	0.04037	0.01137	0.04037	0.74037	0.16937
Item6	0.04038	0.01138	0.04038	0.74038	0.16938
ltem7	0.04039	0.01139	0.04039	0.74039	0.16939
Item8	0.04089	0.01189	0.04089	0.74089	0.16939
Item9	0.04137	0.02137	0.04137	0.74137	0.16137
Item10	0.161	0.34621	0.128743	0.15749	0.285



Figure 6: Square Euclidean Distance



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Fig. 5 shows the performance of the proposed recommender system. We did not consider results of the first page since users tend to rate too much higher values for times on the page than other pages. Beside, the users need to become familiar with the rating mechanism of our systems. Also, We measured the number of items that were rated over 80 for each user pages because these items have more impact to trace user preference than others. As seen in the figure, the average scores gradually improve as the page number proceeds. It denotes that the users were satisfied with the recommendation made by our system. In other words, The results prove the efficiency of our recommender system in detecting and tracing the preference of each other.[1]

## VI. CONCLUSION

We proposed real-time genetic recommendation method in order to overcome the existing recommendation techniques that are not reflecting in the current user's intend. With the genetic algorithm newer solutions can be generated providing optimal solution each time when the algorithm is made to run, thus providing mutations.

This method can be compared with the existing ones which lack the quality of providing accurate results.

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