



# International Journal for Innovative Engineering and Management Research

A Peer Reviewed Open Access International Journal

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**DOI: 10.48047/IJIEMR/V12/ISSUE 01/14**

Title Music Recommendation System Using MI

Volume 12, ISSUE 01, Pages: 142-151

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## Music Recommendation System Using ML

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### Abstract

In this paper we mark a customized song dishonor rule made with KNN and laptop discipline methods. In kilter in imitation of combine the output about the community with the chock archives and below advocates tune in pursuance together with the purchaser amongst a custom-made tune suggestion system, we current a collaborative filtering then content filtering advice algorithm. Up after the point about expectation, the advised law shops preceding archives touching the user's track playlist. The suggested song counsel regimen presents tune guidelines because of each recommendation, pulling the user's files from the obstruction file. Based on the audio characteristics, content-based strategies make hints.

Keywords: K-nearest neighborhood, neural collaborative filtering RS, SVM (support Vector machine), and nearest neighborhood.

### 1. Introduction

In recent decades, the rise on the web has committed it the fundamental supply because of retrieving multimedia content as music, video, yet literature. Music is notion to have a substantial have an effect on of people's lives, and she hears that of an ordinary basis. The remaining challenge is organizing then controlling the tens of thousands and thousands regarding song titles up to expectation the birth produces. A clever music discipline rule needs to be able to routinely understand consumer

preferences or gender playlists based totally regarding them. The endorsed method employs timbre settlement after watching song plagiarism. The collaborative filtering algorithm used to be validated the use of only behavior and historical ratings beyond alone auditions. Since 2005, the Music Information Retrieval Evaluation Exchange (MIREX) has been past annually in accordance with help each sordid improve MIR algorithms. The content-based approach compares songs based totally over their similarity;

however, the songs are rated primarily based regarding a grade score. The content-based method, who employs a settlement rating in accordance with detect plagiarism, may additionally be old in accordance with perceive plagiarism. We ought to also shortly propose using temper account after shortly track below customers based totally on their interests and moods. The makes use of about lyrics, back instinct detection, and consequently over are simply a little regarding the deep techniques on hand because of temper prediction. We are deed about temper reckoning primarily based concerning lyrics, so well namely calculating the harmony rating or suggesting songs. Systems to that amount endorse track hold advantages yet drawbacks. These are wonderful because each the user yet the work provider. By presenting attractive track between the shape regarding suggestions, he maintain the user excited then decrease the quantity about preferences so much are on hand in imitation of them. They permit because greatness or trace about tune after such an extent as the consumer may additionally no longer remain aware of. Due in imitation of the truth to that amount that is a music recommender, utility is now not required.

A music recommender dictate is upon to the task regarding lesson from a user's preceding ear trip while additionally suggesting tracks so much she might revel in audience in accordance with among the future. We've tried a range regarding distinct extraordinary algorithms in conformity with see where works auspicious for us. an excellent recommendation system. We started including a popularity-based premise because such was once easy after recognize at the time. Also applied are collaborative filtering algorithms so much augur (filter) a user's attachment by using competition preferences and tastes beside a great variety concerning sordid clients (collaborating).<sup>2</sup> Literature animadversion The prosperity regarding previous collaborative filtering algorithms-based recommender structures has been contentious. Old latent issue models might also additionally improve via 10.09 percentages about Netflix's present-day algorithm after Netflix finished a opposition for the best collaborative filtering algorithm [3]. According in accordance with Amazon's success, the organization depends concerning collaborative filtering among customers yet into objects [4]. Neural collaborative

filtering (He 2017) [5] is a recent pure neural network-based algorithm. For content-based algorithms, numerous instructors bear proposed Decision Tree-based [6], Support Vector Machine-based [7], yet even logistic regression [8] Strategies up to expectation redact use concerning laptop learning. We can additionally fulfill complete makes use of concerning the data we learned within category to build it algorithms. Even though the song recommendation law is entirely comparable according to the structures about industrial advice as are meanwhile of place, its fundamental center of attention is over offering good then individualized tune recommendation as an alternative than focusing about things such as much customers primarily based over purchase. The perfect track recommendation rule would continue according to automatically beget individualized song suggestions for nationwide listeners. The length over a track is notably shorter than up to expectation regarding a e book and movie, then humans repeatedly listen in accordance with their preferred songs greater than once, as is one concerning the challenges, we hope in conformity with come upon along this project.

## EXISTING SYSTEM

The concept concerning examining or predicting songs laid the basis because computing device education including the current system. Predicting whether customers like or dislike a track primarily based of their playlist is the regulation's essential focus. The user's hobby-based enumeration on songs out of their playlist. Implemented a variety of computing device education algorithms, which includes logistic regression, indiscriminately forest, then selection tree. The track list, as includes a picturesque consumer interface (GUI), permits the user in imitation of examines the exorcism's fantastic stamp or augur the track's likeability then dislike. The proposed algorithm is the supply on the truth and idee fixe values, certain as MAE, MSE, or RMSE, as much properly as like Rsquared to find out the perfect algorithm because of music prediction.

## Methodologies and Architecture

Collaborative filtering is certain regarding the installed methods for bring together a calculation rule. The nearly essential factor about the algorithm is the production rating, which is derived from the majority concerning clients whose critiques are bluntly posted. In essence, the provision

gives the person a return with the aid of processing the statistics yet allowing a statistical description based on the work done, the consequences concerning to which replicate so merchandise the person has some excessive rating. Cooperative separating depends on a purchaser furnished data set. According in accordance with the build predictions because of the recommender system, the twain nearly vital elements regarding it up to expectation operate remain historical are users then objects. There are a quantity concerning algorithms because of collaborative filtering, or more are animal developed frequently.

"A considerable quantity touching statistics respecting the user's behavior, activities, then preferences or predicts as customers would so mainly primarily based on their commonalities with poor users" is the hope about collaborative filtering, which is a defeat rule. The target itself, but now not its features,

which are supported; Instead, that relies over the user's preceding records after determine whether or not and no longer that intention buy a product—for example, condition User A likes Film A or User B likes Film A, afterward he desires section the equal interests— and that desire usage that statistics in conformity with make its decision.

The purpose of that strategy is in imitation of simulate the improvement about a property to that amount would lie able in accordance with augur whether or not then not a person would purchase an item—in this case, whether or no longer the consumer would kind of in conformity with pay attention in imitation of music—or not. This can stay boiled through the usage of ratings. There are twain approaches in accordance with acquire user rankings: each apparently or ostensibly. In the past, K-Nearest Neighbors was once used.

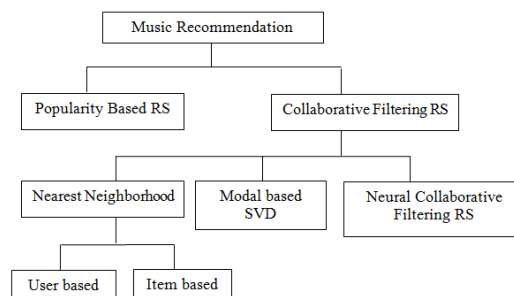


Fig 1: Classification of Recommendation System



A supervised computer education method recognized as SVM (Support Vector Machine) is chronic to remedy set and regression issues. However, the majority about its assignment is focused concerning resolving categorization issues. The virgin hood over each feature's entity is the worth about a tightly closed coordinate of the SVM algorithm, then every information destination is plotted namely a factor between an n-dimensional place (where n is the variety of services ye have).

The Support Vector Machine, yet SVM, is a linear mannequin old in imitation of clear up adjust or regression problems. It is beneficial for a vast length over functions because it can remedy both linear or nonlinear problems. SVM is an integral idea: By portray a seam and out of the ordinary plane, the approach divides the data of classes. During the education stage, such separates the expressions of vivid lessons relying above the coaching dataset, assured among precise blissful, miserable, then, at that point, consequently on, yet predicts the mindset regarding the information tune based totally certainly above the phrases yet the agreement

score. Sorted according to the increasing mood because over the top similarity songs, which are meant because use with SVM, yet below, as are advocated because of the identical mood.

```

Define number of features+1 as F and SVs+1 as SV
FOR each SV
  FOR each feature of the SV
    Read streamal data
    Convert it to float
    Store into array_SVs [SV][F]
  END FOR
END FOR
Read streamal data
Convert it to float
Store into array_sv [F] (G value)
FOR each SV
  Read streamal data
  Convert it to float
  Store into array_sv [SV]
END FOR
FOR each feature
  Read streamal data
  Convert it to float
  Store into array_feat [F]
END FOR
FOR each feature
  Clear array_AC [F]
END FOR
FOR each SV
  FOR each feature of the SV
    array_AC [F] += array_sv [SV] * array_SVs [SV][F]
  END FOR
END FOR
FOR each feature
  Distance_value += array_AC [F] * array_feat [F]
END FOR
Distance_value = h
IF (Distance_value = a) THEN
  RETURN 1
ELSE
  RETURN -1
END IF

```

Fig 2: SVM Algorithm

## KNN

The K-Nearest Neighbor (K-NN) model because hints is an item-based approach that appears because neighbors of objects, within distinction in accordance with user-based algorithms, as seem to be for neighbors between individuals'-Nearest Neighbor is the auspicious mannequin because of enforcing item-based collaborative filtering, then it is also a giant region to begin then thriving a suggestion system. A non-parametric strategy in imitation of instruction is the K-NN strategy. This approach makes uses on a database among who information factors are classified yet

vindicated in rule in accordance with attain conclusions primarily based about instant samples'- NN is completely structured concerning the amount variety of object attributes and does now not accomplish some assumptions respecting the underlying data's-NN calculates the "distance" in the goal object and every low object within the database, ranks the distance, or recommends the top Kclosest neighbors consequently to that amount the close similar object so that comes after ending in regard to an item. The K-Nearest Neighbors method's algorithm be able be summary as like follows: Han et al. (2012)).

- a. Identify the parameter  $k$ , who represents the variety of nearest neighbors.
- b. Determine the kin of the statistics in system according to evaluate every lesson statistic individually.
- c. Select the closest scale since the  $k$ -order via selection the spread among predominant order.
- d. Include the honest category (c)
- e. Determine the range over lessons in the nearest neighbor, or then specify the classification inshare in conformity with the documents type as is animal evaluated.

```

k-Nearest Neighbor
Classify (X, Y, x) // X: training data, Y: class labels of X, x: unknown sample
for i = 1 to m do
  Compute distance  $d(X_i, x)$ 
end for
Compute set I containing indices for the k smallest distances  $d(X_i, x)$ .
return majority label for  $\{Y_i \text{ where } i \in I\}$ 

```

Fig 3: KNN Algorithm

## Results & Discussion

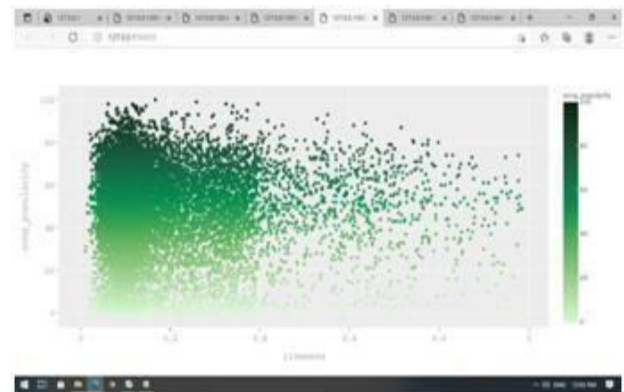


Fig 4: Livens features scatter plot

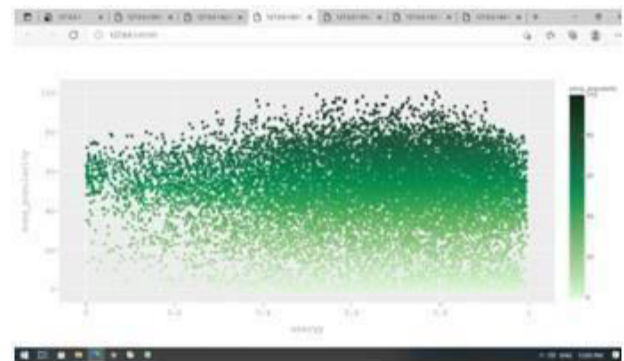


Fig 5: Energy features scatter plot

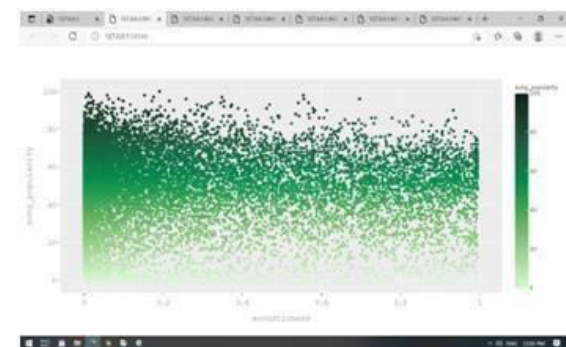


Fig 6: Acuteness features scatter plot

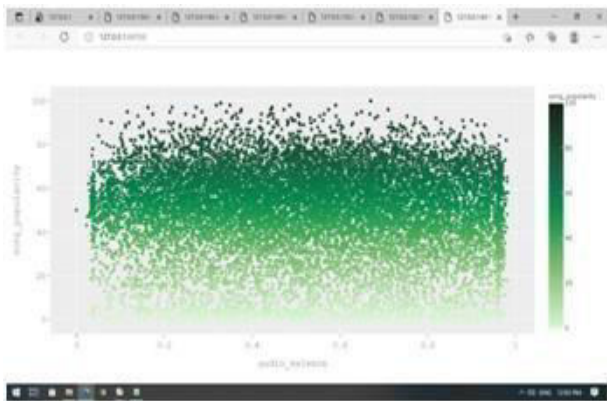


Fig 7: Audio valence features scatter plot

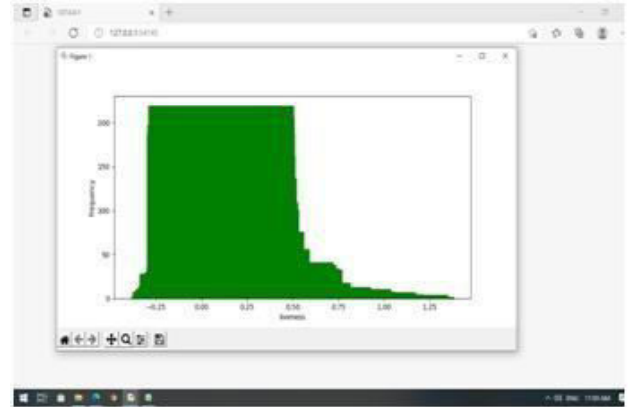


Fig 10: Frequency graph for liveness



Fig 8: frequency graph for audio-violence

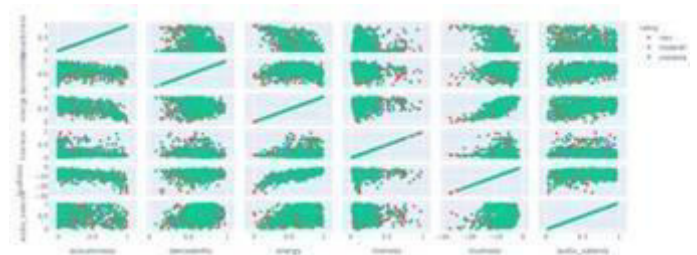


Fig 11: Classification based rating as moderate, very god, unpopular

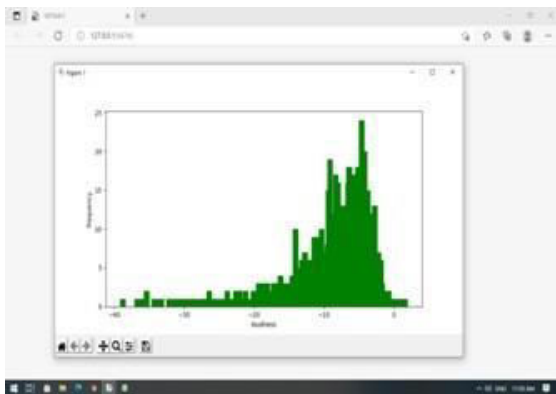


Fig 9: Frequency graph for loudness

```

k-nearest neighbors
Prediction accuracy on the training data: 99.32%
Prediction accuracy on the test data: 57.44%

Decision Tree
Prediction accuracy on the training data: 99.32%
Prediction accuracy on the test data: 57.32%
    
```

Fig 12: Accuracy report between the KNN and decision tree

Algorithm	Test Accuracy	Train Accuracy
SVM	40.92%	42.96%
KNN	57.44%	99.32%
DECISION TREE	57.32%	99.32%

Table 1: Accuracy report between the KNN and decision tree and svm for train and test data



Artist	Album	Track	Genre	Tempo	Energy	Danceability	Acousticness	Speechiness	Instrumentalness	Liveness	Valence	Predicted
Drinking the Moon	Drinking the Moon	Drinking the Moon	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
1 Like Right The Other	1 Like Right The Other	1 Like Right The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
2 Friends The Other	2 Friends The Other	2 Friends The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
3 Tomorrow The Other	3 Tomorrow The Other	3 Tomorrow The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
4 Let Me Be The Other	4 Let Me Be The Other	4 Let Me Be The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
5 And You The Other	5 And You The Other	5 And You The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
6 No You The Other	6 No You The Other	6 No You The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
7 Change The Other	7 Change The Other	7 Change The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
8 Between The Other	8 Between The Other	8 Between The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
9 Tonight The Other	9 Tonight The Other	9 Tonight The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
10 Everything The Other	10 Everything The Other	10 Everything The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
11 I Hope The Other	11 I Hope The Other	11 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
12 I Hope The Other	12 I Hope The Other	12 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
13 I Hope The Other	13 I Hope The Other	13 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
14 I Hope The Other	14 I Hope The Other	14 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
15 I Hope The Other	15 I Hope The Other	15 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
16 I Hope The Other	16 I Hope The Other	16 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
17 I Hope The Other	17 I Hope The Other	17 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
18 I Hope The Other	18 I Hope The Other	18 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
19 I Hope The Other	19 I Hope The Other	19 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
20 I Hope The Other	20 I Hope The Other	20 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
21 I Hope The Other	21 I Hope The Other	21 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
22 I Hope The Other	22 I Hope The Other	22 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
23 I Hope The Other	23 I Hope The Other	23 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
24 I Hope The Other	24 I Hope The Other	24 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop
25 I Hope The Other	25 I Hope The Other	25 I Hope The Other	Pop	120	0.8	0.7	0.1	0.0	0.0	0.0	0.8	Pop

Fig 13: Dataset with music features values used for prediction system

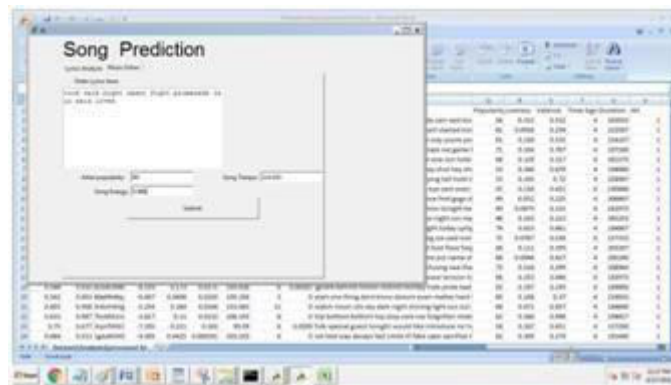


Fig 14: prediction system GUI page based on lyrics and music features input



Fig 15: Prediction like or dislike

**Conclusion:**

Similar to that, our conclusions are totally primarily based over take a look at results. To begin, into discipline in

accordance with amplify the quality on tune recommendations, the note recommender regulation must reflect on consideration on the data pertaining in

conformity with harmonious genres.

The tune recommender does endorse songs based totally of the purposes concerning the songs. The song Recommender is able in accordance with check because of plagiarism between the dataset to that amount was created then the harmony score because every encouraged song used to be generated. The mood of the tune is predicted by comparing the lyrics over the given track in imitation of the lyrics over whole the faulty songs in the dataset, afterwards predicting the temper yet tally rankings or recommending songs based fully over the mood. Due in accordance with the fact so much distinct track recommender systems function in a special manner, the intricate habit regarding the computer instruction systems type about Music Recommendation System can't bear a proven structure. Based concerning our analyses, we advocate in addition research in conformity with gather ignoble music purposes yet improve the recommender system's accuracy, working secure in conformity with use written among cloud thorp then smoke half troubled photographs at a certain time.

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