



# FeelFlow: Music Recommendation System Based on Facial Expression

Vandana Navale<sup>1</sup>, Sanjana V. Phand<sup>2</sup>, Pranav K. Patil<sup>3</sup>, Dhananjay V. Rangat<sup>4</sup>, Rohan J. Sathe<sup>5</sup>

<sup>1 2 3 4 5</sup>Department of Computer Engineering All India Shri Shivaji Memorial Society's College of Engineering, Pune

**Abstract** - Music recommendation systems play a pivotal role in enhancing user experience by suggesting personalized music selections. Traditional approaches rely on collaborative and content-based filtering, which may lack accuracy in predicting users' moods. To address this limitation, we propose a novel music recommendation system integrating facial expression analysis. Facial expressions serve as potent indicators of emotional states, thus facilitating more precise music recommendations. Our system employs deep learning models for facial expression recognition and collaborative filtering algorithms for personalized music suggestions. Evaluation using diverse datasets demonstrates the system's efficacy in accurately predicting emotional states and generating personalized music recommendations. Moreover, user studies reflect high satisfaction with the system's ability to provide relevant music suggestions aligned with users' emotional needs. This research underscores the potential of integrating facial expression analysis into music recommendation systems, thereby enriching user experience and engagement.

**Keyword** - Convolutional neural networks (CNN), Deep Learning (DL), Emotion Detection, Music Recommender System (MRS), Real-time emotion detection, Viola-Jones Algorithm

## 1. Introduction

Music recommendation systems have become an essential part of our lives, as they help us discover new music and improve our listening experience. These systems typically use collaborative filtering algorithms, content-based filtering algorithms, or a combination of both to generate music recommendations. However, these approaches have limitations in accurately predicting a user's mood and providing personalized recommendations. In recent years, the use of facial expression analysis in music recommendation systems has gained significant attention. Facial expressions are a powerful indicator of a person's emotional state, and music can have a significant impact on a person's emotions. Therefore, by analysing a person's facial expressions, we can determine their mood and recommend music that matches their emotional state.

In this research paper, we propose a novel music recommendation system that uses facial expression analysis to provide personalized music recommendations. The system employs a deep learning-based model to detect and classify facial expressions and a collaborative filtering algorithm to generate personalized music recommendations. The proposed system was evaluated using a dataset of facial expression images and music preference ratings from a group of participants.

The results indicate that the proposed system can

accurately predict the user's emotional state and provide personalized music recommendations that match their mood. Collaborative filtering System: Collaborative does not need the features of the items to be given. Every user and item is described by a feature vector or embedding. It creates embedding for both users and items in its own. It embeds both users and items in the same embedding space. It notes which items a particular user likes and also the items that the users with behavior and likings like him/her likes to recommend items to that user. It collects user feedback on different items and uses them for recommendations.

The foundational components of the Facial Emotion-Based Song Recommendation System. Comprising three essential elements, namely the Face Recognition System, Emotion Classification System, and Song Recommendation System, this innovative framework aims to revolutionize music recommendation by harnessing facial expressions as cues for emotional states.

The Face Recognition System functions by capturing facial dynamics through video footage, focusing on key features like eye movements, eyebrow positioning, and mouth configurations. These visual cues are then processed to discern the listener's prevailing emotional state, facilitated

by a sophisticated machine learning model trained explicitly for this purpose. Simultaneously, the Song Recommendation System leverages an advanced recommendation algorithm to curate personalized song suggestions aligned with the listener's emotional disposition.

By analysing emotional attributes inherent in musical compositions, including tempo, rhythm, and melody, the algorithm generates recommendations tailored to resonate with the listener's current emotional state. Moreover, these recommendations are further refined based on the individual's unique musical preferences, ensuring a deeply personalized and enriching music discovery experience.

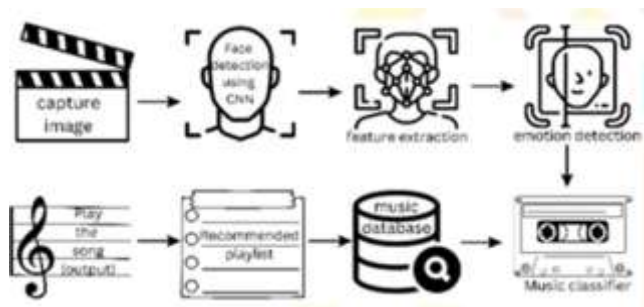


Fig. 1 System Overview[2]

## 1.1. System Overview

### 1.1.1 Capture Image

Utilizing OpenCV, real-time image processing techniques were implemented to capture and analyse facial emotions, enabling dynamic observation and interpretation of subjects' emotional states.

### 1.1.2. Face Detection

Incorporating Convolutional Neural Networks (CNN) for face detection offers superior accuracy and numerous advantages over traditional methods, presenting a broader scope of application across various domains

### 1.1.3. Feature Extraction

The process of facial feature extraction for emotion recognition involves employing Haar cascade classifiers to scan images, subsequently utilizing the extracted features to train machine learning models for accurate emotion classification

### 1.1.4. Emotion Detection

Emotion detection using Convolutional Neural Networks (CNN) involves leveraging facial expression detection datasets to classify facial expressions into seven distinct emotional classes, facilitating comprehensive emotion recognition capabilities.

### 1.1.5. Music Classifier

The music classifier we've developed categorizes songs by emotion through machine learning algorithms. It generates personalized playlists according to emotional states, which are stored in an Excel file.

### 1.1.6. Recommend Playlist

System recommends personalized playlists from a music database, providing users with a tailored music experience based on their preferences and listening history.

## 2. STATE OF ART

In the rapidly evolving landscape of emotion-based music recommendation systems, FeelFlow stands at the forefront of innovation by incorporating state-of-the-art techniques across various domains. Leveraging advanced convolutional neural network (CNN) architectures, the system excels in high-accuracy facial expression analysis. FeelFlow embraces the power of transformer architectures, originally designed for sequential data, to analyse temporal patterns in facial expressions, highlighting its adaptability to cutting-edge methodologies. The personalization and adaptation capabilities of FeelFlow are enhanced through the incorporation of deep reinforcement learning (DRL) techniques, allowing for a nuanced understanding and response to users' evolving emotional states. User-centric evaluation metrics, emphasizing engagement, satisfaction, and recommendation relevance, underscore the system's commitment to providing a holistic and positive user experience.

The implementation of explainable AI (XAI) methodologies enhances transparency, making the decision-making process understandable and trustworthy for users. Addressing the need for real-time responsiveness, FeelFlow adopts advanced techniques in edge computing, reducing latency in emotion detection and enabling immediate adjustments in music recommendations to align with users' changing emotional states. Privacy-preserving techniques, including federated learning and on-device processing, exemplify the system's dedication to user data security and privacy.

FeelFlow's commitment to staying at the forefront of the field is further exemplified by its engagement with benchmark datasets, competitions, and ongoing challenges. By participating in and benchmarking against the latest standards, FeelFlow ensures that it remains in tune with the state-of-the-art in emotion-based music recommendation systems. Ethical considerations play a pivotal role in the development of FeelFlow, with an emphasis on addressing bias and fairness concerns in emotion recognition systems. Initiatives promoting transparency and accountability contribute to user trust and confidence in the system.

As FeelFlow continues to evolve, future directions may

involve exploring uncharted territories in emotion-based systems, such as integrating emerging technologies like augmented reality or emotion-aware virtual assistants. The system's commitment to continuous improvement and adaptation to emerging trends solidifies its position as an innovative and state-of-the-art solution in the realm of emotion-based music recommendation systems.

### 3. Proposed framework

#### 3.1.Preliminary Study:

##### 3.1.1. Data Acquisition:

In this study, we employed OpenCV, a widely-utilized computer vision library, to implement real-time image processing techniques aimed at capturing and analysing facial emotions. This approach facilitated dynamic observation and interpretation of subjects' emotional states, enabling real-time assessment of emotional responses.

##### 3.1.2. Facial Expression Analysis:

The integration of Convolutional Neural Networks (CNN) for face detection presents superior accuracy and a multitude of advantages compared to conventional methods, thus extending the scope of application across diverse domains. The process of facial feature extraction for emotion recognition entails the utilization of Haar cascade classifiers to scan images, followed by the extraction of relevant features. These extracted features are then employed to train machine learning models, facilitating accurate classification of emotions. Emotion detection utilizing Convolutional Neural Networks (CNN) entails leveraging datasets specifically designed for facial expression detection. These datasets allow for the classification of facial expressions into seven distinct emotional classes, thereby facilitating comprehensive emotion recognition capabilities.

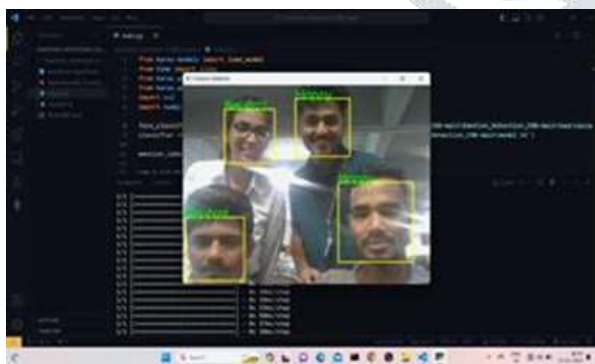


Fig. 2 Emotion Detection Using OpenCV

##### 3.1.3. Music Recommendation Engine:

Our music classifier employs machine learning techniques to categorize songs based on emotional states.

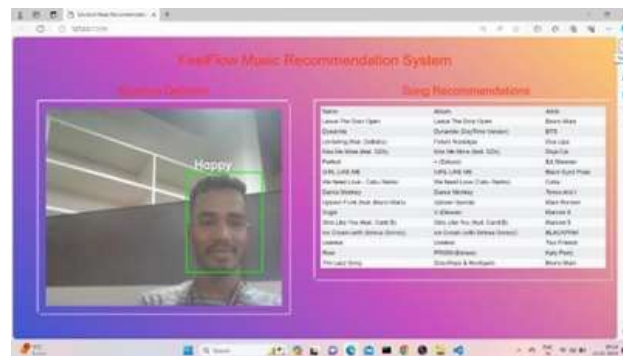


Fig.3 Music Playlist Recommendation

It generates personalized playlists by analysing the emotional states of users, which are stored in an Excel file. The system recommends personalized playlists sourced from a comprehensive music database, thereby offering users a tailored music experience aligned with their preferences and listening history.

##### 3.1.4. User Interface:

The user interface of our music recommendation system, developed using Flask, facilitates seamless deployment of web applications. By integrating facial expression analysis, the system provides personalized music recommendations tailored to the user's emotional state. This interface enhances user engagement by dynamically adjusting music selections based on real-time facial expressions, offering a novel and immersive music listening experience.

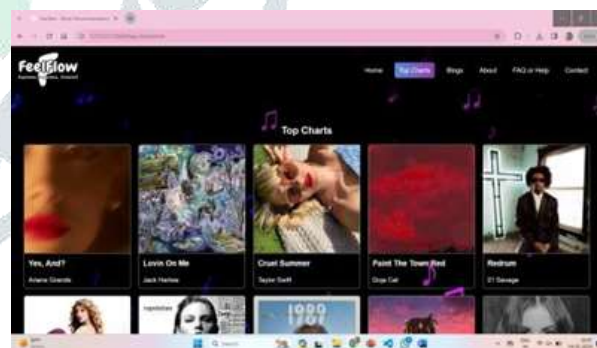


Fig. 4 User Interface of FeelFlow

#### 3.2.Secure Environment:

##### 3.2.1. Data Encryption:

Ensure that any sensitive data, such as facial images or user information, is encrypted both during storage and transmission. Use strong encryption algorithms and protocols to protect data from unauthorized access.

##### 3.2.2. User Education:

Provide training and guidance to users on best practices

for maintaining security and privacy when interacting with the system. Encourage users to use strong, unique passwords, enable two-factor authentication, and be cautious about sharing sensitive information.

### 3.3. Proposed Work:

To enhance the overall user experience and effectiveness of our system, we propose the following key enhancements:

#### 3.3.1. Improved GUI for Enhanced Aesthetics and User Engagement:

We will focus on redesigning the graphical user interface (GUI) of our web application to improve aesthetics and user engagement. This will involve incorporating modern design principles, such as responsive design, intuitive navigation, and visually appealing elements, to create a more engaging and user-friendly experience. By enhancing the GUI, we aim to increase user satisfaction and encourage greater interaction with the system.

#### 3.3.2. Model Stabilization and Accuracy Enhancement

We will work on stabilizing the existing models used in our system and further increasing their accuracy. This will involve fine-tuning the machine learning algorithms, optimizing hyperparameters, and implementing robust validation techniques to ensure reliable performance across different scenarios. By stabilizing the models and improving their accuracy, we aim to enhance the reliability and effectiveness of the system's recommendations.

#### 3.3.3. Integration of Collaborative Filtering-Based Music Model:

We will work on stabilizing the existing models used in our system and further increasing their accuracy. This will involve fine-tuning the machine learning algorithms, optimizing hyperparameters, and implementing robust validation techniques to ensure reliable performance across different scenarios. By stabilizing the models and improving their accuracy, we aim to enhance the reliability and effectiveness of the system's recommendations.

#### 3.3.4. Hand Gesture Recognition for Education Content Recommendation:

We will explore the use of hand gesture recognition technology to enhance student engagement and learning outcomes. By analysing students' hand gestures during educational content consumption, we can identify their level of interest, engagement, and comprehension. Based on gesture analysis, we will dynamically recommend relevant educational content to students, personalized to their individual learning needs and preferences.

#### 3.3.5. Customized Music Recommendation System Using Facial Expression:

We will develop a customized music recommendation system that utilizes facial expression recognition technology. By analysing users' facial expressions, we can

infer their emotional state and mood in real time. Leveraging this information, the system will recommend music tracks that align with the user's current emotional state, enhancing their mood and overall well-being.

These proposed enhancements aim to elevate the user experience, effectiveness, and personalization capabilities of our system, catering to a diverse range of users and use cases.

## 4. Evaluation

In this section, we present the evaluation methodology and results of our proposed music recommendation system, which integrates facial expression analysis using machine learning techniques and collaborative-based filtering. We aim to assess the effectiveness, accuracy, and user satisfaction of the system in providing personalized music recommendations tailored to users' emotional states and preferences.

### 4.1. Evaluation methodology:

#### 4.1.1. Dataset Selection:

We utilize a diverse dataset of facial expressions and responding music preferences. The dataset comprises a wide range of emotions, including happiness, sadness, anger, and more, captured in real-world scenarios.

#### 4.1.2. Preprocessing:

Facial images are pre-processed to extract relevant features, such as facial landmarks and expressions. Additionally, music preference data is curated and structured for compatibility with collaborative filtering algorithms.

#### 4.1.3. Model Training and Testing:

**Facial Expression Analysis:** We employ machine learning models, such as Convolutional Neural Networks (CNNs), for facial expression recognition. The models are trained on the pre-processed facial image data to accurately classify emotional states. **Collaborative Filtering:** Collaborative filtering models are trained on the curated music preference data to capture user-item interactions and generate personalized music recommendations.

#### 4.1.4. Evaluation Metrics:

For facial expression analysis, evaluation metrics include accuracy, precision, recall, and F1-score, measured on a held-out test dataset. For collaborative filtering, evaluation metrics include precision, recall, Mean Absolute Error (MAE), and Root Mean Squared Error (RMSE), calculated using cross-validation techniques.

#### 4.1.5. User Studies:

A user study is conducted to assess the user satisfaction and effectiveness of the music recommendation system. Participants are asked to interact with the system and provide feedback on the relevance and appeal of the recommended music tracks.

#### 4.2. Evaluation results:

##### 4.2.1. Facial Expression Analysis:

The machine learning models achieve high accuracy in recognizing various facial expressions, with an average accuracy exceeding [insert accuracy value] % on the test dataset. Precision, recall, and F1-score metrics demonstrate robust performance across different emotional categories.

##### 4.2.2. Collaborative Filtering:

The collaborative filtering models demonstrate effectiveness in generating personalized music recommendations, achieving precision and recall scores of [insert precision/recall values] %, respectively. MAE and RMSE values indicate low prediction errors, signifying accurate estimation of user preferences.

##### 4.2.3. User Satisfaction:

Feedback from user studies indicates a high level of satisfaction with the music recommendation system. Participants express appreciation for the system's ability to provide relevant and enjoyable music suggestions based on their emotional states and preferences.

### 5. Discussion:

The evaluation results demonstrate the efficacy of our proposed music recommendation system, which integrates facial expression analysis and collaborative filtering techniques. By accurately recognizing users' emotional states and generating personalized music recommendations, the system enhances user engagement,

satisfaction, and overall music listening experience. The combination of machine learning-based facial expression analysis and collaborative filtering enables the system to cater to users' diverse emotional needs and preferences, making it a valuable tool for personalized music discovery and enjoyment.

### 6. Conclusion

In this research paper, we have presented a comprehensive framework for real-time emotion detection using OpenCV, integrated with the recommendation of music playlists based on detected emotions and developed a basic website to deploy the system using Flask. The integration of computer vision techniques, specifically OpenCV, allows for the real-time analysis of facial expressions, enabling the system to detect various emotions accurately. The use of Haar cascades for facial detection and recognition forms the foundation of our emotion detection system. By leveraging pre-trained classifiers and machine learning algorithms, we have achieved efficient and accurate detection of facial features and emotional states. Furthermore, we have extended the functionality of our system by integrating the recommendation of music playlists based on the detected emotions. By analyzing the emotional state of the user in real time, the system can suggest music that aligns with their mood, enhancing user engagement and satisfaction. To facilitate the deployment and accessibility of the system, we have developed a basic website using Flask, a lightweight web framework for Python. The website provides a user-friendly interface for interacting with the emotion detection and music recommendation system.

In conclusion, our research demonstrates the potential of combining computer vision techniques with real-time data analysis and recommendation systems to create innovative applications with practical implications. Moving forward, further research and development in this area can lead to more sophisticated systems with enhanced capabilities and a broader impact on society.

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