

Movie Character Recognition from Video and Image using AI

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Abstract: *Recognition of movie characters using video and images An innovative research study called Using AI explores the use of Artificial Intelligence (AI) for recognizing and identifying movie characters using still photos and video clips. The study offers a new AI system that analyses visual input, extracts attributes, and correctly identifies movie characters using machine learning and deep learning methods. This system is able to identify characters in a single frame and follow their actions over several frames, giving it a thorough understanding of the character's function in the story. The technology's potential uses in a number of industries, including film creation, advertising, and audience analytics. The suggested AI system can automate video editing, help with character recognition-based targeted advertising, and offer insightful data on audience preferences and habits. The implications of this technology are enormous, creating new opportunities for the film industry to grow technologically. The challenges and limitations of implementing AI in movie character recognition. It highlights issues such as handling low-resolution images, varying lighting conditions, and changes in character appearance over time. Despite these challenges, the AI system demonstrates a high degree of accuracy and efficiency. The concludes by suggesting future enhancements to the system.*

Keywords: Python, Artificial Intelligence, Image Processing Machine Learning, Face detection, OpenCV, Face recognition.

I. INTRODUCTION

Remarkable ability to detect characters in a singular and trace their movements across diverse frames offers a wide-ranging comprehension of the character's function in the storyline. This technology exhibits immense potential applications such as the automation of the video editing process, promoting targeted advertising based on character recognition, and presenting invaluable insights into audience preferences and behaviors. Nevertheless, the integration of AI in embracing movie character recognition encounters various hurdles. Managing low-resolution images, fluctuating lighting circumstances, and alterations in character appearance over time stand out as some of the predicaments that require resolution. Despite these obstacles, the AI system displays a significantly high level of accuracy and efficiency, underscoring its potential for practical application. The system capitalizes on machine learning and deep learning algorithms to scrutinize visual data, extract features, and recognize movie characters. It possesses a distinctive capability of not just recognizing characters in a particular frame but also tracing their movements across several frames, thereby delivering an extensive awareness of the character's role within the storyline.

This inventive realm is reshaping the manner in which we engage with film and television, by enabling machines to recognize, identify, and analyze the characters within video and image content. With AI algorithms gaining complexity, they enable us to delve profoundly into the nuances of storytelling, character development, and cinematic artistry, transcending conventional boundaries and fostering fresh possibilities for content creators, researchers, and movie enthusiasts alike. This article embarks on an enthralling odyssey through the captivating domain of movie character recognition, illuminating its transformative impact and the diverse applications that mold the future of visual storytelling.

II. LITERATURE REVIEW

The face recognition algorithm platform developed by Service and Kirby in 2009 followed the Eigenfaces approach. Sirovich and Kirby introduced software capable of identifying a subject's head, comparing its characteristics with known faces in a database, and then recognizing it. Another face recognition algorithm was created using independent component analysis by Bartlett, Movellan, and Sejnowski in 2011. Two algorithm methods for face recognition are under consideration: Principal Component Analysis (PCA) and Independent Component Analysis (ICA). PCA focuses on pixel data relationships, while ICA emphasizes high order statistics.

Researchers Maryam Mollaei and Mohammad Hossein Moattar proposed an enhanced ICA model with improved accuracy in face recognition. The previous Eigenface method faced challenges in identifying faces under varying illumination conditions. To address this issue, a new method called Fisher face using Linear Discriminant Analysis was developed in 1996. Viola and Jones proposed a face detection algorithm using HAAR Cascade and AdaBoost. Shervin Emami and other authors highlight the increasing interest in computer vision over the past decade.

Computer vision has evolved from an esoteric field to a widespread area of research, driven by advancements in computing power. Face detection and recognition are among the successful applications of image analysis and algorithm-based understanding. Computer vision research is not limited to computer science but also extends to neuroscientific and psychological studies, offering insights into brain function through advancements in computer image processing and understanding.

III. ANALYSIS OF PROBLEM

problem of movie character recognition from video and image using AI is super complex and has many layers. The main challenge lies in accurately identifying and recognizing characters from various movie scenes with different lighting, background, and angles at the character's face. Also, characters might have different hairstyles, makeup, or might even age throughout a movie, making it harder to recognize them. Another challenge is processing a huge amount of data because movies are long and have lots of frames. Processing each frame individually for character recognition is super hard and time-consuming. So, having an efficient algorithm that can fast process and analyze these frames is super important. Additionally, the problem is not just about recognizing faces.

The system should be able to track the recognized character across many frames, understand their role in the story, and even predict their actions or dialogues based on past data. And, there are also challenges related to the quality of the video or image. Low-resolution images or videos can really affect the performance of the AI system. Therefore, the system must be capable to handle all qualities of images and videos. Even though movie character recognition from video and image using AI offers a lot of opportunities for improving media consumption and production, it also faces some challenges that need to be taken care of for a successful implementation.

IV. PROPOSED WORK AND OBJECTIVES

The proposed work for "Movie Character Recognition From Video and Image Using AI involves the development of an AI system that can accurately recognize and identify characters from both video clips and still images. This system will leverage machine learning and deep learning algorithms to analyse visual data, extract features, and accurately identify movie characters. The system aims to not only recognize characters in a single frames, but also track their movements across multiple frames, thereby providing a comprehensive understanding of the character's role in the narrative. This technology has vast potential applications, including automating the process of video editing, assisting in targeted advertising based on character recognition, and providing valuable insights into audience preferences and behaviours! However implementing AI in movie character recognition presents its own set of challenges.

These include handling low-resolution images, dealing with varying lighting conditions, and accounting for changes in character appearance over time. Despite these challenges the AI system demonstrates a high degree of accuracy and efficiency. Overall, the development of an AI system for character recognition in both videos and images opens up new opportunities for automation and insights into audience behaviour. The challenges posed by varying image quality and lighting conditions only serve to enhance the complexity of the system, making it a formidable tool in the realm of character recognition.

V. METHODOLOGY

A. Data Collection:

The data collection process in image recognition involves gathering and preparing images to create datasets used for training artificial intelligence and machine learning algorithms specifically tailored for image analysis. This process is crucial for developing accurate computer vision systems. The collected image data must meet quality and quantity requirements to serve as effective training data for machine learning models. The data collection process typically includes steps such as identifying the need for specific data, selecting the appropriate method for data collection, and utilising various tools and techniques to gather relevant images. There are several methods for collecting image data, including using open datasets, creating custom datasets, or collaborating with third parties to acquire datasets. Image annotation is a key aspect of the data collection process, involving manually providing information about the ground truth within the images to train AI models effectively. This annotation process helps in teaching AI models to recognize patterns and objects within images accurately.

B. Data Pre-Processing:

The data pre-processing process in image recognition involves preparing and refining image data to enhance the performance of machine learning models used for image analysis. Essential pre-processing techniques for image recognition include:

Resizing Images: Resizing images to a uniform dimension is crucial to ensure consistency in input data for the model. This step helps in standardizing the images and reducing computational costs.

Normalisation: Normalising pixel values within images to a specific range, such as $[1]$ or $[-1, 1]$, can improve model stability, convergence, and overall performance. It reduces data variance and skewness, enhancing the model's ability to detect features and edges.

Data Augmentation: Data augmentation involves applying random transformations to images like flipping, rotating, or cropping. This technique helps increase the diversity of the dataset, improving model robustness and generalisation capabilities.

Image pre-processing plays a critical role in optimizing the quality of feature extraction and the accuracy of image analysis in computer vision systems. By carefully implementing these pre-processing techniques, researchers and developers can enhance the performance and efficiency of image recognition models.

C. Labelling Actor Data

Assigning unique labels to each actor's face for identity representation. Labeling in a dataset refers to the process of adding descriptive metadata or tags to raw data, which is essential for training machine learning models, particularly in supervised learning. These labels provide context and meaning to the data, enabling AI algorithms to understand and interpret the information accurately. In the context of machine learning, the dataset used for training and evaluating a model is often referred to as the "ground truth," emphasizing the importance of accurate labeling for model performance.

D. Training and testing

Training and testing in image recognition are crucial stages in the development and evaluation of machine learning models designed for image analysis.

Training:

Purpose: Training involves feeding a machine learning model with labeled image data to enable it to learn patterns, features, and relationships within the data. This process helps the model understand how to recognize and classify objects accurately.

Process: During training, the model iteratively adjusts its parameters based on the input data to minimize errors and improve its predictive accuracy. The model learns to differentiate between different classes of objects by analyzing the features extracted from the images.

Testing:

Purpose: Testing is essential for evaluating the model's performance on new, unseen data to assess its accuracy, generalization capabilities, and robustness.

Process: The model is presented with images that were not part of the training dataset to measure its ability to correctly classify and recognize objects. This step helps in determining how well the model can generalize to new data.

A. Local Binary Pattern Histogram (LBPH): The LBPH (Local Binary Pattern Histogram) algorithm is a face recognition algorithm that is used to recognize faces based on their unique characteristics. It is known for its performance and efficiency in identifying individuals from images or videos. The LBPH algorithm works by creating a histogram of local binary patterns, which are representations of the texture and structure of facial features. By analyzing these patterns and histograms, the algorithm can effectively match facial features extracted from images in a database to recognize and identify individuals. LBPH is particularly useful for its ability to recognize faces from various angles, including front-facing and side-facing views, making it a versatile and reliable tool for face recognition applications.

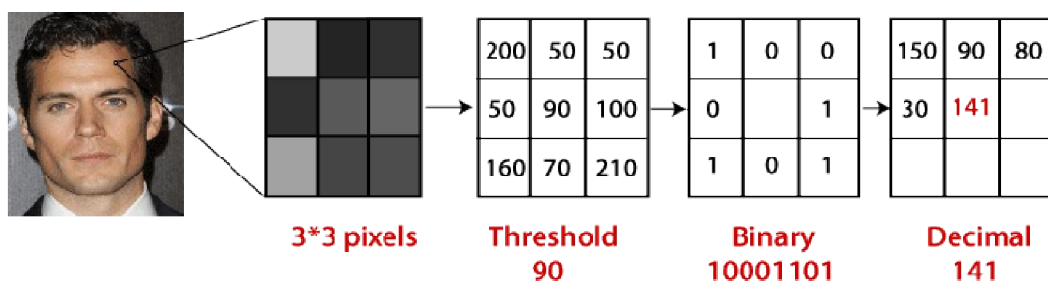


Fig. 1. Face Recognition of LBPH Algorithm.

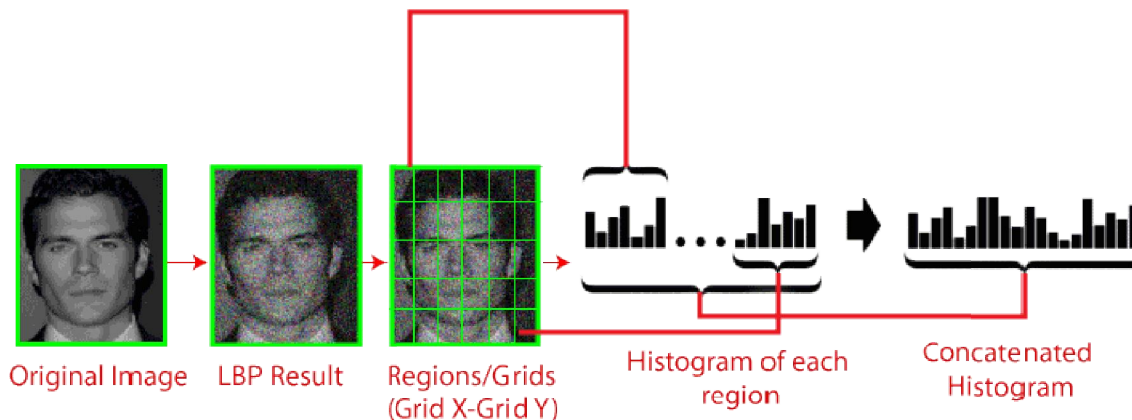


Fig. 2. Extracting the Histogram from the image.

B. Histogram of Oriented Gradients (HOG):

A popular feature descriptor for object detection in computer vision and image processing is the Histogram of Oriented Gradients (HOG) technique. In order to define local intensity gradients and their orientations—which are crucial for recognizing object shapes and structures—it focuses on capturing the distribution of gradient orientations inside a picture. The image is divided into small cells, and each cell's gradient orientation histogram is calculated. Next, the cells are grouped into blocks for normalization, and the normalized histograms are concatenated to generate the final HOG descriptor. With the exception of object orientation, the HOG algorithm's resistance to geometric and photometric alterations makes it useful for tasks like object recognition and pedestrian detection. In computer vision and image analysis, the Histogram of Oriented Gradients (HOG) is a well-liked feature descriptor technique. It analyzes the distribution of edge orientations within an object to describe its shape and appearance. The HOG method involves

computing the gradient magnitude and orientation for each pixel in an image and then dividing the image into small cells.

Process of Calculating the Histogram of Oriented Gradients (HOG) :

We now have a rudimentary understanding of what a HOG feature descriptor entails. It's time to explore the main concept of this article. Let's talk about the detailed steps involved in calculating HOG.

Examine the 180 x 280 pixel image below. Let's examine in more depth how the HOG features for this image will be produced:



Fig. 3. Prepare the Image.

Step 1 : Prepare the Data:

Most of you will be somewhat familiar with this phase. Any machine learning project must include preprocessing of the data, and dealing with photos is no exception. The image must be pre-processed in order to reduce the width to height ratio to 1:2. 64 x 128 pixels should be the ideal image size. This is due to the fact that in order to extract the features, we will be splitting the image into 8*8 and 16*16 patches. Having the given dimensions (64 x 128) will simplify all of our computations. Actually, the original paper utilized this precise value.

Returning to our example, let's assume that the size 64 x 128 is,

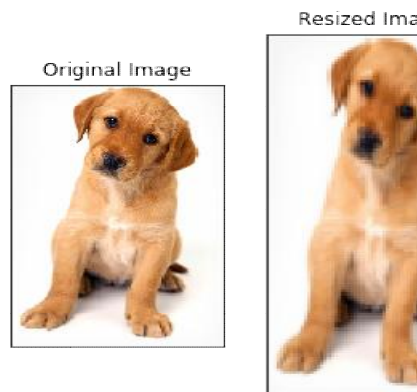


Fig. 4. Resizing Image by Pixelation.

Step 2: Calculating Slopes (course x and y):

In the following step is to calculate the slope for each pixel within the picture. Slopes are the little alter within the x and y headings. Here, I am planning to take a little fix from the picture and calculate the angles on that:nnhog_featuresWe will get the pixel values for this fix.

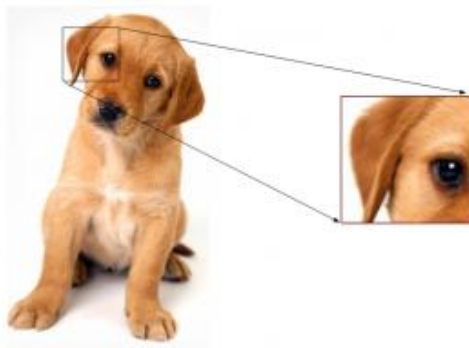


Fig. 5. Calculating Slopes.

Let's say we produce the underneath pixel framework for the given fix (the lattice appeared here is just utilized as an illustration and these are not the first pixel values for the given patch):nnhog_featuresSource: Applied Machine Learning Coursenn I have highlighted the pixel esteem 85. Presently, to determine the slope (or alter) within the x-direction, we got to subtract the esteem on the left from the pixel esteem on the proper. So also, to calculate the slope within the y-direction, we'll subtract the pixel esteem underneath from the pixel esteem over the chosen pixel.

121	10	78	96	125
48	152	68	125	111
145	78	85	89	65
154	214	56	200	66
214	87	45	102	45

Fig. 6. Pixel Orientation X,Y Coordinates.

Hence the resultant slopes within the x and y heading for this pixel are:nnChange in X direction(G_x) = $89 - 78 = 11$ nnChange in Y direction(G_y) = $68 - 56 = 8$ nnThis prepare will grant us two modern lattices – one putting away slopes within the x-direction and the other putting away slopes within the y heading. Typically similar to employing a Sobel Part of measure 1. The greatness would be higher when there's a sharp alter in concentrated, such as around the edges.nn We have calculated the gradients in both x and y heading independently. The same prepare is rehased for all the pixels within the picture. The following step would be to discover the greatness and introduction utilizing these values.

VI. BENEFITS OF THE SYSTEM

Content Cration: Content platforms can curate movie selections based on actors users enjoy, improving user engagement and satisfaction.

Marketing and Promotion: Studios and streaming platforms can target specific audiences by promoting content featuring popular actors.

Enhanced User Experience: Users can easily find information about their favorite actors, including their filmography, biographies, and related content.

Personalization: It allows for personalized recommendations and content suggestions based on the actors individuals prefer.

VII. LIMITATIONS

Complex Backgrounds and Illumination:

Movie scenes often have intricate backgrounds, varying lighting conditions, and dynamic illumination. These factors can interfere with accurate character recognition.

Processing Time and Database Management:

Manually classifying characters for each movie is time-consuming. Managing a database of character information can be cumbersome, especially when dealing with large movie catalogues.

Data Collection and Updates:

Manual data collection, updates, and deletions are necessary for maintaining accurate character profiles. Automating this process is essential to keep character databases up-to-date.

VIII. CONCLUSION

Movie character recognition from video and image using AI is a project that aims to automate the process of character classification in movies. The project involves collecting a dataset of movie characters and training a model that captures their facial features using image processing techniques and algorithm. The trained model can be used to track live video and automate the process of detecting characters. The software requirements for the project include Windows operating system, Python coding language, Anaconda development kit, TensorFlow, OpenCV libraries, and any movie dataset. The project can also be extended to predicting the success of movie scripts, selecting actors, and editing movies using AI algorithms.

In the dynamic realm of film analysis, character recognition from visual media presents both intrigue and complexity. The fusion of technical hurdles such as intricate backgrounds, varying illumination, and inconsistent grey-scale values creates a challenging landscape. Recognizing characters across different scales, adapting to long-form content, and maintaining temporal context are essential considerations. However, the scarcity of domain-specific training data and the need for manual character classification remain significant obstacles. As researchers continue to innovate, automating data management and ensuring up-to-date character databases will unlock the magic of identifying beloved cinematic figures on the silver screen.

REFERENCES

- [1] Musarrat Saber Nipun, Rejwan Bin Sulaiman, Amer Kareem, "Efficiency Comparison of AI classification algorithms for Image Detection and Recognition in Real-time" in School of Computer Science and Technology University of Bedfordshire Vicarage St, Luton, LU1 3JU.
- [2] A. Malik, M. Kuribayashi, S. M. Abdullahi and A. N. Khan, "DeepFake Detection for Human Face Images and Videos: A Survey", in *IEEE Access*, vol. 10, pp. 18757-18775, 2022.
- [3] Dhonushree Banerjee, Swapnil Ingole, Shritika Mandal, Aman Shrivastava, Nischay Dekate, Smit Patil, Preete Karmore "Face Recognition and Identification App using Python" in International Research Journal of Modernization in Engineering Technology and Science Volume. 04, issue.04, May-2022.
- [4] Manish Kumar, Siddharth Sharma, Divy Chaudhary, S. Prakash, "Image Recognition Using Artificial Intelligence" 2021 International Conference on Advance Computing and Innovative Technologies in Engineering (ICACITE) | 978-1-7281-7741-0/20/\$31.00 ©2021 IEEE.
- [5] Souvik Das, Soumyadeep Sett, Subhojyoti Saha, Soumyadeep Saha, Soumyadeep Roychoudhury, Koushik Pal "A Novel Face Detection Technique Using OpenCV" in International Journal of Research in Engineering, Science and Management Volume 4, Issue 7, July 2021.

- [6] Nur Ateqah Binti Mat Kasim, Nur Hidayah Binti Abd Rahman, Zaidah Ibrahim, Nur Nabilah Abu Mangshor "Celebrity Face Recognition using Deep Learning" in Indonesian Journal of Electrical Engineering and Computer Science Volume. 12, No. 2, November 2018.
- [7] Cast2Face: Assigning Character Names Onto Faces in Movie With Actor-Character Correspondence. IEEE Transactions on Circuits and Systems for Video Technology (*Volume: 26, Issue: 12, December 2016*).
- [8] Robust Character Labelling in Movie Videos: Data Resources and Self-Supervised Feature Adaptation IEEE Transactions on Multimedia (*Volume: 24, Issue: 09 July 2021*).
- [9] Robust Identification of Movie Character using Face-Name Graph Matching. NCRTICE – 2013 (Volume 1 – Issue 04). **volume & Issue : NCRTICE – 2013 (Volume 1 – Issue 04) Published (First Online): 30-07-2018.**
- [10] A Novel Method for Movie Character Identification based on Graph Matching. Publish on *IJISSET* - International Journal of Innovative Science, Engineering & Technology, (*Vol. 1 Issue 4, June 2014*).