

Age and Gender Classification using Convolutional Neural Network

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Abstract - Due to its numerous applications in various facial analysis challenges, automatic prediction of age and gender from face images has received a lot of interest recently. The available models, however, are still below the needed accuracy level, which is required for the usage of these models in real-world applications due to the significant intra-class variance of face images (such as difference in lighting, position, scale, and opacity). In this study, we offer a classification model that can accurately identify the gender and age range of facial images using convolutional neural networks

impact on facial recognition systems. This concept is crucial to the new fields of computer vision research that will be investigated.

1.1 Overview with Problem Statement

With the expansion of real-world applications has expanded day-to-day living, researchers have shown more interest in the soft biometrics sector to close the communication gaps between humans and machines. Age, gender, ethnicity, height, face dimensions, and other soft biometrics are included.

Key Words: Convolutional Neural Network, Machine learning, Age classification, Gender Detection.

1. INTRODUCTION

Age, gender, mood, and other characteristics can all be inferred from a person's face. Numerous dynamic aspects that alter over time, such as age, hairstyles, expressions, etc., have an impact on it. Age and gender are regarded as crucial biometric characteristics for identifying humans. For the purpose of human identification and verification, biometric recognition gathers data on a person's physiological and behavioral traits (security models). Age, gender, ethnicity, height, and face measurements are examples of soft biometrics. Hard biometrics are measurements of the body (physical, behavioral, and biological). To speed up data traversal or to categorize unlabeled subjects for different gender and age groups, soft-biometric features can be retrieved, such as skin color, hair color, the distance between the eye and nose, facial shape, etc.

Machines cannot classify patterns as effectively and powerfully as the human brain can. Therefore, our goal is to use technology to imitate the ability of the human brain to determine a person's age and gender. This problem can be solved by developing an application for age and gender detection that can accurately determine a person's age and gender. The age and gender of the person are determined by using their human face as the input. The person's age and gender are the output.

1.2 Challenges and Applications

Age estimation and facial gender classification provide numerous difficulties. Two classes that can be either male or female are subject to gender prediction. While a machine cannot easily classify gender, a human can. Numerous methods and models have been criticized for gender classification based on extra data from hairstyles, body shape, attire, and facial traits. As of now, it is not possible to determine actual age while estimating age. In order to determine age from facial photos, age grouping is still used. Additionally, there aren't enough high-quality datasets for estimating age and classifying gender to support extensive research.

Partial occlusions and poor-quality photos are the most frequent issues when it comes to face detection or age/gender classification. Due to the model's limited data set and difficulty in making predictions, these directly affect the outcome findings. When a human is making the forecast, the same rules still apply. It is more difficult for a human to comprehend what is being viewed in a low-quality image and, as a result, to predict the future.

Age recognition is important in police investigations and intelligence departments because it aids in locating the actual suspect based on his age. They may receive a filtered result of that person who has committed a criminal act or any other activity. When it comes to software, the actual and predicted ages are roughly the same, indicating its dependability, and this dependability serves as a trust factor for many other useful operations in daily life.

We propose a scheme in this paper to bridge the gap between automatic face recognition and age and gender prediction. When there is a large-scale improvement in face recognition, a link between face recognition and Convolution Neural Network (CNN) is proposed, and by studying it further, we created a system in which a limited number of face data sets are used to accurately predict age and gender.

2. SYSTEM ANALYSIS AND DESIGN

2.1 Machine learning techniques:

The various machine learning approaches utilized for categorization and implementation in this system are covered in this section. The optimal model for the system has been determined by comparing the results of all of these models.

Deep learning techniques for computer vision:

In recent times, deep learning techniques proved to be a big success in the Computer Vision discipline. Deep learning enables multi-layered computing models to determine and interpret data with multiple abstraction levels and imitates how the information is perceived and translated by the brain. So, it implicitly captures large-scale data structures. Deep learning has outperformed many of the previously existing techniques. Deep learning has enabled various techniques in computer vision to increase accuracy and efficiency, which includes object detection, action recognition, human emotion recognition, and others. Further, types of deep learning techniques will be discussed with their comprehensive details.

Convolutional Neural Networks(CNN):

Convolutional Network Network (CNN) was first proposed in 1962, and it consists of the following layers.

- **Convolutional Layer:** As it is known that CNN utilizes various kernels, so the convolutional operation of the layer increased the learning time of the developed model.

- **Pooling Layers:** It reduces the spatial dimensions of the input volume for the next convolutional layer. It only affects the length and height, and not the depth.

- **Fully connected Layers:** Several connected neural network layers have been used to perform any high-level reasoning.

There are some difficulties that might also arise with CNN, such as overfitting, which is due to the CNN training of a large number of parameters. The solution to it is the pretraining of parameters, which accelerates the learning process of the model as well as improves the generalising capability of the model. In short, CNN has outperformed the usual and traditional machine learning algorithms.

3. IMPLEMENTATION

Python programming language, as well as numerous computer vision and machine learning packages and libraries, will be used throughout the research implementation. The primary goal of the project will be to create a Python-based, Tensorflow-compatible high-level convolutional neural network API. Python, in and of itself, is a high-level programming language. Python is open-source, object-oriented, and has simple readability and coding. Because it contains so many packages, it is widely used in Big Data, Machine Learning, and Computer Vision. Furthermore, Python was chosen for this experiment because it is free to use, compatible with the Windows operating system, and contains all of the necessary libraries for face recognition, emotion detection, and gender classification.

Face identification consists of three steps. Detect which part of an image is the face, then train our classifier for that dataset of images, and finally, predict the face. OpenCV, a Python open-source library for computer vision, will be used for this. A Haar cascade frontal face default classifier will be used for face detection, which is a pretrained model that is freely available online. A default Haar classifier will be used for gender and age prediction. The first step will be to detect the face in the image using some test images. All of the images used in the training are freely available online and open-source. The entire experiment will be implemented in Keras with Tensorflow as the backend. The entire Convolutional Neural Network will be built on these, as well as the OpenCV computer vision library. Keras can determine whether the model's current epoch outperformed the previously saved epoch. In this case, the best model weights will be saved in a file that will allow the weights to be loaded directly without retraining if the model needs to be used in another

situation. Keras has modularity, extensibility, and Python nativeness when compared to other similar libraries.

Gender identification is the process of determining whether a person is male or female. It is a binary model because there are only two options: male or female. The primary libraries for Gender Classification are OpenCV and Keras.

4. RESULTS

By analyzing human facial features in real-time, this model can predict ages ranging from 0 to 80 and classify genders as Male or Female. Because the model predicts age in real-time, it is subject to change with each webcam frame.

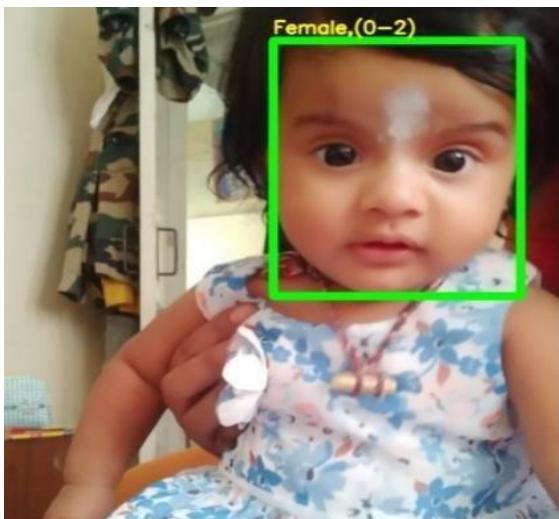


Figure 1: Age and Gender prediction of age range 0-2

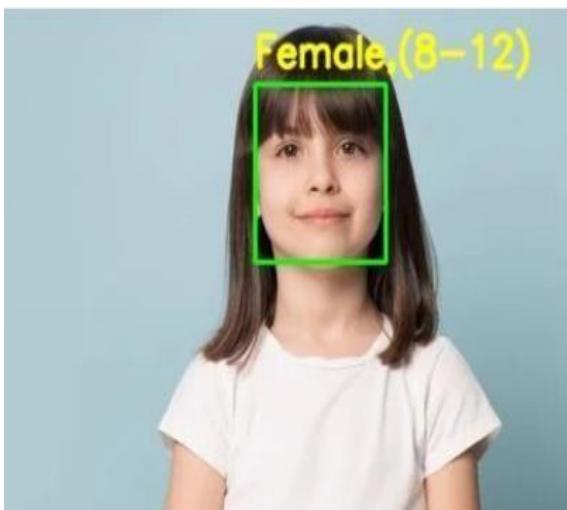


Figure 1: Age and Gender prediction of age range 8-12

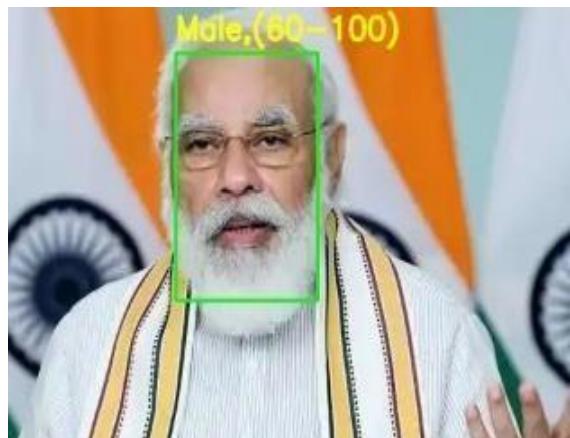


Figure 1: Age and Gender prediction of age range 60-100

Figure 1,2 and 3 shows the prediction results of the developed model.

5. CONCLUSIONS

A thorough literature review of various Machine Learning and Deep Learning techniques is used to discuss all of the techniques and methods that have already been implemented in this field. Facial images have become increasingly important in recent decades, owing primarily to their promising real-world applications in a variety of emerging fields. The proposed system is capable of classifying gender as either male or female and predicting age from 0 to 80.

The model's accuracy is calculated separately to provide a more accurate comparison and interpretation of the study. The proposed architecture was built methodically to improve accuracy and reduce the number of parameters. Gender classification and age prediction have been manually tested, and the results have been astounding. Because gender classification is considered a binary problem in this study, it has proven to be very efficient with the use of Keras and achieves an overall accuracy of about 90%. Age prediction is affected by a variety of external factors, including lighting effects, facial expressions, and skin tones, but it also produces impressive results.

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REFERENCES

- [1] Asad Mustafa, Kevin Meehan, Gender Classification and Age Prediction using CNN and ResNet in Real-Time, Department of Computing Letterkenny Institute of Technology Letterkenny, Ireland, ORCID: 0000-0001- 5447-5878.
- [2] Thakshila R, Kalansuriya, Anuja T, Facial Image Classification Based on Age and Gender, Dharmaratne University of Colombo School of Computing, University of Colombo No 35, Reid Avenue, Colombo 7.
- [3] Vijay Prakash Dwivedi, Deepak Kumar Singh, Saurav Jha, Ranvijay, Gender Classification of Blog Authors: With Feature Engineering and Deep Learning using LSTM Networks, Computer Science and Engineering Department MNNIT Allahabad UP, India.
- [4] Mohammed Kamel Benkaddour, Sara Lahlali, Maroua Trabelsi, Human Age and Gender Classification using Convolutional Neural Network, University Kasdi Marbah, Department of Computer Science and Information Technology, FNTIC Faculty, Ouargla, Algeria.
- [5] Hiromi Kondo, Fumiyo N. Kondo, Convolutional Neural Networks on Multichannel Time Series of Smartphone Applications for Gender or Age Range Classification, Toyota Motor Corporation Customer First Promotion Group C&A Operations Div. Aichi, Japan, University of Tsukuba Division of Policy and Planning Sciences Faculty of Engineering, Information and Systems Ibaraki, Japan.
- [6] Avishek Garain , (Member, Ieee), Biswarup Ray , Pawan Kumar Singh , (Member, Ieee), Ali Ahmadian , (Member, Ieee), Norazak Senu , And Ram Sarkar , (Senior Member, IEEE), A Deep Learning Model for Classification of Age and Gender From Facial Images, Department of Computer Science and Engineering, Jadavpur University, Kolkata. 700032, India, Department of Information Technology, Jadavpur University, Kolkata 700106, India , Institute of Industry Revolution 4.0, The National University of Malaysia (UKM), Selangor 43600, Malaysia, Institute for Mathematical Research, University Putra Malaysia, Serdang 43400, Malaysia
- [7] Insha Rafique, Awais Hamid, Sheraz Naseer, MuhammadAsad, Muhammad Awais, Talha Yasir, Age and Gender Prediction using Deep Convolutional Neural Networks, Department of Software Engineering University of Management and Technology Lahore, Pakistan.
- [8] Md. Nahidul Islam Opu, Tanha Kabir Koly, Annesha Das and Ashim Dey, A Lightweight Deep Convolutional Neural Network Model for Real-Time Age and Gender Prediction, Department of Computer Science & Engineering Chittagong University of Engineering and Technology Chittagong-4349, Bangladesh
- [9] Azliza Mohd Ali, Plamen Angelov, Gender and Age Classification of Human Faces for Automatic Detection of Anomalous Human Behaviour, 2017 3rd IEEE International Conference on Cybernetics (CYBCONF), DOI: 10.1109/CYBConf.2017.7985780
- [10] Seok Hee Lee, Hyuk Jin Kwon, Hyung Il Koo, Nam Ik Cho, Age and gender classification using wide convolutional neural network and Gabor filter, 2018 International Workshop on Advanced Image Technology (IWAIT), DOI: 10.1109/IWAIT.2018.8369721.
- [11] Xuan Liu, Junbao Li, Cong Hu, Jeng-Shyang Pan, Deep convolutional neural networks-based age and gender classification with facial images, 2017 First International Conference on Electronics Instrumentation & Information Systems (EIIS), DOI: 10.1109/EIIS.2017.8298719.
- [12] Mohammed Kamel Benkaddour, Sara Lahlali, Maroua Trabelsi, Human Age and Gender Classification using Convolutional Neural Network, 2020 2nd International Workshop on Human-Centric Smart Environments for Health and Well-being (IHSH).
- [13] Gil Levi, Tal Hassner, Age and gender classification using convolutional neural networks, 2015 IEEE Conference on Computer Vision and Pattern Recognition Workshops DIO:10.1109/CVPRW.2015.7301352.

[14] Avishek Garain, Biswarup Ray, Pawan Kumar Singh, Ali Ahmadian, Norazak Senu, Ram Sarkar, A Deep Learning Model for Classification of Age and Gender From Facial Images, DOI: 10.1109/ACCESS.2021.3085971.

[15] Syed Taskeen Rahman, Asiful Arefeen, Shashoto Sharif Mridul, Asir Intisar Khan, Samia Subrina, Human Age and Gender Estimation using Facial Image Processing, 2020 IEEE Region 10 Symposium (TENSYMP), DOI: 10.1109/TENSYMP50017.2020.9230933.

[16] Sandeep Kumar, Sukhwinder Singh, Jagdish Kumar, A study on face recognition techniques with age and gender classification, Published in: 2017 International Conference on Computing, Communication and Automation DOI: 10.1109/CCAA.2017.8229960.

[17] Sandeep Kumar, Sukhwinder Singh, Jagdish Kumar, A study on face recognition techniques with age and gender classification, 2017 International Conference on Computing, Communication and Automation. DOI: 10.1109/CCAA.2017.8229960.

[18] Jun Beom Ko, Wonjune Lee, Sung Eun Choi, Jahie Kim, A gender classification method using age information, 2014 International Conference on Electronics, Information and Communication, DOI: 10.1109/ELINFOCOM.2014.6914362.

[19] Min Hu, Yaona Zheng, Fuji Ren, He Jiang, Age estimation and gender classification of facial images based on Local Directional Pattern, 2014 IEEE 3rd International Conference on Cloud Computing and Intelligence Systems, DOI: 10.1109/CCIS.2014.7175711.

[20] Chi Xu, Yasushi Makihara; Ruochen Liao, Hirotaka Niitsuma; Xiang Li, Yasushi Yagi, Jianfeng Lu, Real-Time Gait-Based Age Estimation and Gender Classification from a Single Image, 2021 IEEE Winter Conference on Applications of Computer Vision (WACV).

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