

Early Autism Spectrum Disorder Detection

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Abstract: *Autism spectrum disorder (ASD) is a type of mental illness that can be identified through social media data and biomedical imaging. Autism spectrum disorder (ASD) is a neurological disorder that correlates with brain growth and subsequently affects physical expression. ASD children have distinct facial features that significantly differentiate them from typically developing (TD) children. The novelty of the proposed research is to develop a system based on the detection of autism spectrum disorder in social media and facial recognition. Deep learning techniques can be used to recognize such landmarks, but they require precise techniques to extract and generate the correct facial patterns.*

Keywords: Deep Learning, CNN, ASD, Tensorflow

I. INTRODUCTION

Autism Spectrum Disorder (ASD) refers to the following groups: A complex neurodevelopmental disorder of the brain. B. Autism, Childhood Disorders, Asperger's Syndrome. As the word "spectrum" suggests, symptoms and severity vary widely. The disease is now included in the International Statistics on Classification of Diseases and Associated Health Problems. It is found in the Mental and Behavioral Disorders category of the Pervasive Developmental Disorders category. Symptoms of ASD often appear in the first year of life [3-6]. It may also include lack of eye contact, lack of reaction, verbal abuse and indifference to caregivers. Few children develop normally in the first year of life, but by 18 to 24 months of age, symptoms of autism, such as restricted and repetitive behaviours, a narrow range of interests and activities, and poor language skills, can be seen. Appear. These disorders also affect a person's cognition. Five-year-olds can suddenly become introverted or aggressive when interacting with others, making it difficult for them to speak and communicate socially. ASD begins in childhood but tends to persist through adolescence and into adulthood. A convolutional neural network (CNN) algorithm is used on the training data to extract the components of human facial expressions. It has been proposed to use such facial expression recognition algorithms for many neurological diseases.

II. LITERATURE SURVEY

Vaishali R, Sasikala R et al. [1] proposed a method to identify autism using optimal behaviour sets. In this work, his ASD diagnostic dataset containing 21 functions taken from the UCI machine learning repository was experimented with. Experimental alternative hypothesis Machine learning models claim that it is possible to achieve higher classification accuracy with minimal features Subset. Using the Swarm intelligence-based single-purpose binary Firefly feature selection wrapper, it is found that 10 features among 21 features of the ASD dataset are sufficient to distinguish between ASD and non-ASD patients. The results obtained with this approach justify the hypothesis, with average accuracies in the range of 92.12%. 97.95% (approximately equal to the average accuracy of the entire set for the best feature subset) ASD diagnostic record

M. S. Mythili, A. R. Mohamed Shanavas [3] conducted a study on ASD using classification techniques. The primary goal of this study was to establish autism problems and degrees of autism. In this neural network, SVM and Analyze student behaviour and social interactions using fuzzy techniques with the WEKA tool.

Kosmicki1, V. Sochat, M. Duda, D.P. Wall et al. [4] adopt minimum set search method functional description for autism detection. The authors used a machine learning approach to assess clinical assessment by ASD. ADOS was performed on a subset of children's behaviours based on the autism spectrum. Addos Equipped with 4 modules. In this work, eight different machine learning algorithms were used and performed step by step. Reverse the identification of the features

of the 4540 scoresheets. Uses 9 of the module's 28 movements identified ASD risk with an overall accuracy of 98.27% using 2 and 12 of the 28 behaviours in module 3, with an accuracy of 97.66%.

Li B, Sharma A, Meng J, Purushwalkam S, Gowen E (2017) et al. [5] uses a machine learning classifier to identification of adults with autism by imitation methods. The purpose of this study was to explore the underlying issues involved, regarding identification of test conditions and kinematic parameters. The dataset contains a series of hand gestures and was used to extract 40 kinematic constraints from 08 mimic conditions, extracted by using machine learning techniques. This study shows that the machine can be used for small samples. Sensitivity rate achieved by RIPPER, characterised by Va (87.30%), CHI (80.95%), IG (80.95%), Correlation (84.13%), CFS in the AQ youth dataset (84.13%), it was 'no function selection' (80.00%).

III. ARCHITECTURE

Convolutional Neural Network employ user provided input images to detect autism from the picture. In the proposed architecture, Dataset is collected from the Kaggle Platform. They then extract key features from the images, and forecast the features using pooling functions. Here, the softmax function is used to predict the output accurately and for overcoming the over fitting problem.

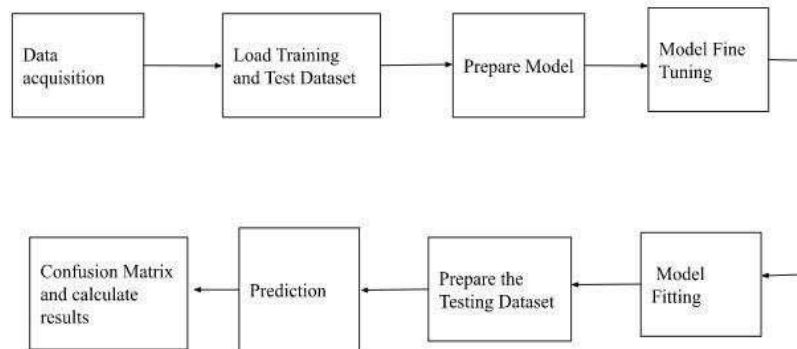


Fig. 1. System Architecture

IV. METHODOLOGY

Facial features can be used to determine whether a child has autism or is normal. Models extract important facial features from images. One of the advantages of deep learning algorithms is the ability to extract very small details from an image that cannot be seen with the naked eye. An overview of our ASD case-finding methods is briefly described below.

Dataset

This study analysed facial images of autistic children and normal children obtained from the publicly available online platform Kaggle. The data consisted of 2,940 facial images, half of which were of autistic children and the other half of non-autistic children. This dataset is collected from online sources such as websites and Facebook pages with an interest in autism.

Preprocessing

The purpose of data processing was to clean and deconvolution the images. Because Piosenka collected data from online resources, it had to be pre-processed before it was used to train a deep learning model. The creator of the dataset automatically extracted the face from the original image. The dataset was then divided into 2540 images for training, 100 images for validation, and 300 images for testing. The normalisation method was used to scale; the dataset changed all image parameters from pixel values [0, 255] to [0, 1].

Convolutional Neural Network (CNN)

AI has been remarkably developed to assist humans in their daily life, for example, through medical applications, which are based on a branch of AI called “computer vision.” Hence, the CNN algorithm has contributed to the detection of diseases and to behavioural and psychological analysis.

Basic Components of the CNN Model.

The convolutional neural network(CNN) is one of the most notorious deep literacy algorithms. It takes the input image and assigns significance to learnable weights and impulses in order to fetch the class of the image. The neuron can be said to be a simulation of the communication pattern of the neurons of Computational Intelligence and Neuroscience the mortal brain through the connection and communication between cells. In this section, we will explain the introductory factors of the CNN model: the input subcaste, convolutional subcaste, cranking function, pooling subcaste, completely connected subcaste, and affair vaticination.

Convolutional Layer with a Pooling Layer

The input of the convolutional subcaste is an image as a matrix of pixel values. The ideal of the convolutional subcaste is to reduce the images into a form that can be fluently reused without losing their important features that will help to describe autism. The first subcaste of the CNN model is responsible for rooting low position features similar to edges and colour. The figure of the CNN model allowed us to add further layers to it to enable it to due to the large number of parameters outputted from the complication subcaste, which may significantly protract the computation operations of the matrices, the number of weights was reduced by using one of the following two ways: maximum pooling or average pooling. Max pooling is grounded on the outside values in each window in the stride, while average pooling is grounded on the mean value of each window in the stride. In this study, the model was grounded on maximum pooling. The slide window of the kernel excerpts the features from the input image and converts the image into a matrix, after which the number of the parameters is mathematically reduced through maximum pooling and average pooling.

Fully Connected Layer and Activation Function

The completely connected(FC) subcaste is a nonlinear combination of the high- position features that admit the input from the retired layers and are represented as labourers. In the FC subcaste, the input image is represented as a column vector. The training of the model has two paths: the forward neural network and backpropagation. The forward neural network feeds form a flattened affair subcaste. In the backpropagation, the neural network minimises the loss crimes and learns further features by applying the number of the training duplications. The utmost deep literacy models show high performance while adding the number of the retired layers and the training duplications, which allows the neural network to price the low- position features deeply. The softmax classifier receives the parameters from the FC subcaste and calculates the parcels to prognosticate the affair. A Softmax affair of 0 means the image belongs to class 0 and a softmax affair of 1 means the image belongs to class 1 In this study, class 0 is the autism class and class 1 is the normal class

Deep Learning Models

The paper is based on Xception models for autism detection using facial feature images: Xception.

1) Xception Model

The Xception model was trained on the ImageNet dataset(2, 3) for the image recognition and bracket task. Xception is a deep CNN that provides new commencement layers. The commencement layers are erected from depthwise complication layers and are followed by a pointwise complication subcaste. Transfer literacy has two generalities: birth and fine- tuning. In this study, the point birth system used the pretraining model, which was trained on the standard dataset to prize the point from the new dataset and to remove the top layers of the model. The new top layers were added to the model for custom brackets grounded on the number of classes. Fine- tuning has been used to acclimatise the general features to a given class to avoid overwriting. The Xception armature used the features maps, followed by a global maximum pooling subcaste and two thick layers, 128 and 64, independently, with a rule activation function. Also, the affine of the thick layers was passed to the flatten subcaste, which took the input as a point chart and the affine as a vector. Batch

normalisation was used to enhance the affair by avoiding overfitting. Keras supported the early- stopping system, which stopped the training when the confirmation loss of the model didn't ameliorate. The last subcaste for the affair vaticination used the Softmax function.

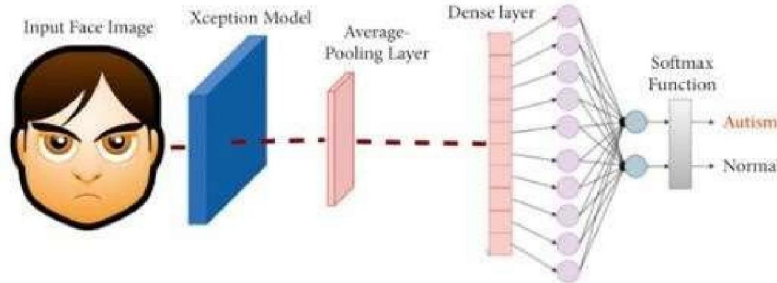


Fig. 2. Xception Model

Figures

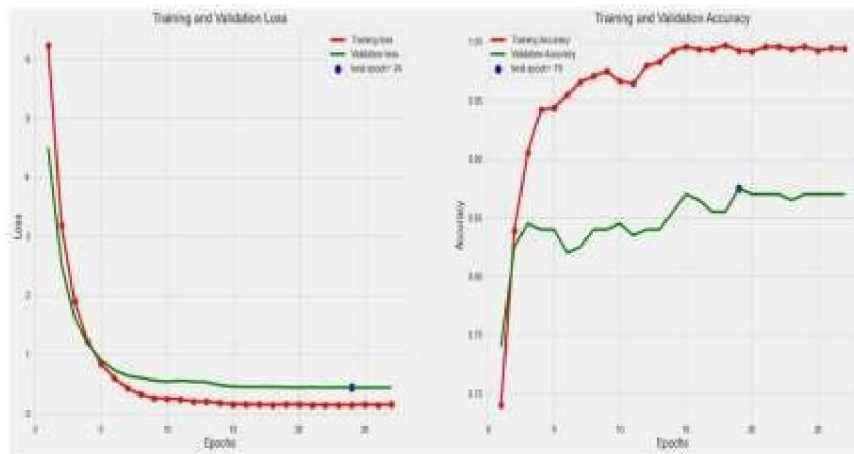


Fig. 3. Plotting Training Data

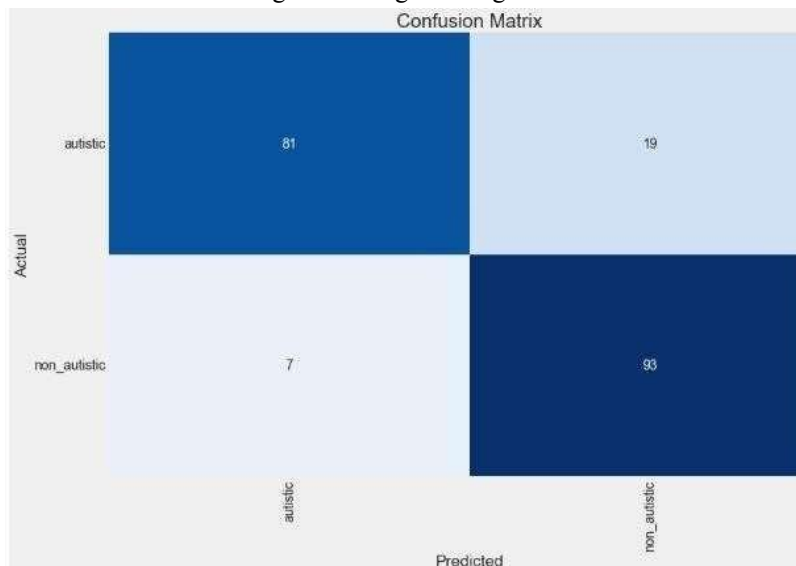


Fig. 4. Confusion Matrix

Fig. 2. Shows the plotting of training data whereas, Fig. 4. shows the Confusion Matrix.

V. CONCLUSION AND FUTURE SCOPE

In this work, discovery of Autism Spectrum Disorder was tried using colourful machine literacy and deep literacy ways. The CNN grounded model was suitable to achieve loftiest delicacy result than all the other considered model structure ways, these results explosively suggest that a CNN grounded model can be enforced for discovery of Autism Spectrum Disorder. The main limitation of this exploration is the small datasets. In future work, we aim to collect large datasets and work with deep literacy styles that integrate point assessment and bracket together for bettered performance. Also, we would like to assay brain signals(e.g., EEG) to relatethis with AQ grounded study in order to develop a more robust ASD discovery algorithm. The results of the model bracket showed us the possibility of using similar models grounded on deep literacy and computer vision as automatic tools for specialists and families to directly and more snappily diagnose autism

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