Capstone Project-5

Deep learning and MLE Project:

Live Class Monitoring System(Face Emotion Recognition)

Team Observant Force Group

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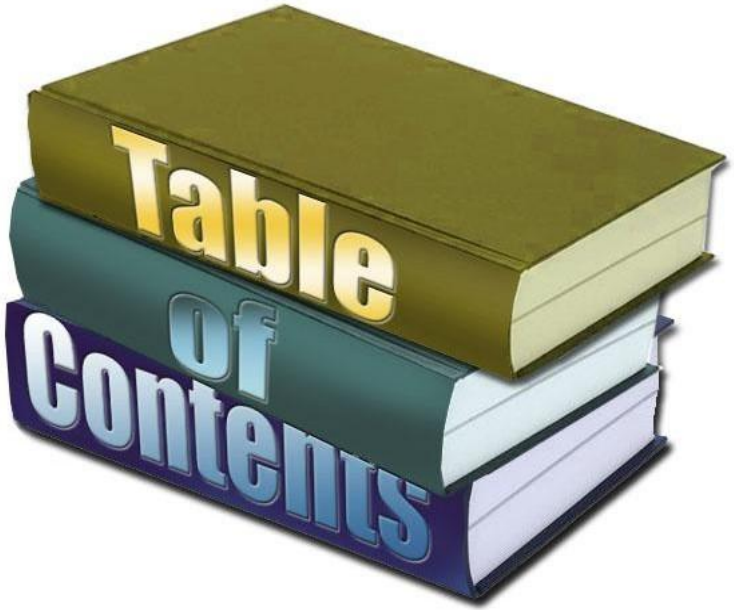


Introduction

The Indian education landscape has been undergoing rapid changes for the past 10 years owing to the advancement of web- based learning services, specifically, eLearning platforms.

Global E-learning is estimated to witness an 8X over the next 5 years to reach USD 2B in 2021. India is expected to grow with a CAGR of 44% crossing the 10M users mark in 2021. Although the market is growing on a rapid scale, there are major challenges associated with digital learning when compared with brick and mortar classrooms. One of many challenges is how to ensure quality learning for students. Digital platforms might overpower physical classrooms in terms of content quality but when it comes to understanding whether students are able to grasp the content in a live class scenario is yet an open-end challenge. In a physical classroom during a lecturing teacher can see the faces and assess the emotion of the class and tune their lecture accordingly, whether he is going fast or slow. He can identify students who need special attention. Digital classrooms are conducted via video telephony software program (ex-Zoom) where it's not possible for medium scale class (25-50) to see all students and access the mood. Because of this drawback, students are not focusing on content due to lack of surveillance.

While digital platforms have limitations in terms of physical surveillance but it comes with the power of data and machines which can work for you. It provides data in the form of video, audio, and texts which can be analyzed using deep learning algorithms. Deep learning backed system not only solves the surveillance issue, but it also removes the human bias from the system, and all information is no longer in the teacher's brain rather translated in numbers that can be analyzed and tracked.

Problem Statements

We will solve the above-mentioned challenge by applying deep learning algorithms to live video data.The solution to this problem is by recognizing facial emotions.

What is Face Emotion Recognition?

Facial emotion recognition is the process of detecting human emotions from facial expressions. The human brain recognizes emotions automatically, and software has now been developed that can recognize emotions as well.

Data Summary:

- I have built a deep learning model which detects the real time emotions of students through a webcam so that teachers can understand if students are able to grasp the topic according to students' expressions or emotions and then deploy the model. The model is trained on the FER-2013 dataset .
- This dataset consists of 35887 grayscale, 48x48 sized face images with seven emotions - angry, disgusted, fearful, happy, neutral, sad and surprised.

- Here is the dataset link:- https://www.kaggle.com/msambare/fer2013

Label	Emotion	Number of images for Training	Number of images for Testing
0	angry	3995	958
1	disgust	436	111
2	fear	4097	1024
3	happy	7215	1774
4	sad	4830	1247
5	surprised	3171	831
6	neutral	4965	1233

Model Creation

1)Using DeepFace

deepface

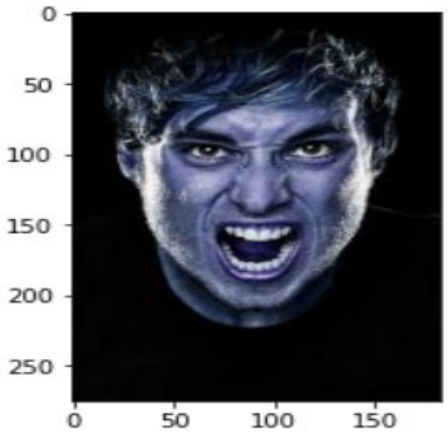
DeepFace is a lightweight face recognition and facial attribute analysis framework for Python.

Face verification

Face recognition

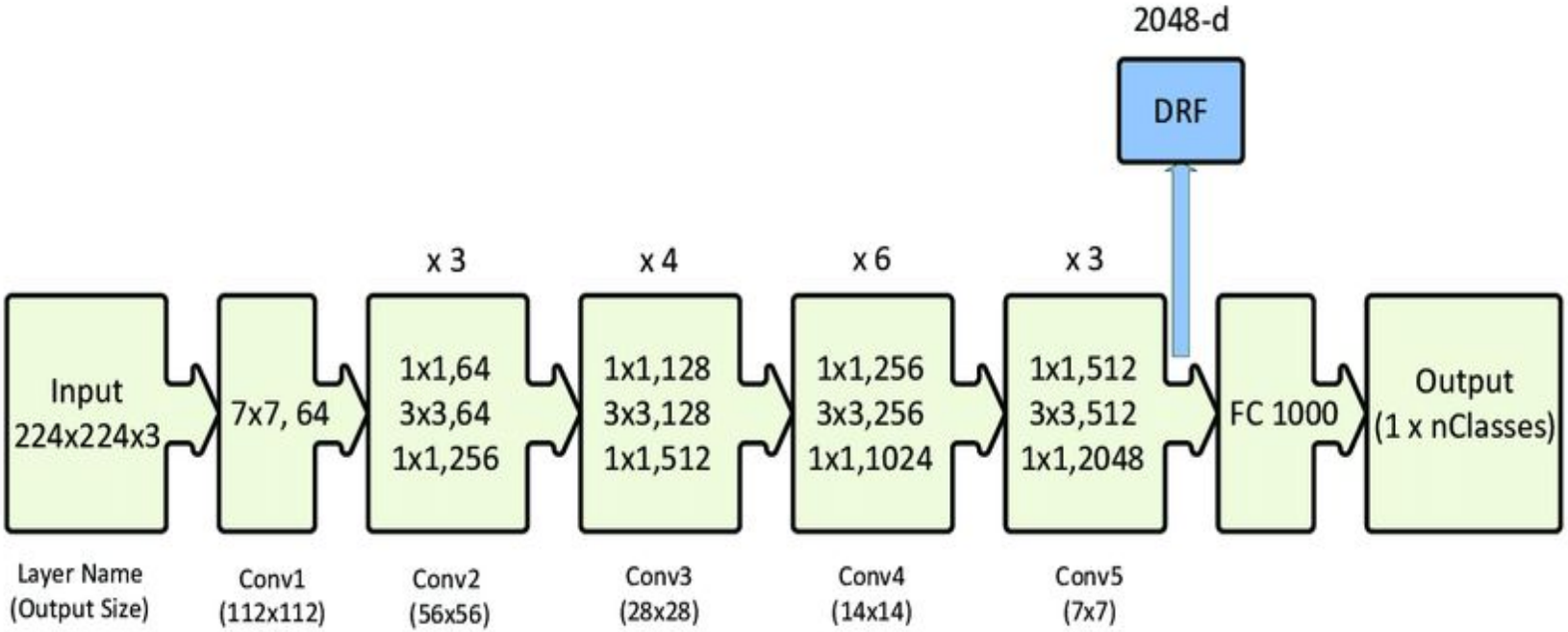
Facial analysis

DeepFace Prediction:

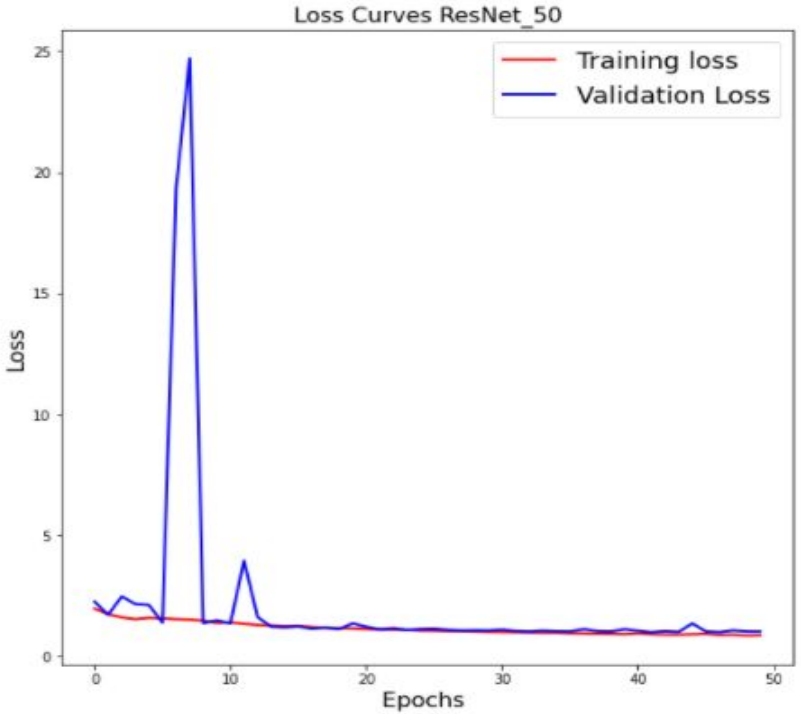


We imported above image which looks angry but our model give us “27 years old white **fear** Man” this result. To get better results we decided to train our own model.

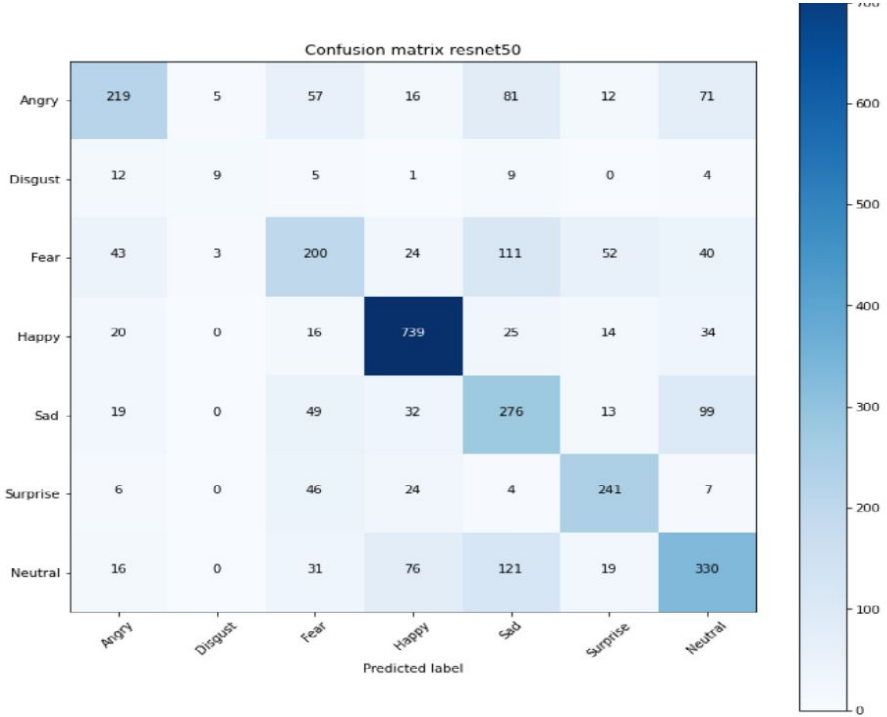
2) Resnet50


2048-d
DRF
x 3
x 4
x 6
x 3
Input
224x224x3
7x7, 64
1x1,64
3x3,64
1x1,256
1x1,128
3x3,128
1x1,512
1x1,256
3x3,256
1x1,1024
1x1,512
3x3,512
1x1,2048
FC 1000
Output
(1 x nClasses)
Layer Name
(Output Size)
Conv1
(112x112)
Conv2
(56x56)
Conv3
(28x28)
Conv4
(14x14)
Conv5
(7x7)

Accuracy and loss plot:



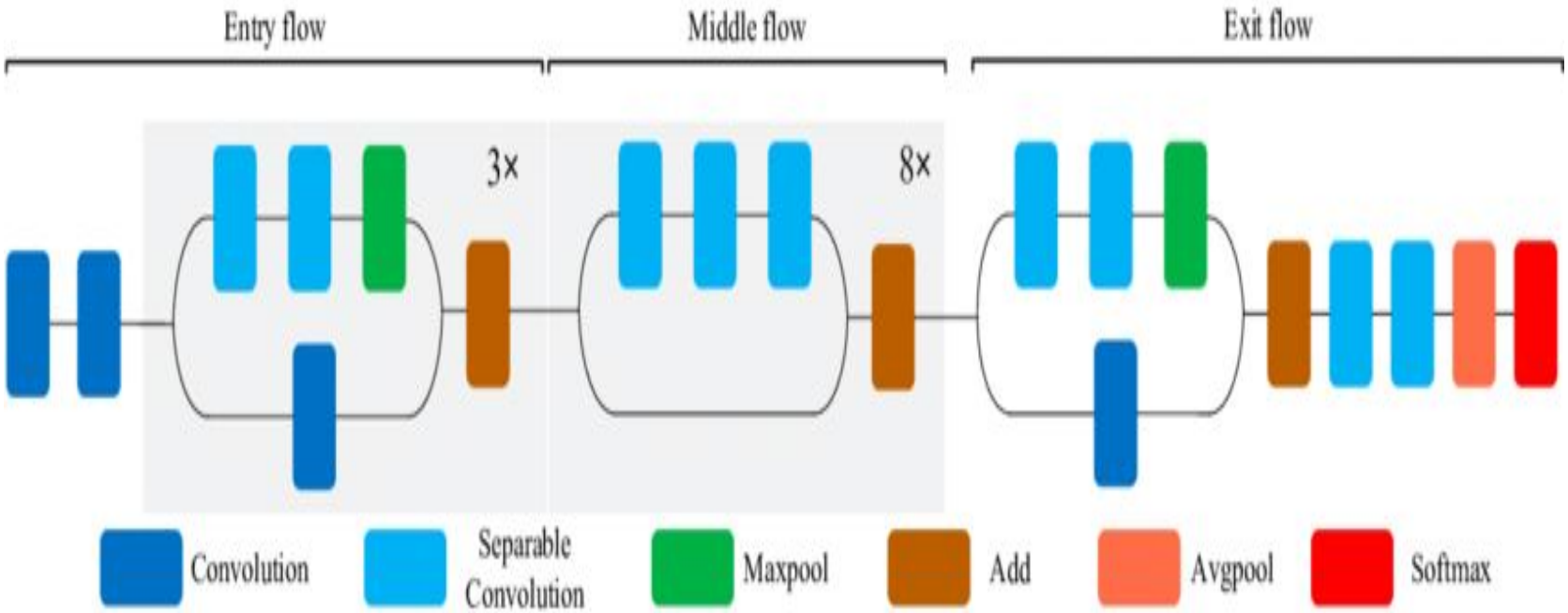
Confusion Matrix



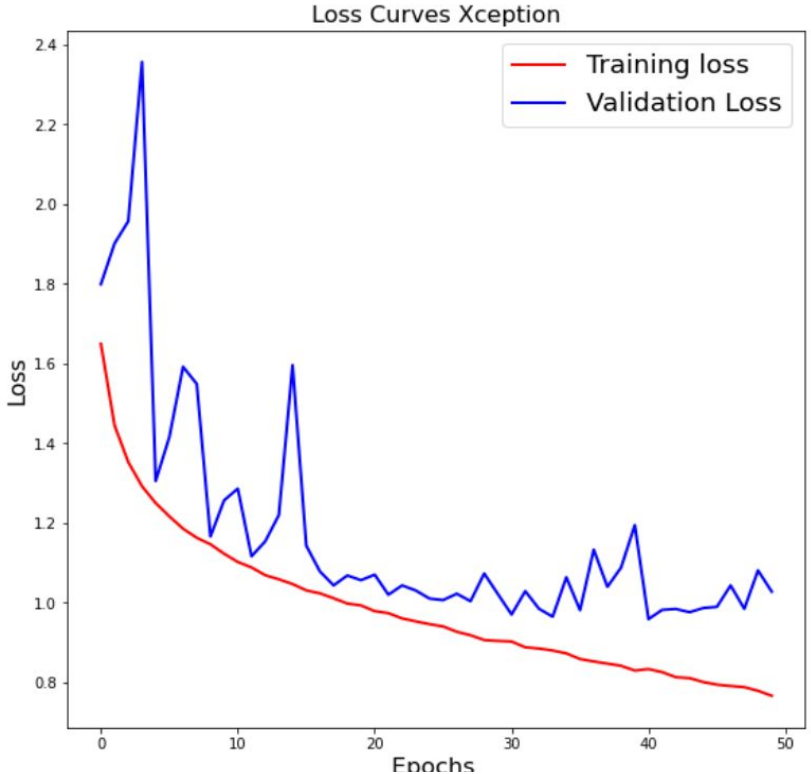
Classification Report

	precision	recall	f1-score	support
0	0.654	0.475	0.550	461
1	0.529	0.225	0.316	40
2	0.495	0.423	0.456	473
3	0.810	0.871	0.840	848
4	0.440	0.566	0.495	488
5	0.687	0.735	0.710	328
6	0.564	0.556	0.560	593
accuracy			0.623	3231
macro avg	0.597	0.550	0.561	3231
weighted avg	0.625	0.623	0.619	3231

4) Xception model



Accuracy and loss plots



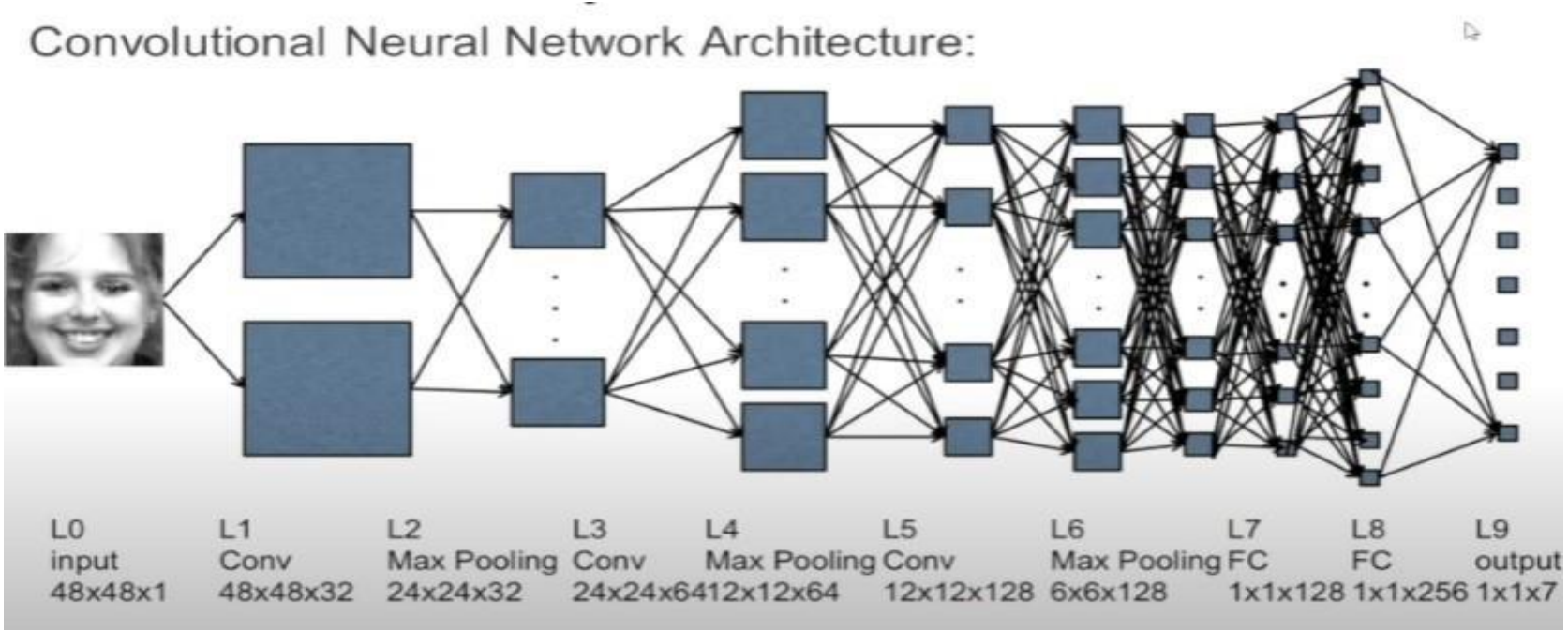
Confusion Matrix

Classification Report

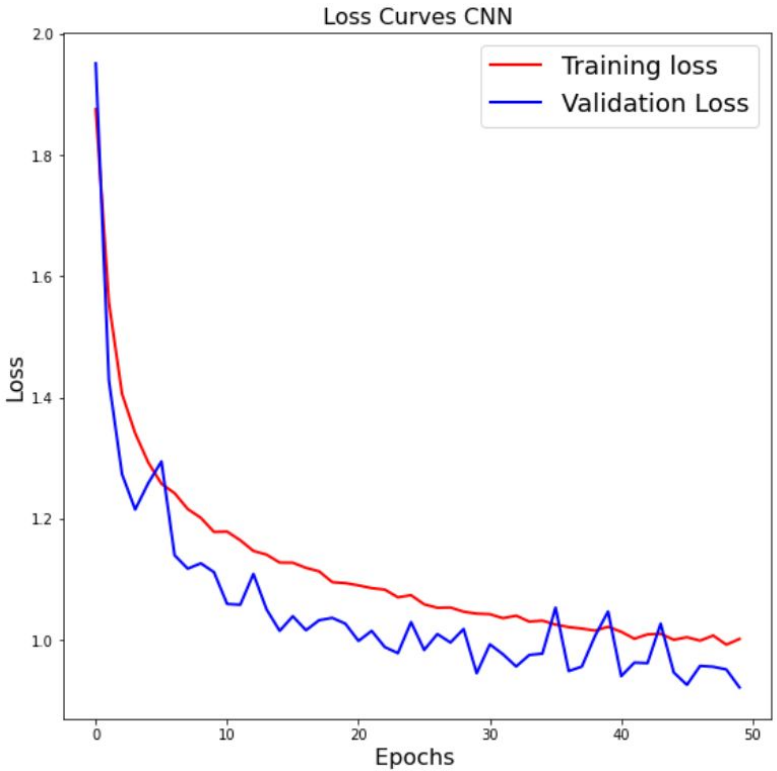
	precision	recall	f1-score	support
0	0.709	0.464	0.561	461
1	0.773	0.425	0.548	40
2	0.494	0.518	0.506	473
3	0.896	0.790	0.840	848
4	0.508	0.469	0.488	488
5	0.724	0.710	0.717	328
6	0.501	0.752	0.601	593
accuracy			0.636	3231
macro avg	0.658	0.590	0.609	3231
weighted avg	0.660	0.636	0.638	3231

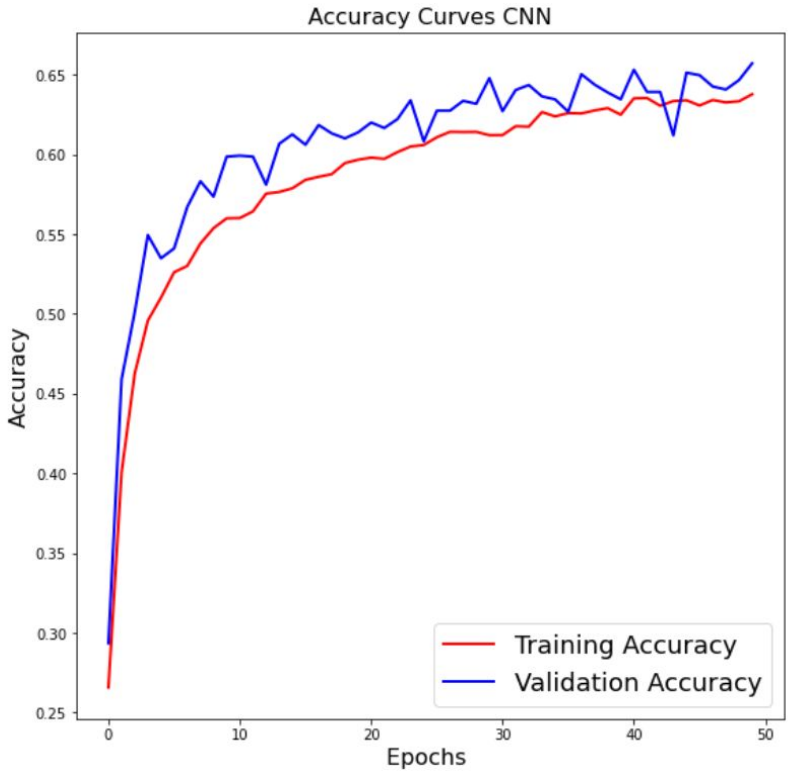
3) Using CNN layers

What is CNN ?

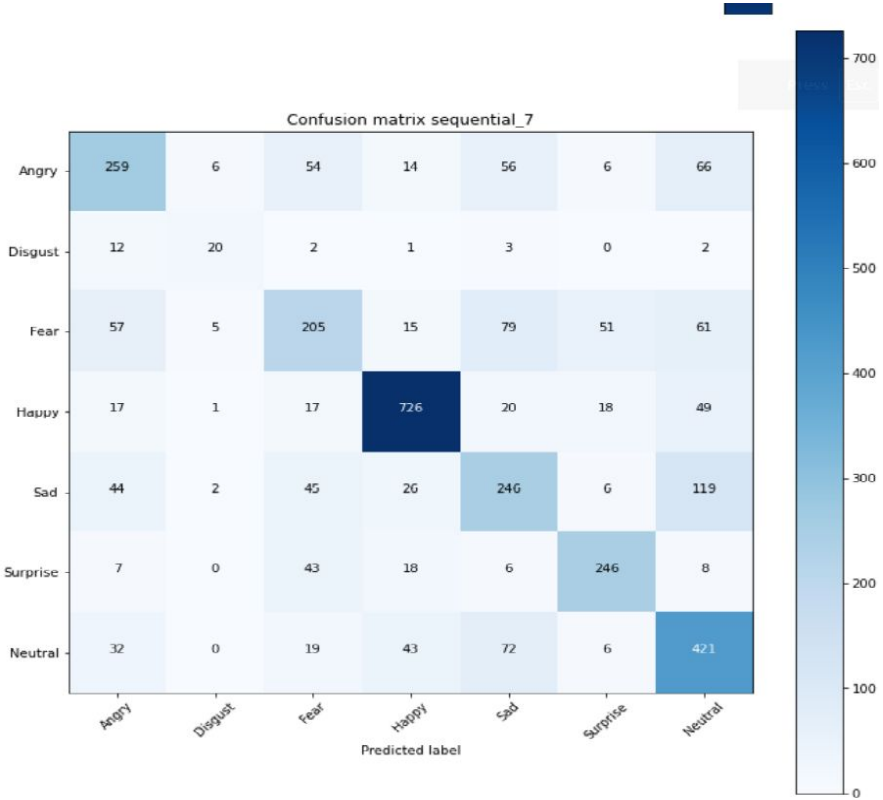


Accuracy and loss plots





Confusion Matrix



Classification Report

	precision	recall	f1-score	support
0	0.605	0.562	0.583	461
1	0.588	0.500	0.541	40
2	0.532	0.433	0.478	473
3	0.861	0.856	0.859	848
4	0.510	0.504	0.507	488
5	0.739	0.750	0.744	328
6	0.580	0.710	0.638	593
accuracy			0.657	3231
macro avg	0.631	0.616	0.621	3231
weighted avg	0.656	0.657	0.654	3231

Real-Time Local Video Face Emotion Detection

- We created patterns for detecting and predicting single faces and as well as multiple faces using OpenCV videocapture in local webcam.

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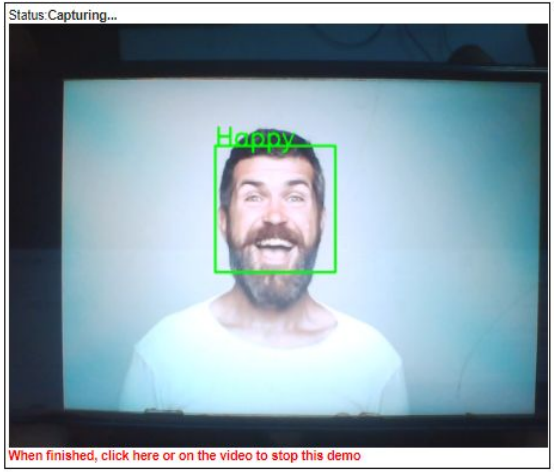
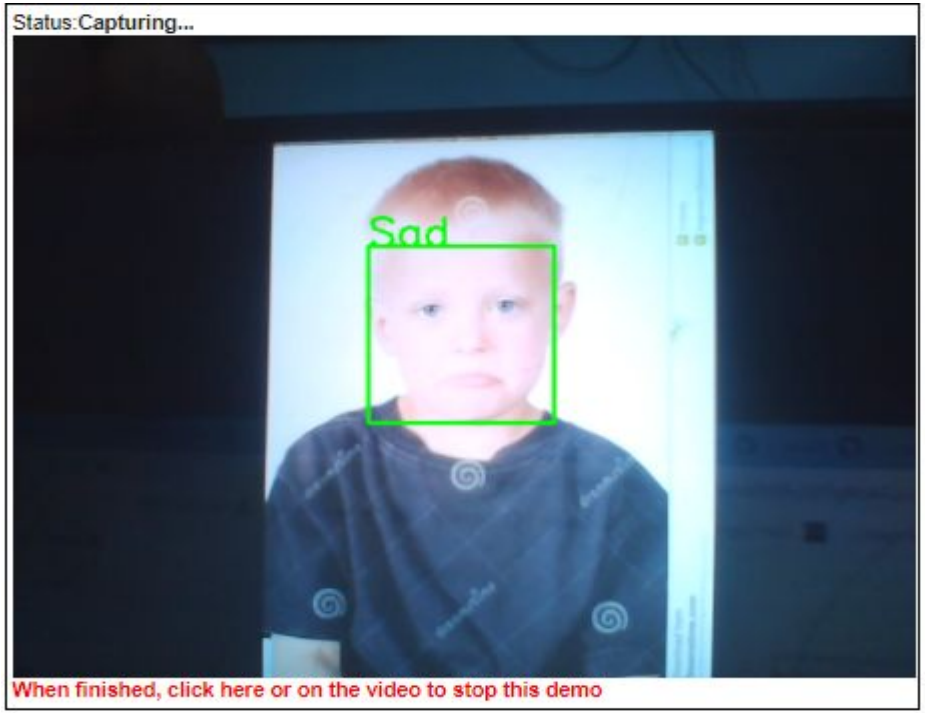
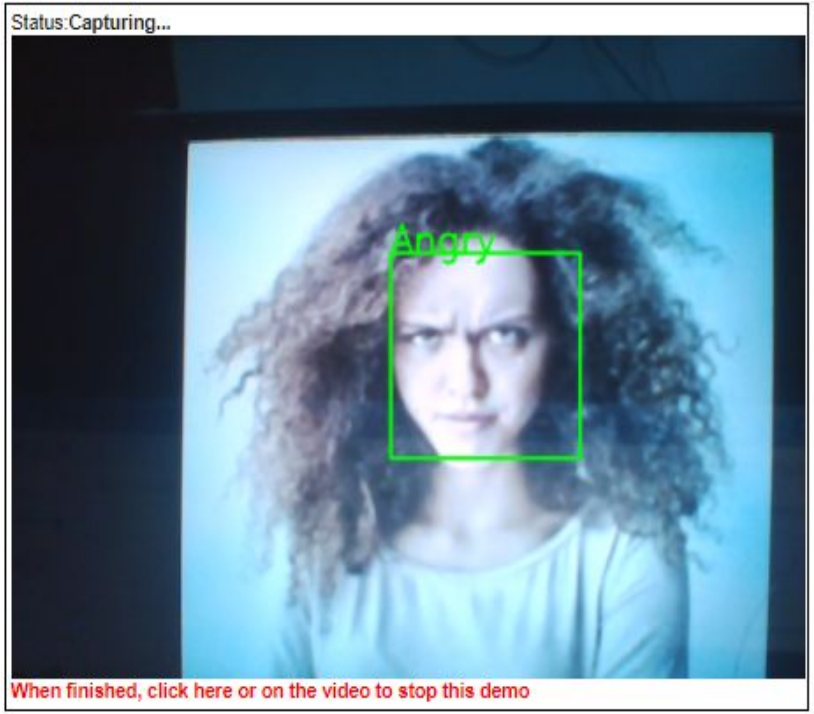
- For Web-app , OpenCV can’t be used. Thus, using Streamlit-Webrtc for front-end application.

Some Predicatied Images



Contd..


Status:Capturing...
Sad
When finished, click here or on the video to stop this demo


Status:Capturing...
Angry
When finished, click here or on the video to stop this demo

Deployment of Streamlite Web App in Heroku and Streamlit:

- We deployed the app in Heroku if you saw in the starting section of github repo you see the all the requirement files are there for creating an app on Heroku of name “face-emotion-recognition-ofg”.

- But due to high slug size the buffering takes time so we have ran our app working on local and it ran properly and app is also fine also we’ve included video on github repo.

Heroku Link: https://face-emotion-recognition-ofg.herokuapp.com/

Challenges

- Large Image Dataset to handle.
- Couldn’t able to connect GPU with Jupyter Notebook.
- Tried creating lot of models till find the best one.
- Continuous Runtime and RAM Crash due to large dataset.
- Carefully tuned Hyper parameters .

Conclusion

- We build the WebApp using streamlit and deployed in Heroku and Streamlit Sharing.
- The model which was created by custom CNN model gave training accuracy of 77% and test accuracy of 69% .
- I have also included the video of my WebApp working in Local.
- Codes which are deployed are in Github Repository.
- It was such an amazing and interesting project. We learnt a lot from this.

Some real life experience form project

- Understand the deep concept of project
- Don't afraid to failure
- Never give up
- Have some patience good things happen
- Try new things and execute your idea

Contributors Role:

Shubham Deshmukh :

- Build Deep-Face model and Resnet_50 (Residual Neural Network) model.
- Ran the local webcam for emotion detection on single as well as multiple face.

Saurabh Yadav:

- Build Xception model and Custom CNN (Convolutional Neural Network) model.
- Created streamlit webapp and deployed on Heroku.

Thank You!