

# **“Face Mask Detection”**

**Major project report submitted in partial fulfilment of the  
requirement for the degree of Bachelor of Technology**

**in**

**Computer Science and Engineering**

**By**

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**UNDER THE SUPERVISION OF Dr. Hari Singh**



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# **I DECLARATION**

**I hereby declare that this project has been done by me under the supervision of Dr. Hari Singh, Assistant Professor (SG) Jaypee University of Information Technology. I also declare that neither this project nor any part of this project has been submitted elsewhere for award of any degree or diploma.**

**Supervised by:  
Dr. Hari Singh  
Assistant Professor(SG),  
Department of Computer Science & Engineering and  
Information Technology  
Jaypee University of Information Technology**

**Submitted by:  
Varun Rangra (181451)  
Computer Science & Engineering Department  
Jaypee University of Information Technology**

## **II CERTIFICATE**

**This is to certify that the work which is being presented in the project report titled “ Face Mask Detection” in the partial fulfilment of the requirements for the award of the degree of B.Tech in Computer Science And Engineering and submitted to the Department of Computer Science And Engineering, Jaypee University of Information Technology, Wagnaghat is an authentic record of work carried out by “Varun Rangra (181451)” during the period from January 2022 to May 2022 under the supervision of Dr. Hari Singh in the Department of Computer Science and Engineering, Jaypee University of Information Technology, Wagnaghat.**

**Varun Rangra (181451)**

**The above statement made is correct to the best of my knowledge.**

**Dr. Hari Singh  
Assistant Professor(SG),  
Computer Science & Engineering and Information Technology  
Jaypee University of Information Technology, Wagnaghat,**

### **III ACKNOWLEDGEMENT**

**Firstly, I express my heartiest thanks and gratefulness to almighty God for His divine blessing makes us possible to complete the project work successfully.**

**I really grateful and wish my profound my indebtedness to Supervisor Dr. Hari Singh, Assistant Professor(SG), Department of CSE Jaypee University of Information Technology, Wakhnaghat. Deep Knowledge & keen interest of my supervisor in the field of “Computer Science” to carry out this project. Her endless patience, scholarly guidance, continual encouragement, constant and energetic supervision, constructive criticism, valuable advice, reading many inferior drafts and correcting them at all stage have made it possible to complete this project.**

**I would like to express my heartiest gratitude to Dr. Hari Singh, Department of CSE, for his kind help to finish my project.**

**I would also generously welcome each one of those individuals who have helped me straight forwardly or in a roundabout way in making this project a win. In this unique situation, I might want to thank the various staff individuals, both educating and non-instructing, which have developed their convenient help and facilitated my undertaking.**

**Finally, I must acknowledge with due respect the constant support and patients of my parents.**

**Varun Rangra (181451)**

## **IV**

### **Abstract**

**This put forward a novel mechanism in which good rules to restrain Covid-19 epidemic needs a good attention to not to impact people health and global economy. In the absence of good antiviral and with the no great medical facilities, many measure are recommended by WHO to able to control the virus rate and rid of the limited medical resources. All countries are mandating to wear the mask in social event. To wards public health, our aim to be high efficiency and a good accuracy technique that can find non-mask faces in social gathering events. We will combine dataset to build the Covid-19 face mask detector with computer vision using Python, OpenCV, and Tensor Flow and Keras. Our goal is to identify whether the person on image/video stream that the person wearing a mask or not with the help of computer vision and deep learning .**

## 1.1 Introduction

Face Mask Detection is the approach for the reduce of corona virus. All over the countries faced this pandemic. The world had made a great loss due to this pandemic. On August 2020, WHO report that the covid-19 disease is spread by the respiratory ailment. It has affected all worldwide people and the world suffers.

1> According to reports and the health experts if we want to control Covid-19 virus we need to maintain the social distance and boost our health system.

2> In recent times, many Universities clear that wearing a mask reduces the spread of covid-19. This is the only way that we can use this to reduce the spread of coronavirus.

We need to make it compulsory to wear a mask in social gathering events, so now it's our duty to provide good facilities to force people to apply a mask.

The corona virus can spread between people for example, speaking, coughing, or sneezing — even if those people are wearing masks. The recent information also gives the information the traces of a new strain of a corona virus, the mutant corona virus which, the virus has to change its structure and it becomes a strong mutant. The new strain is not even able to detect using the RT-PCR test we use now.

Our aim of the face mask detection is to determine that if there are any faces in the image or video without masks.

There are multiple faces present at the same time, each face has boundaries with boundary bins and hence we understand the locations of the faces.

Human faces aren't an easy painting for a version because there are numerous variables that can be exchanged. For example: face expressions, orientations, lighting fixtures, situations and in some conclusions inclusive of sunglasses, scarves, mask and so on.

The end result of the detection is to offer the face locations parameter that it could be required in lots of various shapes of instances, a rectangle overlaying the crucial part of the face and eye centre.

It includes eye, nose and mouth corner, eyebrow, nose, ears, and many others.



We can use Face mask detection to see the people is wear a mask or not. With this system, we can help our government to check who wears the mask and who not. We can use this system in Malls, social gathering and also for the security of people. It plays a very good role in the computer vision and the recognition.

This is highly effective and find the good accuracy

We need rules to restrain Covid-19 epidemic needs a good attention to not to impact people health and global economy. In the absence of good antiviral and with the no great medical facilities, many such as measure are recommended by WHO to be able to control the virus rate and risk of the limited.

We need to implement the rules that everyone can wear masks in social events.

We create a dataset to construct the Covid-19 face masks detector with computer imaginative and prescient the use of Python, OpenCV, and Tensor Flow and Keras. Our purpose is to become aware of whether the person on image video movement that the character wearing a mask or no longer/with the assist of laptop imaginative and prescient and deep getting to know .

1: This system is a great for the future to use this system in the public platforms, to get control in the spread of covid-19.

2: This system integrates with the highly resolutions, surveillance devices.

## 1.2 Problem Statement

All over the countries faced this pandemic. The world had made a great loss due to this pandemic. On August 2020, WHO report that the covid-19 disease is spread by the respiratory ailment. It has affected all worldwide people and the world suffers millions of deaths and tragedies.

The world has face the economy issues and people had lost their family members. Covid-19 made a huge impact in our daily to daily lives and also change our lifestyle of living. We need this system to apply on public platforms for the good of the people and it reduces the risk of spread the covid19.

According to reports and the health experts if we want to control Covid-19 virus we need to maintain the social distance and boost our health system.

We need rules to restrain Covid-19 epidemic needs a good attention to not to impact people health and global economy. In the absence of good antiviral and with the no great medical facilities, as many as measure are recommended by WHO to able to control the virus rate and risk of the limited.

We need to implement the rules that everyone can wear masks in social events. The world of covid, multidisciplinary acts has been organized to slow the spread of this pandemic. particular, they develop the method for monitor social distance or identify face mask has made the headlines in these days. But all they hype to show off result as soon as possible, to add up the usual AI overpromise factors, may be signals the wrong ideas.

That solves some of the use cases is almost trivial due to small powers of AI. In this effort to paint a good picture, we decided to show the creative process behind a solution for a seems simple use case in computer visions.

## 1.3 Objectives:

The major goal of face mask detection system is to decrease the spread of coronavirus disease is caused by acute respiratory syndrome. I build a Face Mask Detector using Keras, Tenserflow, MobileNet and Open CV. With further need improvement these type of model could be integrated with CCTV cameras to detectand identify people without masks.

The symptoms of a person with coronavirus while it is spreading fast may experience include: illness, fever, pains in joints and respiratory problem. The Face Mask Detector did not use any morphed masked photos into datasets. The model is accurate, and for the reason that MobileNetV2 architecture is used, its also computationally efficient and making it easierto set up the model to embedded systems(Raspberry Pi, Google).

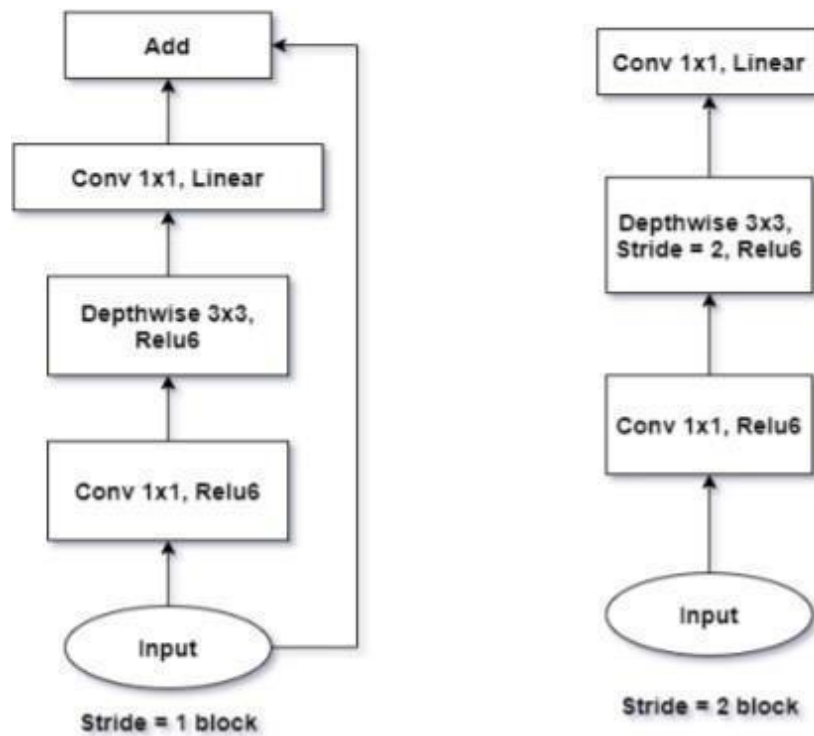
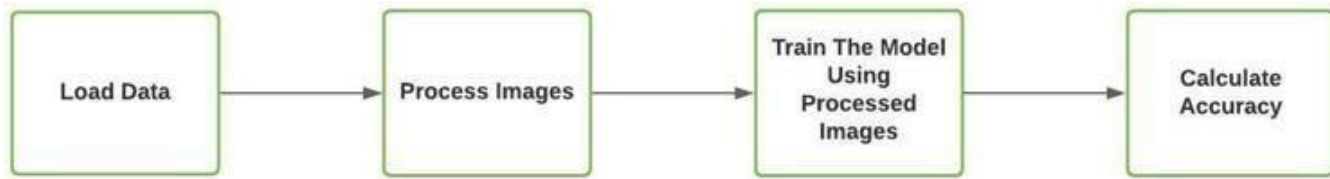
This device can therefore be utilized in actual time packages whichrequire face mask detection for protection forpurpose due to outbreak of covid regulations. We create adataset to construct the Covid-19 face mask detector with pc. Imaginative and prescient using Python, OpenCV, and Tensor Flowand Keras. Our intention is to become aware of whether the person onimage videostream that the individual carrying a masks or now not/. Vision using Python, OpenCV, and Tensor Flow and Keras. Our purpose is to identify whether or not the character onimage video streamthat the man or woman carrying a masks or now not/with the help of laptop vision and deep studying .

The world of covid, multidisciplinary acts has been organizes to slows the spread of this pandemic. All theAI communities has part of these endeavor.

In particular, they development the method for monitorsocial distance or identify face mask has made-the-headlines in these days.

Our major approach is we biuild a two databases. In one database we use images without masks of peoplesand the another one we uses images with masks.

# METHODOLOGY



*Figure1. MobileNetV2*

(It shows firstly we load data and process and train model and then find accuracy )

There are two main approaches for Face Detection System:

1> Feature Based Approach

2> Image Based Approach

We use Image based approach.

### Image Based approach:

It depend upon the strategies of statistical strategies and evaluation and device learning to find the applicable characters of face and non-face pics. The found out characteristics in the form of distribution models or discriminant characteristic which could be consequently used for face detection.

### Feature Base Approach

Object are basically recognizes to their unique feature. There are many feature in a human face, which we can be recognizes between a face and many other object. It located face by the extractes structural feature like eye, nose, mouth etc. Thus we uses them to detect the faces.

There are sort classifiers qualified to separate b/w facial and non-facial region.

1: Human face has particular texture which can be used for the differentiation between the face and other object.

2: Moreover, another feature can help us to detects the object from faces. 3: We

implements a feature-based approach by uses OpenCV.

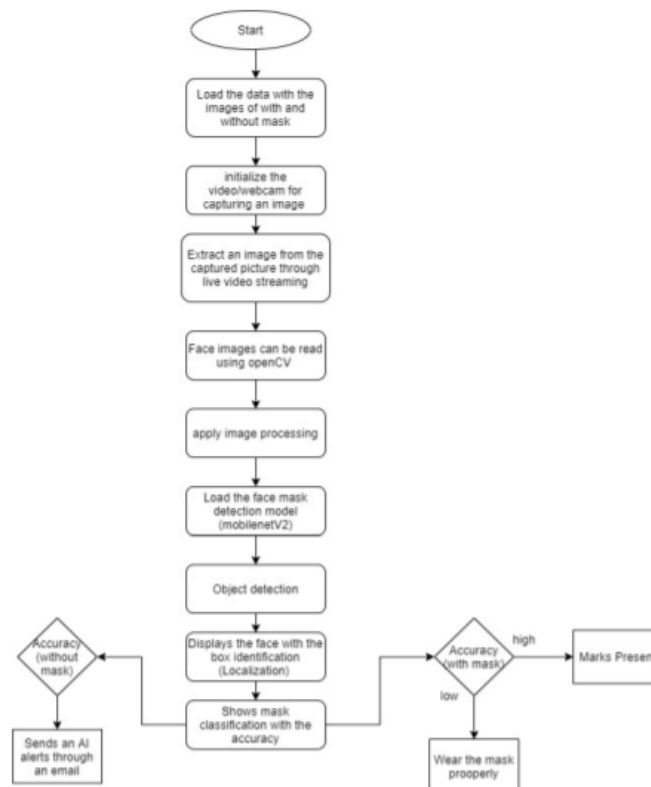
## CHAPTER 2

### 1. Face Mask Detector

Single shot detector structure is used for item detection motive. In this way to machine, face mask detector can be developed in lots of areas like buying shops, airports and heavy traffic place to monitor the public and to keep away from the unfold of the sickness by way of the use of checking who's following fundamental rule and the who is not.

It takes excessive time in load Google Colab Notebook. We are processed a machine computationally green using MobileNetV2 which makes easy to Extract the Data set.

We can use CNN structure for better performance. We can restoration it any kind of cameras.



fig(2)

## 2 > Face Detection Techniques: a review, Artificial

Human beings have not the capability to identify different faces than structures. So automatic face detection devices play a critical function in face detection, head pose estimation and so on. It has some mistakes like face occlusion, and non-uniform illumination. We use Neural Networks to hit upon faces in live video flow.

TensorFlow is likewise used inside the device. In modern times they use AdaBoost algorithm, we're able to overcome all issues in paper.

Three> Real Time Face Mask recognition with alarm tool the use of deep studying:

This process offers quicker consequences for face mask detection. Raspberry pi based real time face mask popularity that captured face photo. This device uses the characteristic of VGG-16 as the main network for face detector.

Deep learning techniques are done to gather a classifier to collect images of the individual wearing a mask and no masks.

It shows accuracy in detecting individuals wearing face masks and no longer sporting it.

This research presents the useful tool within preventing the unfold of COVID-19 virus.

The world of COVID, multidisciplinary acts have been organized to slow the spread of this pandemic. They develop the method for monitor social distance or identify face mask has made the headline in these days.

But all they hype to show off results as soon as possible, to add up the usuals.

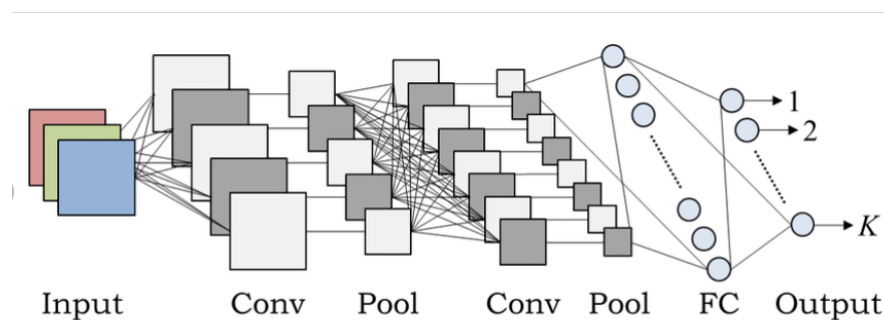
AI overpromise factors, may be signals the wrong ideas.

That solves some of the use cases is almost trivial due to small powers of AI.

#### 4> Multiple Stage CNN Architecture for Face Mask Detection

This includes a dual degree CNN structure able to detecting masked and unmasked faces and can be included with pre- installed CCTV cameras. This may be helped tracked for protection violations, and promote the use of masks and ensured protection environment. Datasets wer collected from public domain at the side of some records scraped from the internet.

We can use any digital digicam to dtecting faces. It can be beneficial for public accumulating locations and society to save you them from virus transmission. Here we use live video detection the use of open cv



An example of CNN architecture.

Fig-3(e.g input>conv>pool>conv>pool>fc>output)



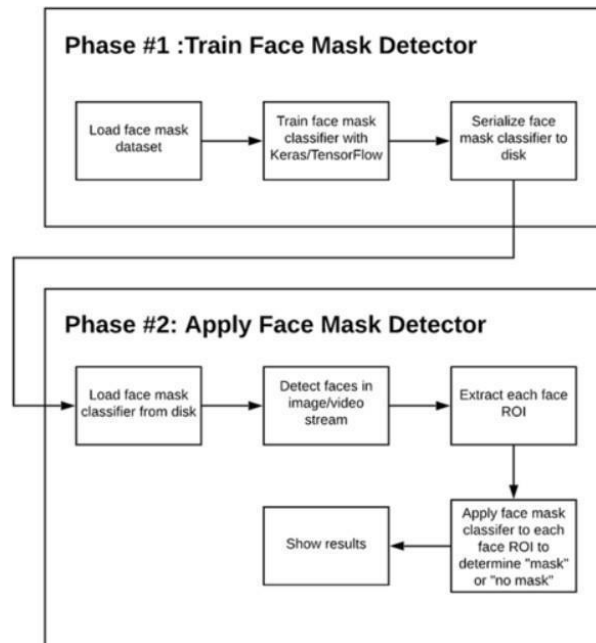


fig-4 (Training and Data Loading)

First, we load the datasets with face masks and non face masks and then we train the face mask classifier with keras/tensor flow and serialize to disk.

## CHAPTER 3- SYSTEM DEVELOPMENT

In Face Mask Detection, in recent trends, the approach of deep learning and detecting faces with and without masks was a good trend, as the Face Mask is obligatory for everyone while travelling without masks. I have created a model here that detects face masks with 3 colour channels trained on 7553 images (RGB)

The accuracy of training for the custom CNN architecture models amounted to 94% and validation exactitude to 98%.

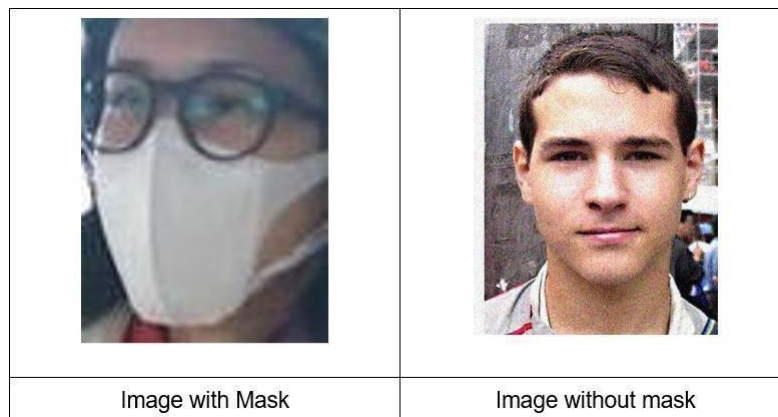


Fig-5 (We make a two datasets images with masks and non-masks)

# CONTENT

Data set comprises 7553 RGB images in mask and without mask in the two folders. Images are also referred as an maskand mask-free images.

Mask pix are 3725 face photo and 3828 faceless maskimages.N  
order to educate a custom face mask detector , we want to break  
our challenge into distinct levels, every with its very own respective  
sub- steps:

1> Training:

Here we all recognition on loading our face masks detection dataset  
from the disk, schooling a model(the usage of keras/Tener waft) on  
thisdataset, and serialise the masks detector to disk.

2> Deployment:

When the face mask detector is skilled, we are able to then flow on  
to the loading the masks detector,acting face detection, after which  
it classifying each face as with mask or with out masks.

## Mask



fig-6(It contains the dataset of  
people with masks. )

When the face mask detector is trained, we are able to then pass directly to loading the mask detector, performing face detection, after which it classifying every face as with mask or with out masks.

To create this dataset, we need to infer the vicinity of our facial systems, consisting of:

- 1 Eyes
- 2 Eyebrows
- 3 Nose
- 4 Mouth
- 5 Jawline

To use facial landmarks to construct a dataset of face carrying masks, we first need to go together with individual with out a masks.

We can use a hard and fast of photographs used to generate face mask samples as non-face mask samples,our version emerge as closely biased and fail to generaise properly.

Avoid this with the aid of taking the time to gather new examples of our faces with out mask..

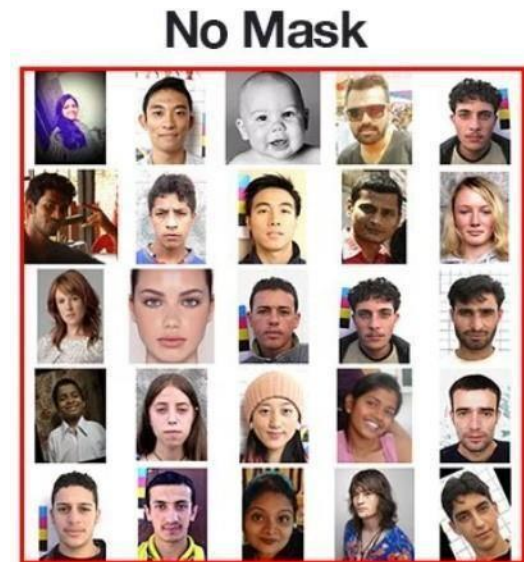


fig-7(Dataset of people without masks.)

Object are bascially recognizes to their unique feature. There are many feature in a human face, which we can be recognizes between a face and many other object.It located face by the extract structural feature like eye, nose, mouth etc.

Thus we uses them to detect the faces.

There are sort classifiers qualified to separate b/w facial and non-facial region.

Fig. 1

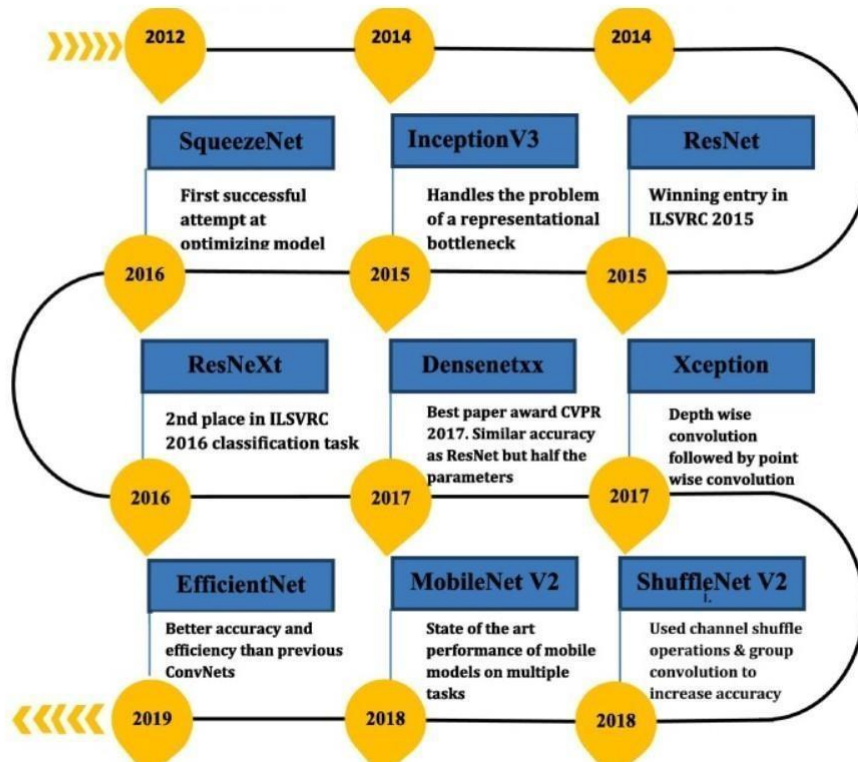


fig-8 (It shows that sequence of every year for the better accuracy or multiple tasks.)

# Face Mask Detector

Single shot detector structure can be used for object detection. In this device, face masks detector can be evolved in lots of regions like shopping branch stores, airports and heavy website site visitors locations to screen the general public and to avoid the spread of the disease with the resource of checking who's following primary rule and who isn't always. It takes excessive time in load Google Colab Notebook. We re processed a device computationally efficient the usage of the MobileNetV2 which makes smooth to Extract the Data set.

- 1: We can use CNN structure for higher performance.
- 2: We can repair it any type of cameras
- 3: Each image is divided into three bins of  $1 \times 1$ ,  $2 \times 2$ ,  $3 \times 3$  size.
- 4: We can use that for VGG-f for v-support regression.
- 5: We can use MobileNet model for predict the class of images and the fast R-CNN model on ResNet50 for predict hard image.

# Training & Loading :

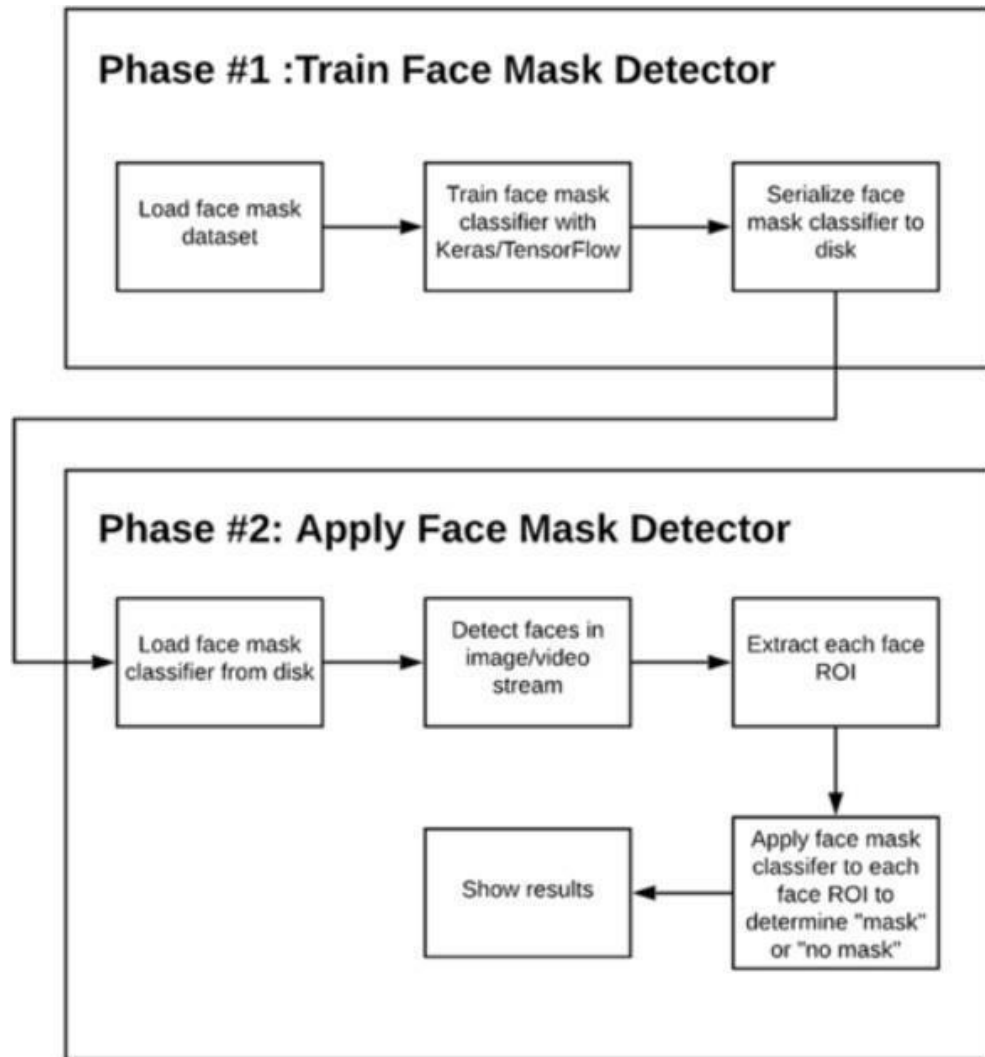


fig-9(It shows the training of data face masks and non-masks. We train the data with Keras and tensorflow)

Comparison between MobileNet-SSD, ResNet50 and Their Various Combinations based on Random vs. Hard/Soft Complexity of Test Data.

Comparison Parameters	MobileNet-SSD to ResNet50 (Left to Right)				
	100-0%	75-25%	50-50%	25-75%	0-100%
Random split (mAP)	0.8868	0.9095	0.9331	0.9650	0.9899
Soft/hard split (mAP)	0.8868	0.9224	0.9631	0.9892	0.9899
Image complexity prediction time (ms)	—	0.05	0.05	0.05	—
Mask detection time (ms)	0.05	1.92	3.08	5.07	6.02
Total Computation Time (ms)	0.05	1.97	3.13	5.12	6.02

fig-10 (We compare image complexity prediction time b/w Mobile NetSSD to ResNet50)

We compare all the data with various test data parameters.



## Design of Problem Statement:

After detect the face with wear masks and not wear masks in searches ,the unwearred masked face are passed separate to a neural networks for the further explore of the person safety and to violate facemask rules. These steps required a fixed sizes and inputs. Only one way to get fixed sizes inputs to reshape the face into boxes 96\*96 pixel.

The system is designed up by the load different pretrained model uses the torch vision packages. These model are fine tuned on to dataset with images.

:Created a Face Mask Detector includes data and collection of the person images.

Our system can be used in real time software which require face- face detection for protection purposes.

: Working at the dataset and preparing it for records analysis.

This model used MobileNetV2 architecture and it is also efficient in its miles less difficult to install the version to embedded system like raspberrypi,Google Coral, and so forth.

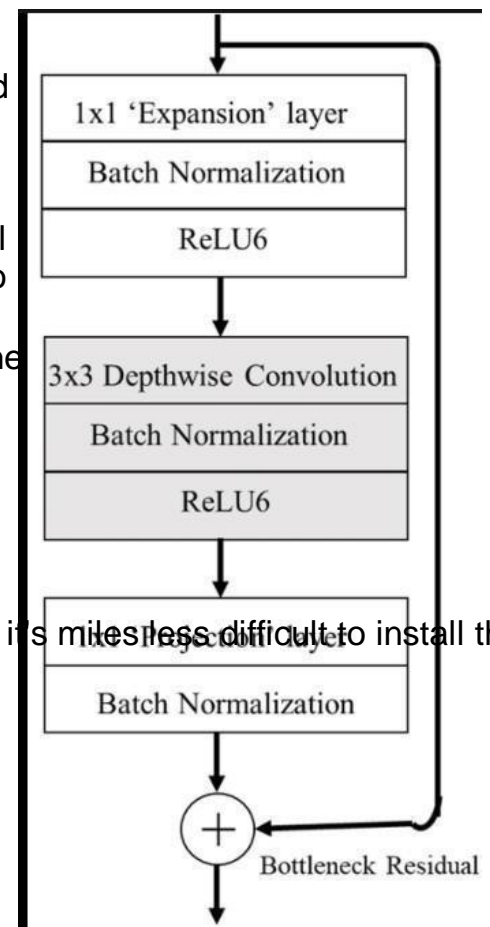


fig-11 (This model uses MobileNet V2 architecture )

Project with screenshots of various stages of projects:

## 1 Data Preprocessing:

Before we begin the analysis, we will adjust the data to ensure accurate and correct analysis.

```
1 # Import the necessary packages
2 from tensorflow.keras.preprocessing.image import ImageDataGenerator
3 from tensorflow.keras.applications import MobileNetV2
4 from tensorflow.keras.layers import AveragePooling2D
5 from tensorflow.keras.layers import Dropout
6 from tensorflow.keras.layers import Flatten
7 from tensorflow.keras.layers import Dense
8 from tensorflow.keras.layers import Input
9 from tensorflow.keras.models import Model
10 from tensorflow.keras.optimizers import Adam
11 from tensorflow.keras.applications.mobilenet_v2 import preprocess_input
12 from tensorflow.keras.preprocessing.image import img_to_array
13 from tensorflow.keras.preprocessing.image import load_img
14 from tensorflow.keras.utils import to_categorical
15 from sklearn.preprocessing import LabelBinarizer
16 from sklearn.model_selection import train_test_split
17 from sklearn.metrics import classification_report
18 from imutils import paths
19 import matplotlib.pyplot as plt
20 import numpy as np
21 import os
22
```

fig-12 (We preprocessed the data files with masks-non-face masks)

```
print("[INFO] loading images...")

data = []
labels = []

for category in CATEGORIES:
    path = os.path.join(DIRECTORY, category)
    for img in os.listdir(path):
        img_path = os.path.join(path, img)
        image = load_img(img_path, target_size=(224, 224))
        image = img_to_array(image)
        image = preprocess_input(image)

        data.append(image)
        labels.append(category)

# perform one-hot encoding on the labels
lb = LabelBinarizer()
labels = lb.fit_transform(labels)
labels = to_categorical(labels)

data = np.array(data, dtype="float32")
```

fig-13(In this fig-13 we print the datasets.)

## 2> MobileNet V2 :

It is a enormously green architecture that may be embedded devices with restricted computational capacity( Raspberry pi , Google Coral, NVIDIA Jetson Nano, and so on).

Deploying to our face mask detector to embedded gadgets should decrease the value of producing such face masks detection systems. Hence why we select to apply this architecture.

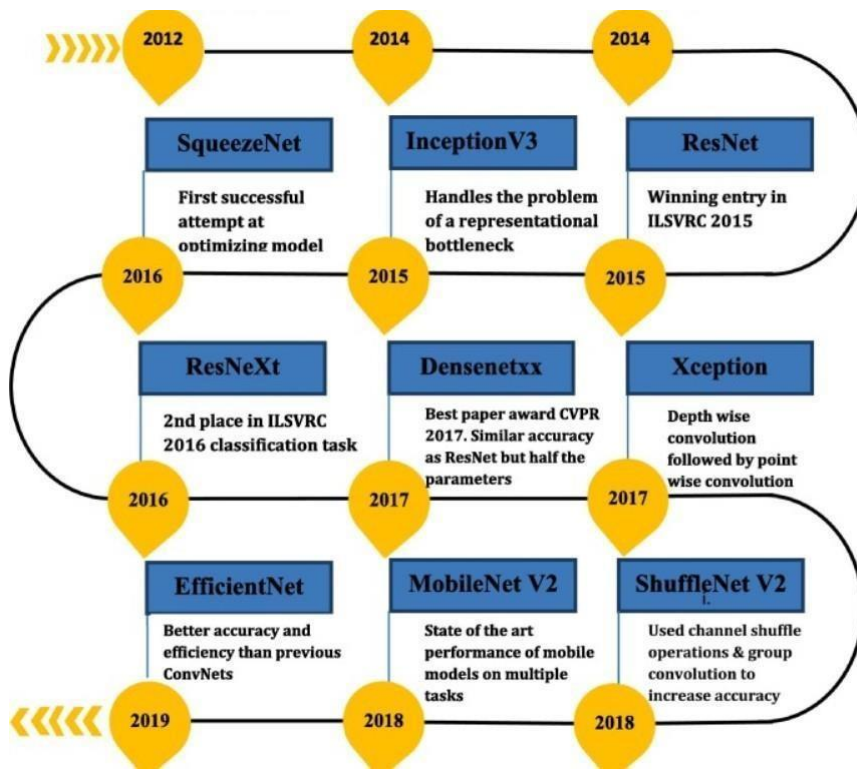


fig-15 (In our model we use MobileNet V2 for the better performance in our model)

## Command Line Arguments:

: Our command line arguments that are required to launch our script into a terminal.

Our command line arguments protected:

: Dataset : The course is the enter dataset of faces and faceswith masks.

: Plot: The direction to our output educate history plotn which willgenerate.

: Model: the path is the resulting serialized face mask type version.

```
print("[INFO] loading images...")
imagePaths = list(paths.list_images(args["dataset"]))
data = []
labels = []

# loop over the image paths
for imagePath in imagePaths:
    # extract the class label from the filename
    label = imagePath.split(os.path.sep)[-2]

    # load the input image (224x224) and preprocess it
    image = load_img(imagePath, target_size=(224, 224))
    image = img_to_array(image)
    image = preprocess_input(image)

    # update the data and labels lists, respectively
    data.append(image)
    labels.append(label)

# convert the data and labels to NumPy arrays
data = np.array(data, dtype="float32")
labels = np.array(labels)
```

fig-16(In the fig, we use command line arguments for the dataset,plot& model )

## Chapter 4: Performance Analysis

Tensor flow : It is a Google team-designed software library or framework for the easy execution of engineering and profound learning concepts.

The computational optimising algebra combines many mathematical expression for easy calculation.

Important features of Tensor Flow:

- : It contain a feature that easily define, optimises and calculate mathematical expression.
- : It includes the deep neural networks and machineteaching.
- : It also includes highly scable computing feature with different data sets.
  - : TensorFlow uses automated management, GPUcomputing.

Keras:

It is a development kit which offered an artificial neural network python interface. The TensorFlow library interface is Keras. Keras has supported several backends until the version 2.3 such as Tenseerflow, Microsoft Cognitive toolkit, Theano and PlaidML. Conceived to allowed for the fast experiment with profound neural networks.

:It is user-friendly, modular and expandable.

:It was develop as part of research effort of "ONEIROS" project and its principal author

:For the import of different image preprocessing features, we used the Tenserflow library.

The Tenserflow Keras interface is also used for MobileNets and MTCNN algorithms.



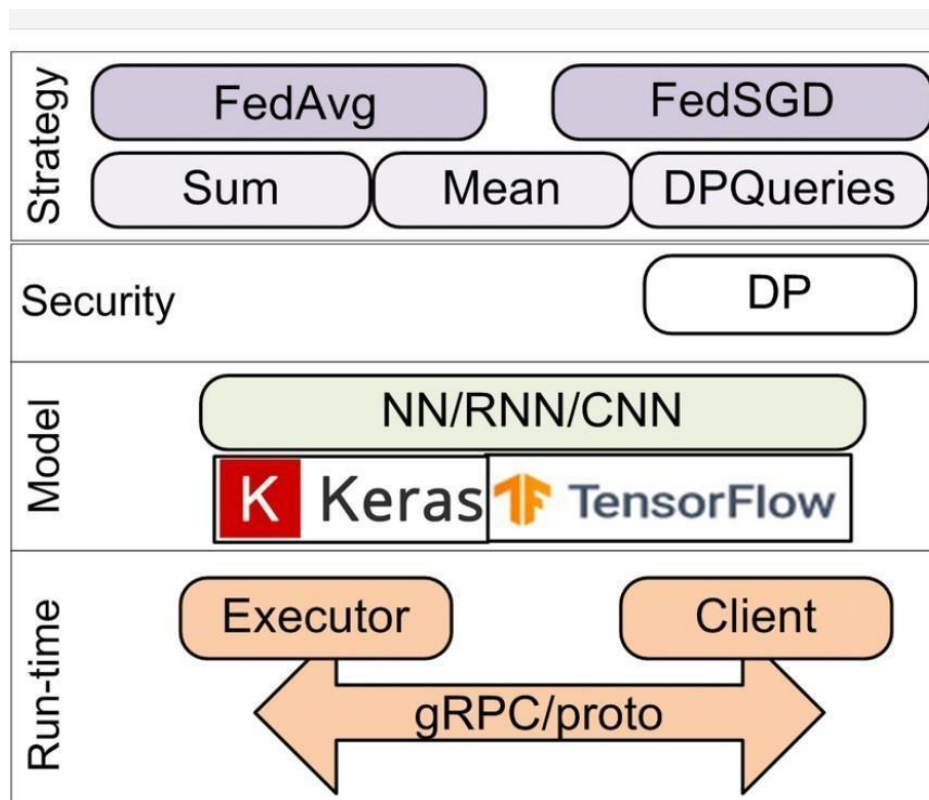


fig-17 (It shows working of Keras & tensorflow, its runtime executor and strategy.)

## Fine tuning of Pretrained model

In this work, we can achieve facemask detection by a deep neural network because of its good performance than others classification algos.

It is expensive.

Because is a time consumed task and required the high power to trains with the networks faster.

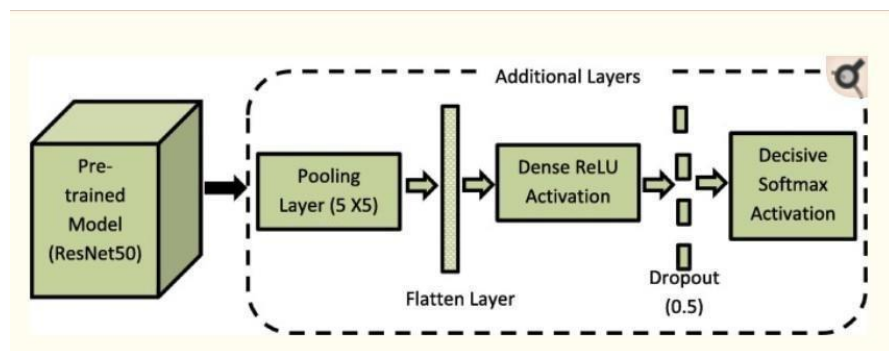


fig-18 (This shows how we done fine tuning in our model)

Supervises the pre training with the domain specific fine tuning

Firstly removed inherit biases present in the datasets and executes it with supervised learning over a specific balanced datasets.

This formula used for computing the imbalance ratio:

$$\rho = \frac{\text{Count}(\text{majority}(D_i))}{\text{Count}(\text{minority}(D_i))}$$

## Real time analysis flowchart:

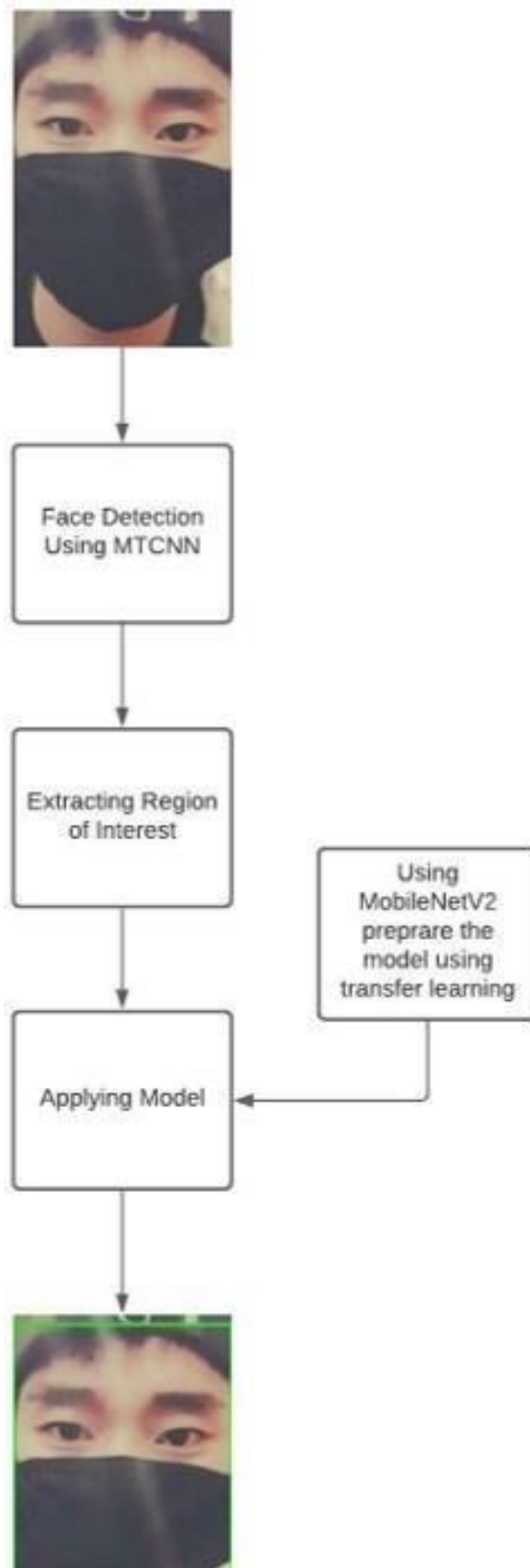


fig-19(This is the real time analysis  
it shows our model works and  
how its done. )

We are now ready ready to compile and train our face mask detector:

```
# compile our model
print("[INFO] compiling model...")
opt = Adam(lr=INIT_LR, decay=INIT_LR / EPOCHS)
model.compile(loss="binary_crossentropy", optimizer=opt,
              metrics=["accuracy"])

# train the head of the network
print("[INFO] training head...")
H = model.fit(
    aug.flow(trainX, trainY, batch_size=BS),
    steps_per_epoch=len(trainX) // BS,
    validation_data=(testX, testY),
    validation_steps=len(testX) // BS,
    epochs=EPOCHS)
```

fig-20( we compile our model and find the accuracy and print the training head. )

We will evaluate the resulting model on test set:

```
print("[INFO] evaluating network...")
predIdxs = model.predict(testX, batch_size=BS)

# for each image in the testing set we need to find the index of the
# label with corresponding largest predicted probability
predIdxs = np.argmax(predIdxs, axis=1)

# show a nicely formatted classification report
print(classification_report(testY.argmax(axis=1), predIdxs,
    target_names=lb.classes_))

# serialize the model to disk
print("[INFO] saving mask detector model...")
model.save(args["model"], save_format="h5")
```

fig-21(we print each image in the testing sets  
and serialize the model to disk.)

Our last step is to plot our accuracy and loss curves:

```
# plot the training loss and accuracy
N = EPOCHS
plt.style.use("ggplot")
plt.figure()
plt.plot(np.arange(0, N), H.history["loss"], label="train_loss")
plt.plot(np.arange(0, N), H.history["val_loss"], label="val_loss")
plt.plot(np.arange(0, N), H.history["accuracy"], label="train_acc")
plt.plot(np.arange(0, N), H.history["val_accuracy"], label="val_acc")
plt.title("Training Loss and Accuracy")
plt.xlabel("Epoch #")
plt.ylabel("Loss/Accuracy")
plt.legend(loc="lower left")
plt.savefig(args["plot"])
```

fig-22 (We plot our training loss and accuracy)

- Training and Testing accuracy VS EPOCHS

## ACCURACY :



fig-23 (This is the final result of our training and accuracy test)



## Chapter 5: Conclusion

We can use face mask detection for the purpose of people to enforced people to wears a mask in the public platforms, social gathering events. In the end its only our loved ones health and to protect them from this pandemic.

### Applications:

Because of coronavirus,we can use face mask detection to control th spread of coronavirus. As a technology advanceswe try to give the cost effective and accurate diagonosis.

## Limitation of the Project:

We tried several machine learning methods in the project, but we still can't obtain very good accuracy and due to lack of data.

## Future Work:

Finally, this works will be opened for the future planned for the countries to apply the face mask detection. In the public platforms, to get control in the spread of the covid. This system integrates with the highly resolutions surveillance devices. This system is a great for the future.

## References:

- 
1. World Health Organization et al. Coronavirus disease 2019 (covid-19): situation report, 96. 2020. - Google Search. (n.d.). [https://www.who.int/docs/default-source/coronaviruse/situation-reports/20200816-covid-19-sitrep-209.pdf?sfvrsn=5dde1ca2\\_2](https://www.who.int/docs/default-source/coronaviruse/situation-reports/20200816-covid-19-sitrep-209.pdf?sfvrsn=5dde1ca2_2).
  2. DataSet Kaggle:  
<https://www.kaggle.com/omkargurav/face-mask-dataset>
  3. Garcia Godoy L.R. Facial protection for healthcare workers during pandemics: a scoping review, *BMJ. Glob. Heal.* 2020;5(5) doi: 10.1136/bmjgh-2020-002553. [[PMC free article](#)] [[PubMed](#)] [[CrossRef](#)] [[Google Scholar](#)]
  4. Eikenberry S.E. To mask or not to mask: Modeling the potential for face mask use by the general public to curtail the COVID-19 pandemic. *Infect. Dis. Model.* 2020;5:293–308. doi: 10.1016/j.idm.2020.04.001. [[PMC free article](#)] [[PubMed](#)] [[CrossRef](#)] [[Google Scholar](#)]

## Appendices:

```
1 # import the necessary packages
2 from tensorflow.keras.preprocessing.image import ImageDataGenerator
3 from tensorflow.keras.applications import MobileNetV2
4 from tensorflow.keras.layers import AveragePooling2D
5 from tensorflow.keras.layers import Dropout
6 from tensorflow.keras.layers import Flatten
7 from tensorflow.keras.layers import Dense
8 from tensorflow.keras.layers import Input
9 from tensorflow.keras.models import Model
10 from tensorflow.keras.optimizers import Adam
11 from tensorflow.keras.applications.mobilenet_v2 import preprocess_input
12 from tensorflow.keras.preprocessing.image import img_to_array
13 from tensorflow.keras.preprocessing.image import load_img
14 from tensorflow.keras.utils import to_categorical
15 from sklearn.preprocessing import LabelBinarizer
16 from sklearn.model_selection import train_test_split
17 from sklearn.metrics import classification_report
18 from imutils import paths
19 import matplotlib.pyplot as plt
20 import numpy as np
21 import os
22
23 # initialize the initial learning rate, number of epochs to train for,
24 # and batch size
25 INIT_LR = 1e-4
26 EPOCHS = 20
27 BS = 32
```

```
28
29 DIRECTORY = r"C:\Mask Detection\CODE\Face-Mask-Detection-master\dataset"
30 CATEGORIES = ["with_mask", "without_mask"]
31
32 # grab the list of images in our dataset directory, then initialize
33 # the list of data (i.e., images) and class images
34 print("[INFO] loading images...")
35
36 data = []
37 labels = []
38
39 for category in CATEGORIES:
40     path = os.path.join(DIRECTORY, category)
41     for img in os.listdir(path):
42         img_path = os.path.join(path, img)
43         image = load_img(img_path, target_size=(224, 224))
44         image = img_to_array(image)
45         image = preprocess_input(image)
46
47         data.append(image)
48         labels.append(category)
49
50 # perform one-hot encoding on the labels
51 lb = LabelBinarizer()
52 labels = lb.fit_transform(labels)
53 labels = to_categorical(labels)
54
```

```

55 data = np.array(data, dtype="float32")
56 labels = np.array(labels)
57
58 (trainX, testX, trainY, testY) = train_test_split(data, labels,
59     test_size=0.20, stratify=labels, random_state=42)
60
61 # construct the training image generator for data augmentation
62 aug = ImageDataGenerator(
63     rotation_range=20,
64     zoom_range=0.15,
65     width_shift_range=0.2,
66     height_shift_range=0.2,
67     shear_range=0.15,
68     horizontal_flip=True,
69     fill_mode="nearest")
70
71 # load the MobileNetV2 network, ensuring the head FC layer sets are
72 # left off
73 baseModel = MobileNetV2(weights="imagenet", include_top=False,
74     input_tensor=Input(shape=(224, 224, 3)))
75
76 # construct the head of the model that will be placed on top of the
77 # the base model
78 headModel = baseModel.output
79 headModel = AveragePooling2D(pool_size=(7, 7))(headModel)
80 headModel = Flatten(name="flatten")(headModel)
81 headModel = Dense(128, activation="relu")(headModel)
82 headModel = Dropout(0.5)(headModel)
83 headModel = Dense(2, activation="softmax")(headModel)
84
85 # place the head FC model on top of the base model (this will become
86 # the actual model we will train)
87 model = Model(inputs=baseModel.input, outputs=headModel)

```

```

89 # loop over all layers in the base model and freeze them so they will
90 # *not* be updated during the first training process
91 for layer in baseModel.layers:
92     layer.trainable = False
93
94 # compile our model
95 print("[INFO] compiling model...")
96 opt = Adam(lr=INIT_LR, decay=INIT_LR / EPOCHS)
97 model.compile(loss="binary_crossentropy", optimizer=opt,
98               metrics=["accuracy"])
99
100 # train the head of the network
101 print("[INFO] training head...")
102 H = model.fit(
103     aug.flow(trainX, trainY, batch_size=BS),
104     steps_per_epoch=len(trainX) // BS,
105     validation_data=(testX, testY),
106     validation_steps=len(testX) // BS,
107     epochs=EPOCHS)
108
109 # make predictions on the testing set
110 print("[INFO] evaluating network...")
111 predIdxs = model.predict(testX, batch_size=BS)
112
113 # for each image in the testing set we need to find the index of the
114 # label with corresponding largest predicted probability
115 predIdxs = np.argmax(predIdxs, axis=1)
116
117 # show a nicely formatted classification report
118 print(classification_report(testY.argmax(axis=1), predIdxs,
119                             target_names=lb.classes_))
120
121 # serialize the model to disk
122 print("[INFO] saving mask detector model...")
123 model.save("mask_detector.model", save_format="h5")
124

```

```
124
125 # plot the training loss and accuracy
126 N = EPOCHS
127 plt.style.use("ggplot")
128 plt.figure()
129 plt.plot(np.arange(0, N), H.history["loss"], label="train_loss")
130 plt.plot(np.arange(0, N), H.history["val_loss"], label="val_loss")
131 plt.plot(np.arange(0, N), H.history["accuracy"], label="train_acc")
132 plt.plot(np.arange(0, N), H.history["val_accuracy"], label="val_acc")
133 plt.title("Training Loss and Accuracy")
134 plt.xlabel("Epoch #")
135 plt.ylabel("Loss/Accuracy")
136 plt.legend(loc="lower left")
137 plt.savefig("plot.png")
```



```
1  # import the necessary packages
2  from tensorflow.keras.applications.mobilenet_v2 import preprocess_input
3  from tensorflow.keras.preprocessing.image import img_to_array
4  from tensorflow.keras.models import load_model
5  from imutils.video import VideoStream
6  import numpy as np
7  import imutils
8  import time
9  import cv2
10 import os
11
12 def detect_and_predict_mask(frame, faceNet, maskNet):
13     # grab the dimensions of the frame and then construct a blob
14     # from it
15     (h, w) = frame.shape[:2]
16     blob = cv2.dnn.blobFromImage(frame, 1.0, (224, 224),
17                                  (104.0, 177.0, 123.0))
18
19     # pass the blob through the network and obtain the face detections
20     faceNet.setInput(blob)
21     detections = faceNet.forward()
22     print(detections.shape)
23
24     # initialize our list of faces, their corresponding locations,
25     # and the list of predictions from our face mask network
26     faces = []
27     locs = []
28     preds = []
29
```

```

--
30     # loop over the detections
31     for i in range(0, detections.shape[2]):
32         # extract the confidence (i.e., probability) associated with
33         # the detection
34         confidence = detections[0, 0, i, 2]
35
36         # filter out weak detections by ensuring the confidence is
37         # greater than the minimum confidence
38         if confidence > 0.5:
39             # compute the (x, y)-coordinates of the bounding box for
40             # the object
41             box = detections[0, 0, i, 3:7] * np.array([w, h, w, h])
42             (startX, startY, endX, endY) = box.astype("int")
43
44             # ensure the bounding boxes fall within the dimensions of
45             # the frame
46             (startX, startY) = (max(0, startX), max(0, startY))
47             (endX, endY) = (min(w - 1, endX), min(h - 1, endY))
48
49             # extract the face ROI, convert it from BGR to RGB channel
50             # ordering, resize it to 224x224, and preprocess it
51             face = frame[startY:endY, startX:endX]
52             face = cv2.cvtColor(face, cv2.COLOR_BGR2RGB)
53             face = cv2.resize(face, (224, 224))
54             face = img_to_array(face)
55             face = preprocess_input(face)
56
57             # add the face and bounding boxes to their respective
58             # lists
59             faces.append(face)
60             locs.append((startX, startY, endX, endY))
61

```

```

62         # only make a predictions if at least one face was detected
63         if len(faces) > 0:
64             # for faster inference we'll make batch predictions on *all*
65             # faces at the same time rather than one-by-one predictions
66             # in the above `for` loop
67             faces = np.array(faces, dtype="float32")
68             preds = maskNet.predict(faces, batch_size=32)
69
70         # return a 2-tuple of the face locations and their corresponding
71         # locations
72         return (locs, preds)
73
74     # load our serialized face detector model from disk
75     prototxtPath = r"face_detector\deploy.prototxt"
76     weightsPath = r"face_detector\res10_300x300_ssd_iter_140000.caffemodel"
77     faceNet = cv2.dnn.readNet(prototxtPath, weightsPath)
78
79     # load the face mask detector model from disk
80     maskNet = load_model("mask_detector.model")
81
82     # initialize the video stream
83     print("[INFO] starting video stream...")
84     vs = VideoStream(src=0).start()
85

```

```

86 # loop over the frames from the video stream
87 while True:
88     # grab the frame from the threaded video stream and resize it
89     # to have a maximum width of 400 pixels
90     frame = vs.read()
91     frame = imutils.resize(frame, width=400)
92
93     # detect faces in the frame and determine if they are wearing a
94     # face mask or not
95     (locs, preds) = detect_and_predict_mask(frame, faceNet, maskNet)
96
97     # loop over the detected face locations and their corresponding
98     # locations
99     for (box, pred) in zip(locs, preds):
100         # unpack the bounding box and predictions
101         (startX, startY, endX, endY) = box
102         (mask, withoutMask) = pred
103
104         # determine the class label and color we'll use to draw
105         # the bounding box and text
106         label = "Mask" if mask > withoutMask else "No Mask"
107         color = (0, 255, 0) if label == "Mask" else (0, 0, 255)
108
109         # include the probability in the label
110         label = "{}: {:.2f}%".format(label, max(mask, withoutMask) * 100)
111

```

```
111
112         # display the label and bounding box rectangle on the output
113         # frame
114         cv2.putText(frame, label, (startX, startY - 10),
115                     cv2.FONT_HERSHEY_SIMPLEX, 0.45, color, 2)
116         cv2.rectangle(frame, (startX, startY), (endX, endY), color, 2)
117
118     # show the output frame
119     cv2.imshow("Frame", frame)
120     key = cv2.waitKey(1) & 0xFF
121
122     # if the `q` key was pressed, break from the loop
123     if key == ord("q"):
124         break
125
126 # do a bit of cleanup
127 cv2.destroyAllWindows()
128 vs.stop()
```

```
1 tensorflow>=1.15.2
2 keras==2.3.1
3 imutils==0.5.3
4 numpy==1.18.2
5 opencv-python==4.2.0.*
6 matplotlib==3.2.1
7 scipy==1.4.1
```