

# Real-time Face Mask Detection using MTCNN, PCA, and SVM

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November 2020

## 1 Introduction

Quite some effort has been put into public-space computer vision applications to help combat the ongoing COVID-19 pandemic. One such application, **Real-time face mask detection**, involves first the detection of faces in a given frame of a video feed, and then the classification of each face to detect if the person is wearing a mask or not. If implemented nationwide, such Computer Vision solutions can have a profound impact on the spread of the virus that has created a pandemic.

**Support Vector Machines (SVMs)** have been proven to be great at a variety of classification and regression problems. We aim to create an application using a machine learning model that utilizes **Principal Component Analysis (PCA)** to reduce the dimensionality of image data, and then classify the image of a face as **with\_mask**, **without\_mask**, or as **mask\_wearred\_incorrect** (n.b: this name was a grammatical error in the labels. We will use it as-is to stay in agreement with the dataset used) with the help of an SVM.

We aim to create a real-time classification application that detects faces from a webcam using a pre-trained Multi-Task Cascaded Convolutional Neural Network (**facenet-pytorch MTCNN**).

## 2 Application Structure

Our real-time face mask detection application consists of 4 steps:

1. Sampling frames from the live feed from a camera
2. Finding all the faces through their bounding box coordinates given a frame, through MTCNN
3. Passing each individual face to the PCA-SVM pipeline and classifying them as either **with\_mask**, **without\_mask**, or as **mask\_wearred\_incorrect**

4. Displaying the result as the frame with a colored frame overlay which is green (**with\_mask**), orange (**mask\_wearred\_incorrect**), or red (**without\_mask**)

### 3 Dataset Selection for the Classifier

For training the model required for classifying face images, we chose the **face-mask-detection** dataset from Kaggle. The dataset consists of two directories:

- **annotations/**, which contains 853 XML files with metadata including:
  1. The image filename
  2. Labels for each face in each image (**with\_mask**, **without\_mask**, or **mask\_wearred\_incorrect**)
  3. Face bounding boxes for the corresponding training image.
  4. The name of the XML file is formatted as: [name\_of\_image].xml
- **images/**, which contains 853 PNG image files

The data was processed using Python into:

1. A numpy array, **data**, where each element is a flattened float32 array of length (224x224x3) generated from 224x224 scaled RGB images of extracted faces based on the annotation XML file data.
2. A numpy array, **labels**, that contains the string label for each corresponding sample in the **data** array.
3. A numpy array, **label\_data**, that contains an integer label (0, 1, 2) for each corresponding sample in the **data** array.

The **data** and **labels** array construction process was borrowed from [4].

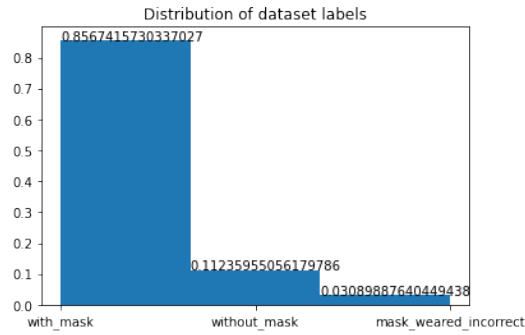


Figure 1: Dataset Label Histogram

## 4 Classifier Model - Training and Analysis

We used the PCA algorithm offered by the **scikit-learn** library to extract the top 140 principal components and thus reduce the number of features the SVM classifier will need to deal with.

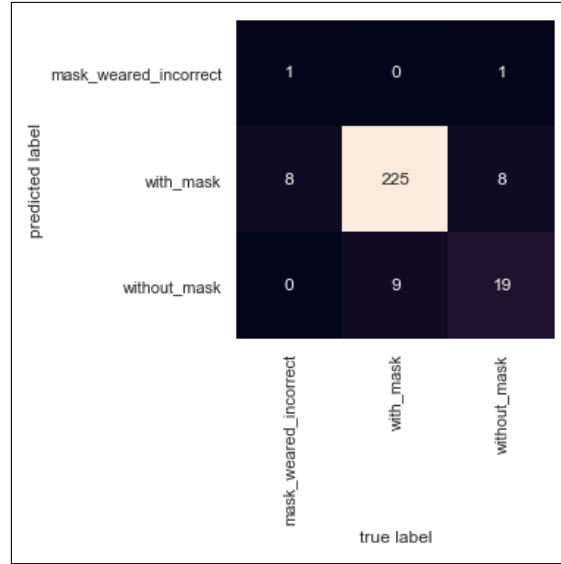


Figure 2: Confusion Matrix for the PCA-SVM classifier

The output of the PCA algorithm was pipelined into the SVM classifier provided by the **scikit-learn** library. The SVM classifier uses a radial basis function (**rbf**) kernel, which is a proven method [3] for clustering high-dimensional data.

	precision	recall	f1-score	support
mask_wearred_incorrect	0.50	0.11	0.18	9
with_mask	0.93	0.96	0.95	234
without_mask	0.68	0.68	0.68	28
accuracy			0.90	271
macro avg	0.70	0.58	0.60	271
weighted avg	0.89	0.90	0.89	271

Table 1: Classification Report for the PCA-SVM classifier

We achieved an accuracy of 90.40% with hyperparameters  $C=1$  and  $\gamma=0.005$  for the SVM, found using the **GridSearchCV** cross validation helper from the **scikitlearn** library. From the confusion matrix (Figure 2), and the recall values in the classification report (Table 1), it is clear that the model is able to

differentiate between the **with\_mask** and **without\_mask** classes. Due to lack of enough training samples however, it fails to detect improperly worn masks.

## 5 Face Detection using MTCNN

MTCNN provides a robust, pre-trained neural network model ideal for detecting faces on a live camera feed. We used it on our application to provide a cropped and scaled image of only the detected face to the PCA-SVM pipeline on the fly. The model output is then used to color the bounding box of the detected face.

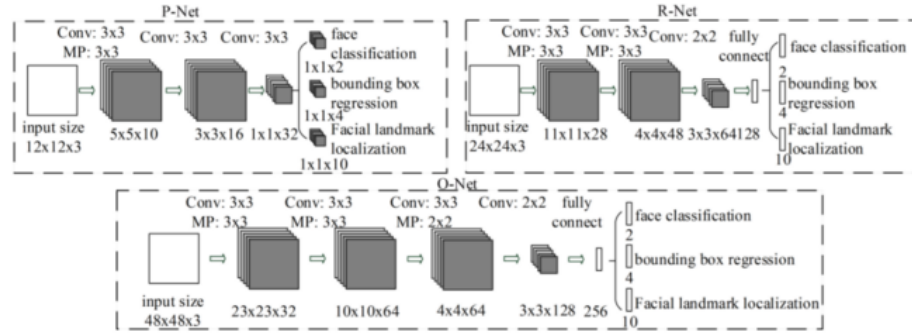


Figure 3: A quick overview of the basic structure of the MTCNN Model.[2]

MTCNN features 3-stages, each having a deep neural network with several layers:

1. The P-net - determines the probability of a face in a given kernel
2. The R-net - refines the coordinates of the boxes bounded by the P-Net
3. The O-net - takes R-net bounds and then determines facial landmarks

## 6 Conclusion

Our model fails to usually detect the **mask\_worned\_incorrect** class. However, Figures 4 and 2 show that our model is able to perform the essential classification between the **with\_mask** and **without\_mask** classes.

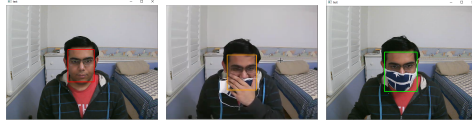


Figure 4: App output: red - no mask, orange - improper, green - with mask

## References

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