

Introduction

Motivation: According to the World Health Organization, one of the best protective measures is wearing a mask to combat COVID-19. But its effectiveness is reliant on how one is wearing them.

Problem Definition: Create machine learning models that can differentiate between a person wearing a mask correctly, incorrectly, and not wearing a mask at all

Approach: We will be building our own models and compare their performances to our baseline research paper's results. Also will be testing for biases in our dataset and models

Datasets

We combined 2 different datasets from Kaggle and images from github user cabani together to form a three class classification. With over 17,000 images and additional images for people of color and people who are white

<https://www.kaggle.com/ashishjangra27/face-mask-12k-images-dataset?>

<https://github.com/cabani/MaskedFace-Net>

Approaches

KNN

We created the KNN model using the optimal parameters of $n_neighbors = 2$, $weights = distance$, and $distance = Manhattan\ distance$, which was discovered during hyperparameter tuning. Dimensionality Reduction was used to remove unnecessary features.

Decision Tree

We created the Decision Trees model using the optimal parameters for $max_depth = 10$, and $max_features = 1138$, which were discovered during hyper parameter tuning. Dimensionality reduction was explored

SVM

Used 50% of the dataset to train the SVM model. The optimal parameters were $kernel = RBF$, $decision_func_shape = ovr$.

Naive Bayes

Used GaussianNB and MultinomialNB, along with hyperparameter tuning for an optimal image classification task.

CNN Model:

We created our own CNN model using Tensorflow and Keras. Built using 3 convolution and 3 max-pooling layers followed by some drop out layers to make our model less prone to overfitting.

Evaluation Method:

- The Accuracy, Recall, and F1 score metrics
- Comparing our performance to research paper's result
- Bias test with datasets of people of color and whites

Results

KNN

The optimal parameters resulted in 86.7% accuracy, and performed lower when dimensionality reduction techniques were included.

Decision Tree

The optimal parameters resulted in 84% accuracy, and performed lower when dimensionality reduction techniques were included.

SVM

Reached an accuracy of ~93% with our testing dataset.

Naive Bayes

Because of naive bayes assumption and how much pixels correlate with each other, it performed with ~68% for multinomial and ~77% for Gaussian.

CNN:

Best performing: Reached an accuracy of ~98% with our testing data.

Conclusions

In conclusion our best performing model was CNN, achieving 98% testing accuracy. Models that were comparable to our baseline paper underperformed. In regards to our POC and W datasets, all models also underperformed dramatically, with an emphasis on poorer results for the W dataset