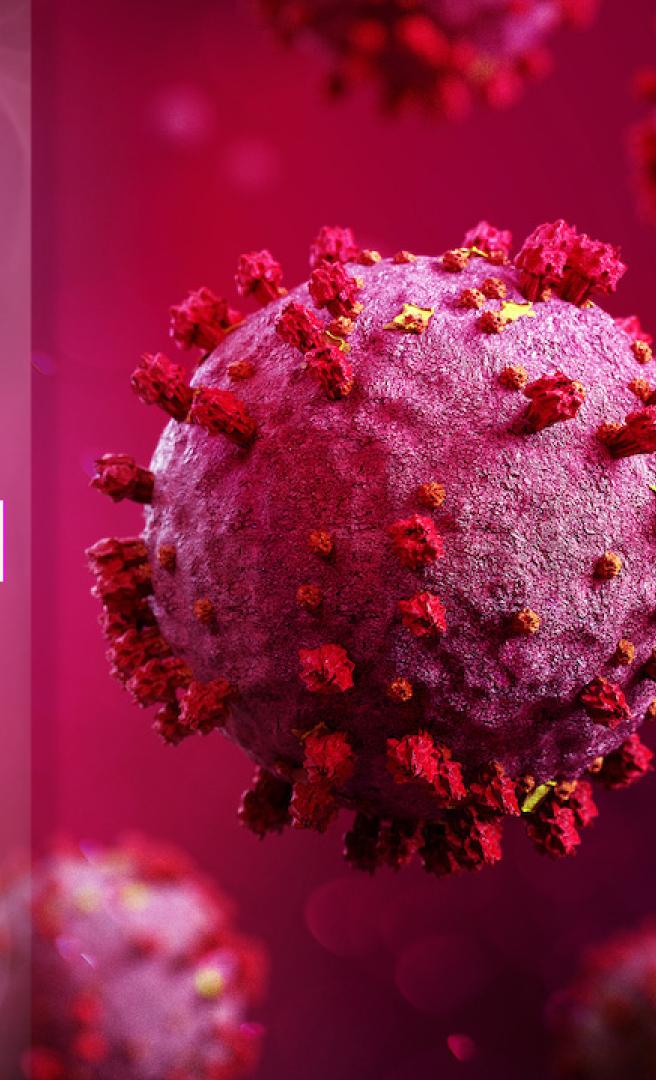


COVID CLASSIFICATION

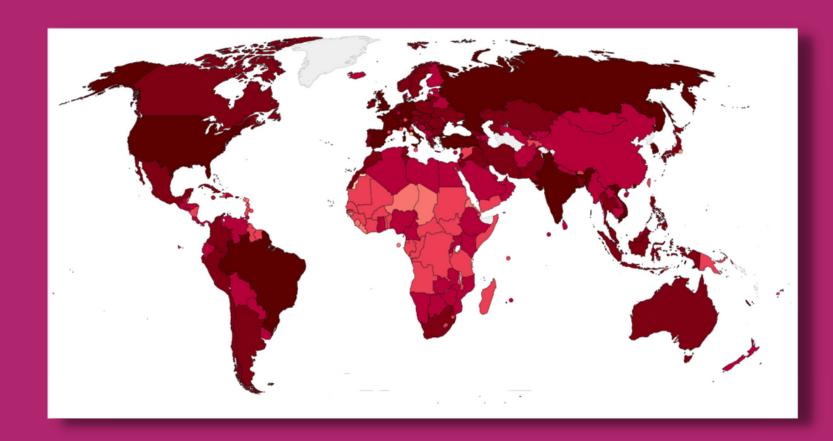
USING X RAY IMAGING

DOMAIN- HEALTHCARE / MACHINE LEARNING



CORONA-VIRUS

- The novel coronavirus diseases 2019 has created a critical and urgent threat to global health.
- This pandemic continues to challenge medical systems worldwide, and have become the biggest threat to humanity in the recent times.





Problem

Manual detection of COVID-19 diseases by the clinicians is time consuming and complex

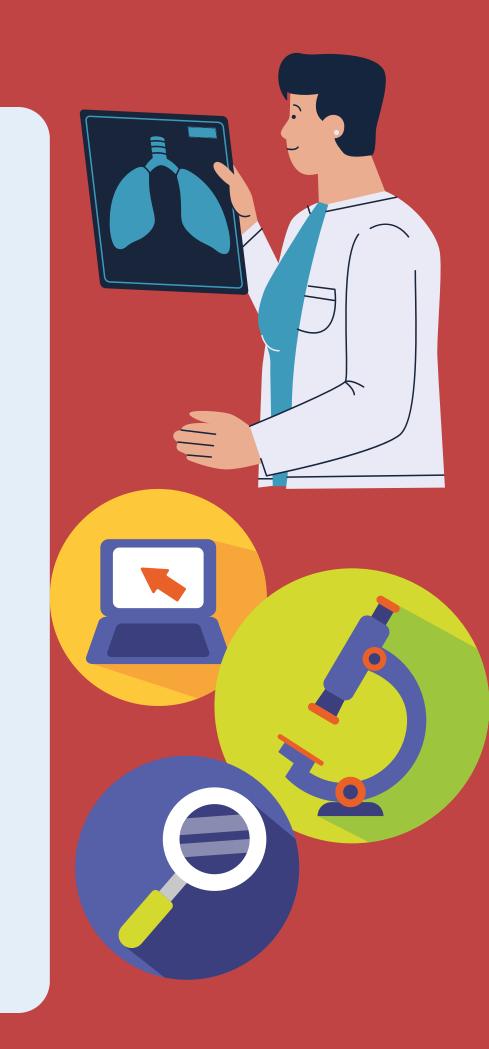
Problem

The most validated diagnosis test (RT-PCR) has long been in shortage in developing countries.

....

HEALTHCARE-MACHINE LEARNING-IMAGE PROCESSING

- Al methods (ML), coupled with biomedical analysis can play a crucial role during pandemics.
- The model is designed to meticulously classify the covid-19 patients using IMAGE PROCESSING.
- The model is able to detect COVID-19 diseases by analyzing chest X-RAYS images.
- The image processing technique is a rapid and finer approach.
- The model provides effective inspection just by analyzing an image.
- The model imparts the specific area in the images for better treatment



• PRE-PROCESSING

Data preprocessing is the concept of changing the raw data into a clean data set.

FEATURE EXTRACTION

Local Binary Pattern (LBP) has been used for feature extraction from the pre-processed images

LBP is an efficient texture operator which thresholds the neighboring pixels based on the value of the current pixel.

FEATURE SELECTION

Dominant features are then selected for further classification by getting rid of noise in data.

CLASSIFICATION

Support Vector Machine(SVM) using the linear kernal has been used for the identification of new observations on the basis of trained data.

PROJECT PIPELINE

INPUT

FEATURE EXTRACTION-LBP

CONCATENATION

CLASSIFICATION-SVM

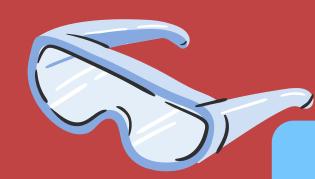
ANALYSIS

```
from google.colab import drive
     drive.mount('/content/drive')
    Mounted at /content/drive
[ ] %cd drive/
     /content/drive
[ ] %cd MyDrive/
     /content/drive/MyDrive
[ ] %cd COVID/Normal/
     /content/drive/MyDrive/COVID/Normal
[ ] %cd ...
     /content/drive/MyDrive/COVID
[ ] import glob
     import os
     from matplotlib import pyplot as plt
     import numpy as np
     from sklearn.svm import SVC
     from sklearn.metrics import accuracy_score
     from sklearn.metrics import confusion_matrix
[ ] %ls-1
    total 8
     drwx----- 2 root root 4096 Apr 9 12:49 Affected/
     drwx----- 2 root root 4096 Apr 9 07:51 Normal/
[ ] normal_files = os.listdir('Normal/')
     print(normal files)
    affected_files = os.listdir('Affected/')
     print(affected_files)
```

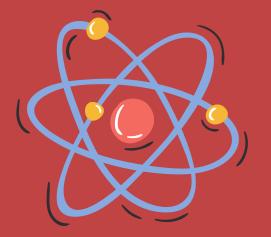
```
%cd Normal/
   /content/drive/MyDrive/COVID/Normal
def extract_feat(img):
      out_img= np.zeros((img.shape[0], img.shape[1]))
      for i in range(1, img.shape[0]-1):
        for j in range(1, img.shape[1]-1):
         n=[0]*8
         cen= img[i, j]
          n[0]=img[i, j-1]>cen
          n[1]=img[i-1, j-1]>cen
          n[2]=img[i-1, j]>cen
          n[3]=img[i-1, j+1]>cen
          n[4]=img[i, j+1]>cen
          n[5]=img[i+1, j+1]>cen
          n[6]=img[i+1, j]>cen
          n[7]=img[i+1, j-1]>cen
          temp=0
          for k in range(len(n)):
           temp+=n[k]*(2**k)
          out_img[i, j]=temp
      return out_img
   def final feat(img):
     lbp img=extract feat(img)
     hist1=np.histogram(lbp_img, bins=256)
     return(hist1[0])
   img_arr=plt.imread(normal_files[0])
    print(img_arr)
   plt.imshow(img_arr)
   [[1 1 1 ... 8 7 2]
     [1 1 1 ... 8 7 2]
     [1 1 1 ... 8 6 2]
     [0 0 0 ... 5 4 2]
     [0 0 0 ... 5 3 0]
     [0 0 0 ... 5 3 0]]
    <matplotlib.image.AxesImage at 0x7fdbaf313890>
     500
     3000
    2500
     2000
                 1000 1000 2000 2500
             500
```

```
plt.imshow(lbp_img)
<matplotlib.image.AxesImage at 0x7fdbaddd0d90>
  500
 3000
 2500
 2000
         500 1000 1500 2000
hist1=np.histogram(lbp_img, bins=256)
print(hist1[0])
[975589 128618 43874 89249 96247 7584 78154 107698 44612 10892
   2335 4035 78543 10589 100482 94297 128769 11203
                     10974
                            6961
                      684
                            4326
                                  1114
                       435
                       579
                             1083
   7629
          501
                       393
                             518
                                          434
                                                 131
                596
                                    38
    886
          894
                       161
                             2516
          454
                      8471
                             4376
                      4012
                            15264
              8626
                      7055 45876
                      8579
                 651
                 763
                      7115
              81782 110310
                            4288 82587
                                         6229
        1765 9029 15065 108261 102072 9197 106817
                            948 9284 10278 9281 23091 99931
 105120 44604 11291 44656 2553 1221 9308 15392 109231 44721
   9506 75070 53483 15389 100464 216535]
hist1[0].shape
(256,)
normal_features=np.zeros((1, 256))
for file in normal files:
  img1=plt.imread(file)
  feat1=final feat(img1)
  feat1=np.reshape(feat1, (1, feat1.shape[0]))
  normal_features=np.vstack((normal_features, feat1))
normal_features= normal_features[1:,:]
normal_features.shape
(20, 256)
```

```
affected_features= affected_features[1:,:]
     affected_features.shape
     (20, 256)
X=np.vstack((normal_features, affected_features))
     X.shape
_, (40, 256)
y=[0]*20 + [1]*20
     1]
     clf=SVC(C=0.04, kernel = "linear")
     clf.fit(X,y)
     y_pred = clf.predict(X)
     acc = accuracy_score(y, y_pred)
     tn, fp, fn, tp = confusion_matrix(y, y_pred).reshape(-1)
     recall = tp/(tp+fn)
     precision = tp/(tp+fp)
     f1_score = (2*recall*precision)/(recall + precision)
[ ] acc
```



PROJECT SPECIALITY





MODEL IS NOT RESTRICTED TO PARTICULAR VARIANT



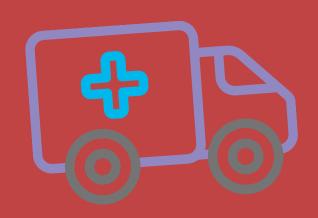
MODEL CAN BE USED WORLDWIDE FOR MEDICAL
TREATMENT AND RESEARCH



MODEL WILL ASSIST THE CLINISTS/DOCTORS AND MITIGATE THE BURDEN ON HEALTH CARE SYSTEM



MODEL ABILITY INCREASES WITH QUANTITY OF DATA AND MAY LEAD TO BETTER PERFOMANCE















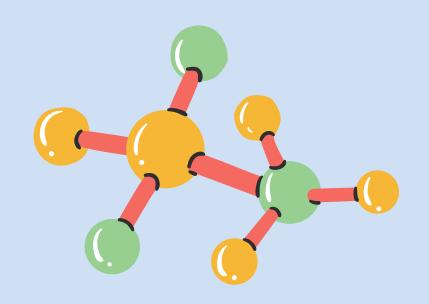
FUTURE SCOPES

- BEST COMBINATIONS OF FEATURE EXTRACTOR AND CLASSIFIER FOR GENERIC RESULTS
- CONFIGURE WITH DEEP LEARNING
- FEATURE OF VIRUS DETECTION
- PROCESSING WITH LARGE DATA
- WEB APPLICATION
- ANDROID / IOS APPLICATION

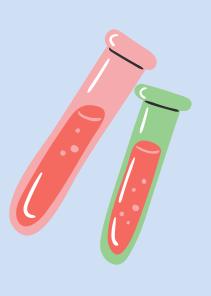
WE STAND WITH EVERY

FRONTLINE WORKER



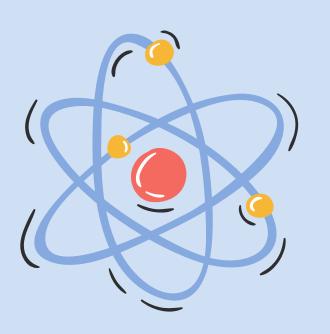








STAY SAFE



MANAV RACHNA UNIVERSITY

TANNU SANGWAN

B.TECH CSE, 4th semester

ABHEER MEHROTRA

B.TECH CSE, 4th semester