```
In [1]:
        !unzip '/content/drive/My Drive/Covid19Pred/Dataset kaggle.zip'
        Archive: /content/drive/My Drive/Covid19Pred/Dataset kaggle.zip
           creating: Dataset_kaggle/
           creating: Dataset kaggle/COVID/
         extracting: Dataset kaggle/COVID/Covid (1).png
          inflating: Dataset kaggle/COVID/Covid (10).png
          inflating: Dataset kaggle/COVID/Covid (100).png
          inflating: Dataset kaggle/COVID/Covid (1000).png
          inflating: Dataset_kaggle/COVID/Covid (1001).png
          inflating: Dataset kaggle/COVID/Covid (1002).png
          inflating: Dataset kaggle/COVID/Covid (1003).png
          inflating: Dataset kaggle/COVID/Covid (1004).png
          inflating: Dataset_kaggle/COVID/Covid (1005).png
          inflating: Dataset kaggle/COVID/Covid (1006).png
          inflating: Dataset_kaggle/COVID/Covid (1007).png
          inflating: Dataset_kaggle/COVID/Covid (1008).png
          inflating: Dataset kaggle/COVID/Covid (1009).png
          inflating: Dataset kaggle/COVID/Covid (101).png
          inflating: Dataset_kaggle/COVID/Covid (1010).png
          inflating: Dataset kaggle/COVID/Covid (1011).png
        import os
In [2]:
        import cv2
        import numpy as np
        from PIL import Image
        import matplotlib.pyplot as plt
        from sklearn.model selection import train test split
        from tensorflow import keras
        from keras.models import Sequential
        from keras.layers import Conv2D,MaxPooling2D,Dense,Flatten,Dropout
        from keras.layers.normalization import BatchNormalization
        Using TensorFlow backend.
        yes=os.listdir('/content/Dataset kaggle/COVID')
In [3]:
        no=os.listdir('/content/Dataset_kaggle/non-COVID')
In [4]:
        data=np.concatenate([yes,no])
        len(data)==len(yes)+len(no)
Out[4]: True
In [5]:
        target x=np.full(len(yes),1)
        target y=np.full(len(no),0)
        data_target=np.concatenate([target_x,target_y])
In [6]:
        yes_values=os.listdir('/content/Dataset_kaggle/COVID')
        no values=os.listdir('/content/Dataset kaggle/non-COVID')
```

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In [7]: X_data =[]
         for file in yes_values:
             img = cv2.imread('/content/Dataset_kaggle/COVID/'+file)
             face = cv2.resize(img, (227, 227) )
              (b, g, r)=cv2.split(face)
             img=cv2.merge([r,g,b])
             X_data.append(img)
 In [8]: for file in no_values:
              img = cv2.imread('/content/Dataset_kaggle/non-COVID/'+file)
             face = cv2.resize(img, (227, 227) )
              (b, g, r)=cv2.split(face)
             img=cv2.merge([r,g,b])
             X_data.append(img)
In [9]: X = np.squeeze(X data)
In [10]: X = X.astype('float32')
         X /= 255
In [11]: x_train,x_test,y_train,y_test=train_test_split(X, data_target, test_size=0.2, rain)
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```
#Now let us define AlexNet CNN model to train the model
In [12]:
         model=Sequential()
         #1 conv layer
         model.add(Conv2D(filters=96,kernel size=(11,11),strides=(4,4),padding="valid",ac
         #1 max pool layer
         model.add(MaxPooling2D(pool size=(3,3),strides=(2,2)))
         model.add(BatchNormalization())
         #2 conv layer
         model.add(Conv2D(filters=256,kernel_size=(5,5),strides=(1,1),padding="valid",act
         #2 max pool layer
         model.add(MaxPooling2D(pool_size=(3,3),strides=(2,2)))
         model.add(BatchNormalization())
         #3 conv laver
         model.add(Conv2D(filters=384,kernel size=(3,3),strides=(1,1),padding="valid",act
         #4 conv layer
         model.add(Conv2D(filters=384,kernel size=(3,3),strides=(1,1),padding="valid",act
         #5 conv layer
         model.add(Conv2D(filters=256,kernel size=(3,3),strides=(1,1),padding="valid",act
         #3 max pool layer
         model.add(MaxPooling2D(pool size=(3,3),strides=(2,2)))
         model.add(BatchNormalization())
         model.add(Flatten())
         #1 dense layer
         model.add(Dense(4096,input shape=(227,227,3),activation="relu"))
         model.add(Dropout(0.4))
         model.add(BatchNormalization())
         #2 dense Layer
         model.add(Dense(4096,activation="relu"))
         model.add(Dropout(0.4))
         model.add(BatchNormalization())
         #3 dense Layer
         model.add(Dense(1000,activation="relu"))
         model.add(Dropout(0.4))
         model.add(BatchNormalization())
```

#output layer

model.add(Dense(20,activation="softmax"))

model.summary()

Model: "sequential_1"

Layer (type)	Output	Shape	Param #
conv2d_1 (Conv2D)		55, 55, 96)	34944
max_pooling2d_1 (MaxPooling2	(None,	27, 27, 96)	0
batch_normalization_1 (Batch	(None,	27, 27, 96)	384
conv2d_2 (Conv2D)	(None,	23, 23, 256)	614656
max_pooling2d_2 (MaxPooling2	(None,	11, 11, 256)	0
batch_normalization_2 (Batch	(None,	11, 11, 256)	1024
conv2d_3 (Conv2D)	(None,	9, 9, 384)	885120
conv2d_4 (Conv2D)	(None,	7, 7, 384)	1327488
conv2d_5 (Conv2D)	(None,	5, 5, 256)	884992
max_pooling2d_3 (MaxPooling2	(None,	2, 2, 256)	0
batch_normalization_3 (Batch	(None,	2, 2, 256)	1024
flatten_1 (Flatten)	(None,	1024)	0
dense_1 (Dense)	(None,	4096)	4198400
dropout_1 (Dropout)	(None,	4096)	0
batch_normalization_4 (Batch	(None,	4096)	16384
dense_2 (Dense)	(None,	4096)	16781312
dropout_2 (Dropout)	(None,	4096)	0
batch_normalization_5 (Batch	(None,	4096)	16384
dense_3 (Dense)	(None,	1000)	4097000
dropout_3 (Dropout)	(None,	1000)	0
batch_normalization_6 (Batch	(None,	1000)	4000
dense_4 (Dense)	(None,	•	20020
Total params: 28.883.132			

Total params: 28,883,132 Trainable params: 28,863,532 Non-trainable params: 19,600

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In [13]: model.compile(optimizer="adam", loss="sparse categorical crossentropy", metrics=
```

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In [14]: history=model.fit(x_train, y_train, epochs=1000, batch_size=128,validation_data=
     Train on 1984 samples, validate on 497 samples
     Epoch 1/1000
     curacy: 0.6003 - val loss: 586.0645 - val accuracy: 0.0000e+00
     Epoch 2/1000
     uracy: 0.7999 - val_loss: 474.6297 - val_accuracy: 0.4990
     Epoch 3/1000
     uracy: 0.8387 - val loss: 461.3928 - val accuracy: 0.4990
     Epoch 4/1000
     uracy: 0.8765 - val loss: 147.5300 - val accuracy: 0.4990
     Epoch 5/1000
     uracy: 0.9022 - val loss: 141.1609 - val accuracy: 0.4990
     Epoch 6/1000
     1984/1984 [============= ] - 3s 2ms/step - loss: 0.2466 - acc
     uracy: 0.9047 - val loss: 85.4918 - val accuracy: 0.4990
```

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In [15]: final_loss, final_acc = model.evaluate(x_test, y_test, verbose=0)
print('The final accuracy is ',final_acc)
```

The final accuracy is 0.9738430380821228

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In [15]:
```