

1

Topic:-

Covid-19 Chest XRay Image Recognition

Problem Statement:-

Build an image classification model which will Identify covid positive patient by looking into the chest Xray images.

General Lifecycle of Image classification project

- 1) Data Collection
- 2) Importing libraries
- 3) Data Preprocessing
- 4) Model Building
- 5) Optimizing Model
- 6) Checking/ Validating model
- 7) Result

Importing all the important Library

In [1]:

```
1 import matplotlib.pyplot as plt
2 import cv2
3 import pathlib
4 from keras.layers import Dense, Conv2D, MaxPool2D, Dropout, Flatten
5 from keras.models import Sequential
```

In [28]:

```
1 #import imp lib
2 import numpy as np
3 import os
4 import pandas as pd
5 import tensorflow as tf
6 from tensorflow import keras
7 from tensorflow.keras import layers
8 import keras_tuner
9 from keras_tuner import RandomSearch
10 from keras_tuner.engine.hyperparameters import HyperParameters
11 from sklearn.metrics import f1_score, accuracy_score
12 from tensorflow.keras.metrics import Accuracy
13 from tensorflow.keras.preprocessing import image
14 from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

Importing Data

About Dataset

Context Helping Deep Learning and AI Enthusiasts like me to contribute to improving COVID-19 detection using just Chest X-rays.

Content It is a simple directory structure branched into test and train and further branched into the respective 3 classes which contains the images.

Acknowledgements The University of Montreal for releasing the images.

Inspiration Help the medical and researcher community by sharing my work and encourage them to contribute extensively.

In [3]:

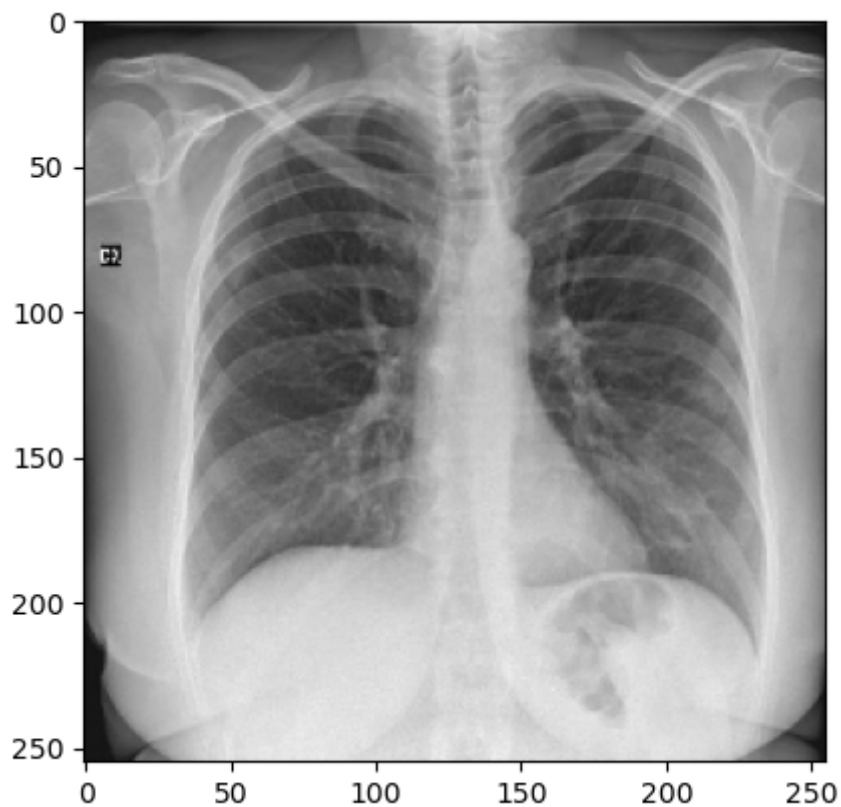
```
1 img = cv2.imread('C:/Users/Shivam Ranshur/ANACONDA JUPYTER PROGRAMMING/Covid19-datas
2 img = cv2.resize(img, (255, 255))
3
4 img = img / 255.0
```

In [4]:

```
1 plt.imshow(img)
2 img.shape
```

Out[4]:

(255, 255, 3)

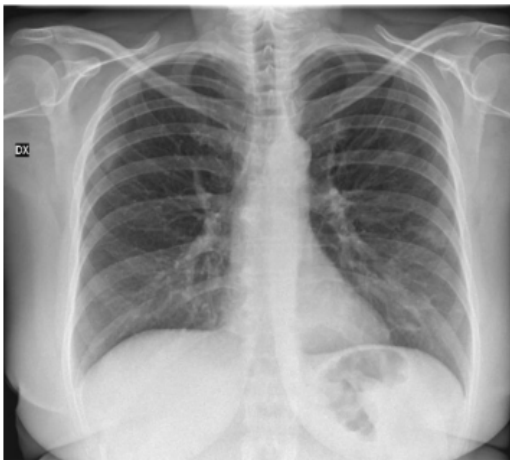


Heatmap of xray just to show how covid Xray looks

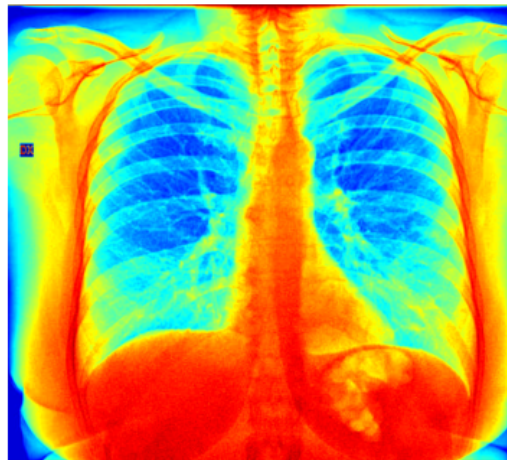
In [5]:

```
1 import cv2
2 import numpy as np
3 import matplotlib.pyplot as plt
4
5 # Load the image
6 image_path = 'C:/Users/Shivam Ranshur/ANACONDA JUPYTER PROGRAMMING/Covid19-dataset/t
7 image = cv2.imread(image_path)
8
9 # Apply the colormap to the image
10 heatmap = cv2.applyColorMap(image, cv2.COLORMAP_JET)
11
12 # Convert the heatmap to RGB format
13 heatmap_rgb = cv2.cvtColor(heatmap, cv2.COLOR_BGR2RGB)
14
15 # Display the original image and the heatmap overlay
16 fig, axs = plt.subplots(1, 2, figsize=(10, 5))
17 axs[0].imshow(cv2.cvtColor(image, cv2.COLOR_BGR2RGB))
18 axs[0].set_title('Original Image')
19 axs[0].axis('off')
20 axs[1].imshow(heatmap_rgb)
21 axs[1].set_title('Heatmap Overlay')
22 axs[1].axis('off')
23
24 plt.show()
25
```

Original Image



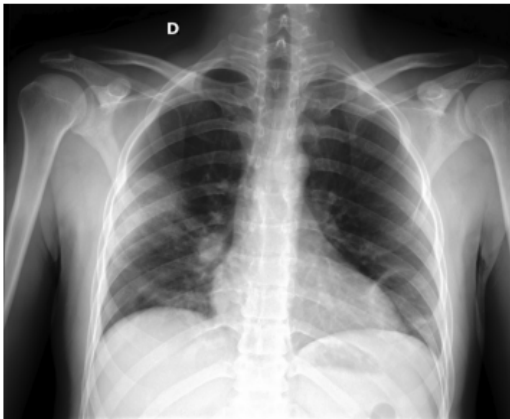
Heatmap Overlay



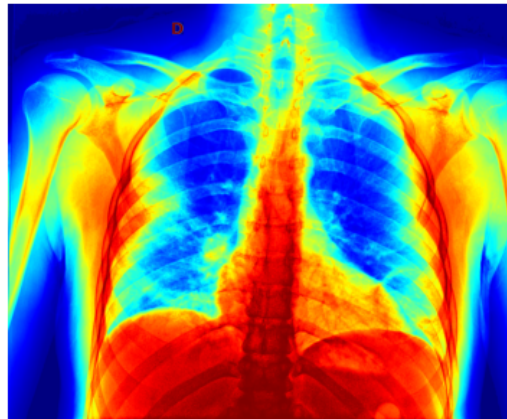
In [6]:

```
1 # Load the image
2 image_path = "C:/Users/Shivam Ranshur/ANACONDA JUPYTER PROGRAMMING/Covid19-dataset/t
3 image = cv2.imread(image_path)
4
5 # Apply the colormap to the image
6 heatmap = cv2.applyColorMap(image, cv2.COLORMAP_JET)
7
8 # Convert the heatmap to RGB format
9 heatmap_rgb = cv2.cvtColor(heatmap, cv2.COLOR_BGR2RGB)
10
11 # Display the original image and the heatmap overlay
12 fig, axs = plt.subplots(1, 2, figsize=(10, 5))
13 axs[0].imshow(cv2.cvtColor(image, cv2.COLOR_BGR2RGB))
14 axs[0].set_title('Original Image')
15 axs[0].axis('off')
16 axs[1].imshow(heatmap_rgb)
17 axs[1].set_title('Heatmap Overlay')
18 axs[1].axis('off')
19
20 plt.show()
21
```

Original Image



Heatmap Overlay



Data PreProcessing

In [7]:

```
1 # we have 251 images in training data which is very less for CNN so we need to do da
2 # Data augmentation is a technique to create more data using images we have currentl
3 # we'll do this on training data
```

Data Augmentation

In [9]:

```
1 # Data augmentation is use to generate more data using exixting data
2 train_data = "C:/Users/Shivam Ranshur/ANACONDA JUPYTER PROGRAMMING/Covid19-dataset/t
3 target_size = (256, 256)
4 batch_size = 16
5
6 train_datagen = ImageDataGenerator(
7     rescale=1 / 255,
8     rotation_range=0.2,
9     width_shift_range=0.1,
10    height_shift_range=0.1,
11    shear_range=0.2,
12    zoom_range=0.2,
13    horizontal_flip=True,
14    fill_mode='nearest'
15 )
```

In [10]:

```
1 train_generator = train_datagen.flow_from_directory(
2     train_data,
3     target_size=target_size,
4     batch_size=batch_size,
5     class_mode='categorical'
6 )
7
8 # Calculate the total number of samples in the generator
9 num_samples = len(train_generator) * batch_size
10 print("Total number of samples in the ImageDataGenerator:", num_samples)
```

Found 251 images belonging to 3 classes.

Total number of samples in the ImageDataGenerator: 256

Getting our labels

In [11]:

```
1 # other way is
2 train_generator.class_indices # we got our labels of data
```

Out[11]:

```
{'Covid': 0, 'Normal': 1, 'Viral Pneumonia': 2}
```

In [12]:

```
1 # doing same for test data
2 # we could have keep same data augmentation code for test and train both , but for e
3 test_data = "C:/Users/Shivam Ranshur/ANACONDA JUPYTER PROGRAMMING/Covid19-dataset/te
4 target_size = (256, 256)
5 batch_size = 16
6
7 test_datagen = ImageDataGenerator(
8     rescale=1 / 255,
9     rotation_range=0.2,
10    width_shift_range=0.1,
11    height_shift_range=0.1,
12    shear_range=0.2,
13    zoom_range=0.2,
14    horizontal_flip=True,
15    fill_mode='nearest'
16 )
17
18
19 test_generator = test_datagen.flow_from_directory(
20     test_data,
21     target_size=target_size,
22     batch_size=batch_size,
23     class_mode='categorical'
24 )
25
26 # Calculate the total number of samples in the generator
27 num_samples = len(test_generator) * batch_size
28 print("Total number of samples in the ImageDataGenerator:", num_samples)
```

Found 66 images belonging to 3 classes.

Total number of samples in the ImageDataGenerator: 80

In [13]:

```
1 test_generator.class_indices # labels for test data
```

Out[13]:

```
{'Covid': 0, 'Normal': 1, 'Viral Pneumonia': 2}
```

In []:

```
1
```

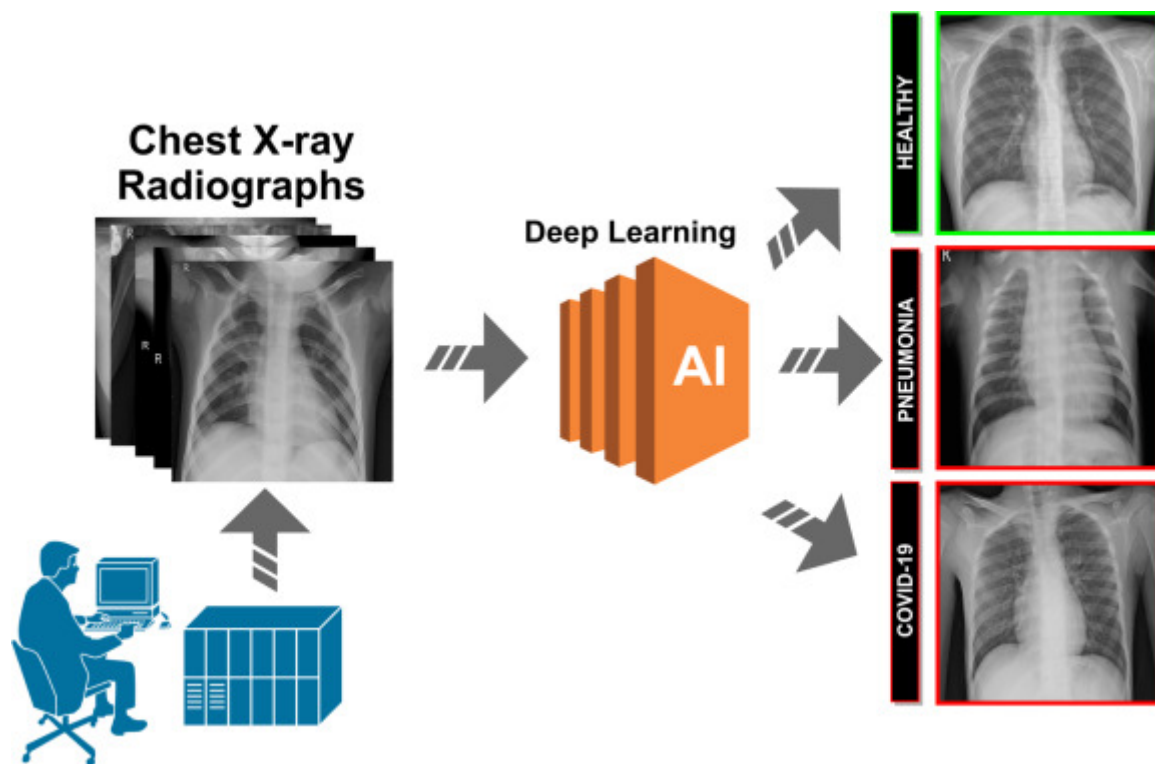
MODEL(CNN)

In [14]:

```

1 Xray_model = Sequential()
2
3 Xray_model.add(Conv2D(filters=32, kernel_size=(3,3), activation="relu",input_shape=(
4 Xray_model.add(MaxPool2D())
5 Xray_model.add(Dropout(rate=0.23))#Dropout is use to prevent overfitting
6
7 Xray_model.add(Conv2D(filters=64, kernel_size=(3,3), activation="relu"))
8 Xray_model.add(MaxPool2D())
9 Xray_model.add(Dropout(rate=0.25))
10
11 Xray_model.add(Conv2D(filters=128, kernel_size=(3,3), activation="relu"))
12 Xray_model.add(MaxPool2D())
13 Xray_model.add(Dropout(rate=0.25))
14
15 Xray_model.add(Flatten())
16 Xray_model.add(Dense(units = 64, activation="relu"))
17 Xray_model.add(Dropout(rate=0.40))
18 Xray_model.add(Dense(units=3, activation="softmax"))# 3 ints is used because we hav
19
20 Xray_model.compile(loss="categorical_crossentropy", optimizer= "adam",metrics= ["acc

```



Model summary

In [15]:

```
1 Xray_model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 254, 254, 32)	896
max_pooling2d (MaxPooling2D)	(None, 127, 127, 32)	0
dropout (Dropout)	(None, 127, 127, 32)	0
conv2d_1 (Conv2D)	(None, 125, 125, 64)	18496
max_pooling2d_1 (MaxPooling2D)	(None, 62, 62, 64)	0
dropout_1 (Dropout)	(None, 62, 62, 64)	0
conv2d_2 (Conv2D)	(None, 60, 60, 128)	73856

Model fitting

In [21]:

```
1 Xray_model.fit(train_generator, epochs=20, steps_per_epoch= 8, validation_steps=2)
```

```
Epoch 1/20
8/8 [=====] - 18s 2s/step - loss: 0.3742 - accuracy: 0.8594
Epoch 2/20
8/8 [=====] - 16s 2s/step - loss: 0.3279 - accuracy: 0.8750
Epoch 3/20
8/8 [=====] - 17s 2s/step - loss: 0.2922 - accuracy: 0.8750
Epoch 4/20
8/8 [=====] - 17s 2s/step - loss: 0.3036 - accuracy: 0.8906
Epoch 5/20
8/8 [=====] - 17s 2s/step - loss: 0.3289 - accuracy: 0.8537
Epoch 6/20
8/8 [=====] - 17s 2s/step - loss: 0.3425 - accuracy: 0.8984
Epoch 7/20
8/8 [=====] - 17s 2s/step - loss: 0.2139 - accuracy: 0.9106
Epoch 8/20
8/8 [=====] - 17s 2s/step - loss: 0.3018 - accuracy: 0.8780
Epoch 9/20
8/8 [=====] - 17s 2s/step - loss: 0.3151 - accuracy: 0.8906
Epoch 10/20
8/8 [=====] - 17s 2s/step - loss: 0.4197 - accuracy: 0.9024
Epoch 11/20
8/8 [=====] - 16s 2s/step - loss: 0.3732 - accuracy: 0.9106
Epoch 12/20
8/8 [=====] - 17s 2s/step - loss: 0.4382 - accuracy: 0.8374
Epoch 13/20
8/8 [=====] - 16s 2s/step - loss: 0.3878 - accuracy: 0.8699
Epoch 14/20
8/8 [=====] - 17s 2s/step - loss: 0.3197 - accuracy: 0.8984
Epoch 15/20
8/8 [=====] - 17s 2s/step - loss: 0.3241 - accuracy: 0.8594
Epoch 16/20
8/8 [=====] - 17s 2s/step - loss: 0.4894 - accuracy: 0.8438
Epoch 17/20
8/8 [=====] - 17s 2s/step - loss: 0.3203 - accuracy: 0.8862
Epoch 18/20
8/8 [=====] - 17s 2s/step - loss: 0.2996 - accuracy: 0.9268
Epoch 19/20
8/8 [=====] - 17s 2s/step - loss: 0.2674 - accuracy: 0.9024
Epoch 20/20
8/8 [=====] - 17s 2s/step - loss: 0.3468 - accuracy: 0.9106
```

Out[21]:

<keras.callbacks.History at 0x185494ec130>

In []:

```

1 # our model is giving final accuracy of 91.06
2 #which is fine
3 #we can optimize model further but it is not always necessary to complicate models
4 # if it is working perfectly consuming less computational power
5 # if while validating it dosent poerform well well optimize it further

```

In [22]:

```

1 y_pred = Xray_model.predict(test_generator)

```

5/5 [=====] - 5s 864ms/step

Validating Model

We'll do this by passing images of different categories

Labels

{'Covid': 0, 'Normal': 1, 'Viral Pneumonia': 2}

In [30]:

```

1 test_path= "C:/Users/Shivam Ranshur/ANACONDA JUPYTER PROGRAMMING/Covid19-dataset/tra
2 img= image.load_img(test_path, target_size=(256,256))
3 img = image.img_to_array(img)/255
4 img= np.array([img])
5 print(img.shape)

```

(1, 256, 256, 3)

In [32]:

```

1 # Predict the class probabilities for the image
2 class_probabilities = Xray_model.predict(img)
3
4 # Get the predicted class label
5 predicted_class = np.argmax(class_probabilities, axis=1)[0]
6
7 print("Class Probabilities:", class_probabilities)
8 print("Predicted Class:", predicted_class)

```

1/1 [=====] - 0s 276ms/step

Class Probabilities: [[0.98729765 0.01164208 0.00106034]]

Predicted Class: 0

In [34]:

```

1 # we can see name in path its normal
2 test_path= "C:/Users/Shivam Ranshur/ANACONDA JUPYTER PROGRAMMING/Covid19-dataset/tes
3 img= image.load_img(test_path, target_size=(256,256))
4 img = image.img_to_array(img)/255
5 img= np.array([img])
6 print(img.shape)
7
8 # Predict the class probabilities for the image
9 class_probabilities = Xray_model.predict(img)
10
11 # Get the predicted class label
12 predicted_class = np.argmax(class_probabilities, axis=1)[0]
13
14 print("Class Probabilities:", class_probabilities)
15 print("Predicted Class:", predicted_class)

```

(1, 256, 256, 3)

1/1 [=====] - 0s 28ms/step

Class Probabilities: [[4.7880650e-04 6.5907425e-01 3.4044698e-01]]

Predicted Class: 1

In [35]:

```

1 # we can see name in path its pneumonia
2 test_path= "C:/Users/Shivam Ranshur/ANACONDA JUPYTER PROGRAMMING/Covid19-dataset/tes
3 img= image.load_img(test_path, target_size=(256,256))
4 img = image.img_to_array(img)/255
5 img= np.array([img])
6 print(img.shape)
7
8 # Predict the class probabilities for the image
9 class_probabilities = Xray_model.predict(img)
10
11 # Get the predicted class label
12 predicted_class = np.argmax(class_probabilities, axis=1)[0]
13
14 print("Class Probabilities:", class_probabilities)
15 print("Predicted Class:", predicted_class)

```

(1, 256, 256, 3)

1/1 [=====] - 0s 26ms/step

Class Probabilities: [[0.00521721 0.38212913 0.6126536]]

Predicted Class: 2

For all 3 categories its working perfectly as we wanted its classifying perfectly

RESULT

Proposed CNN model achieves 91% accuracy in X-ray image classification, effectively detecting COVID-19, normal, and pneumonia cases with high precision.

Feel free to use this project and tune it further for your personal projects