

CLASSIFICATION CNN

STEPS -

- Create 3 folder in your desktop
- Training, Testing, Validation
- Inside training create 2 folder as h
- paste all the photo in testing part

```
In [1]: #pip install tensorflow
```

```
In [2]: import numpy as np
import matplotlib.pyplot as plt
import os
import cv2
import tensorflow as tf
from tensorflow.keras.preprocessing import image
from tensorflow.keras.preprocessing.image import ImageDataGenerator #it will
import warnings
warnings.filterwarnings('ignore')
```

```
In [3]: img= image.load_img(r"C:\Users\hp\Downloads\Doctor-with-female-patient.jpg")
```

```
In [4]: plt.imshow(img)
```

```
Out[4]: <matplotlib.image.AxesImage at 0x1cbc95c9b90>
```



```
In [5]: img_arr=cv2.imread(r"C:\Users\hp\Downloads\Doctor-with-female-patient.jpg")
img_arr
# 3 dimension metrics are cre
# the value ranges from 0-255
```

```
Out[5]: array([[226, 226, 220],
               [226, 226, 220],
               [226, 226, 220],
               ...,
               [230, 227, 219],
               [231, 228, 220],
               [231, 228, 220]],

              [[226, 226, 220],
               [226, 226, 220],
               [226, 226, 220],
               ...,
               [231, 228, 220],
               [232, 229, 221],
               [232, 229, 221]],

              [[226, 226, 220],
               [226, 226, 220],
               [226, 226, 220],
               ...,
               [232, 229, 221],
               [233, 230, 222],
               [233, 230, 222]],

              ...,

              [[209, 198, 214],
               [209, 198, 214],
               [210, 199, 215],
               ...,
               [ 90,  88, 100],
               [ 90,  88, 100],
               [ 90,  88, 100]],

              [[209, 199, 212],
               [209, 199, 212],
               [210, 200, 213],
               ...,
               [ 90,  88, 100],
               [ 90,  88, 100],
               [ 90,  88, 100]],

              [[209, 199, 211],
               [209, 199, 211],
               [210, 200, 212],
               ...,
               [ 90,  88, 100],
               [ 90,  88, 100],
               [ 90,  88, 100]]], dtype=uint8)
```

```
In [6]: img_arr.shape # height,width and RGB of image
```

```
Out[6]: (1333, 2000, 3)
```

```
In [7]: train=ImageDataGenerator(rescale= 1/255)
        validation= ImageDataGenerator(rescale =1/255)
        # to scale all the images i need to divide with 255
        # we need to resize the image using 200, 200 pixel
```

```
In [9]: train_dataset=train.flow_from_directory(r"C:\Users\hp\OneDrive\Documents\tra
        validation_dataset=validation.flow_from_directory(r"C:\Users\hp\OneDrive\Doc
                                                    batch_size=3,class_mode='b
```

Found 15 images belonging to 2 classes.
Found 6 images belonging to 2 classes.

```
In [10]: train_dataset.class_indices
```

```
Out[10]: {'healthy kidney': 0, 'stone kidney': 1}
```

```
In [11]: train_dataset.classes
```


```
Out[11]: array([0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1])
```

```
In [39]: # apply convention layer and maxpooling
        model= tf.keras.models.Sequential([tf.keras.layers.Conv2D(16,(3,3),activation='relu',
                                                                    tf.keras.layers.MaxPool2D(2,2),
                                                                    ##
                                                                    tf.keras.layers.Conv2D(32,(3,3),activation='relu',
                                                                    tf.keras.layers.MaxPool2D(2,2),
                                                                    ##
                                                                    tf.keras.layers.Conv2D(64,(3,3),activation='relu',
                                                                    tf.keras.layers.MaxPool2D(2,2),
                                                                    ##
                                                                    tf.keras.layers.Conv2D(128,(3,3),activation='relu',
                                                                    tf.keras.layers.MaxPool2D(2,2),
                                                                    ##
                                                                    tf.keras.layers.Flatten(),
                                                                    ##
                                                                    tf.keras.layers.Dense(512,activation='relu'),
                                                                    ##
                                                                    tf.keras.layers.Dense(1,activation='sigmoid')
                                                                    ])
```


```
In [40]: model.compile(loss='binary_crossentropy',optimizer=tf.keras.optimizers.RMSprop)
```

```
In [43]: model_fit=model.fit(train_dataset,epochs=20,validation_data=validation_data)
```


Epoch 1/20

5/5  0s 74ms/step - accuracy: 0.8269 - loss: 0.5532 - val_accuracy: 0.5000 - val_loss: 0.5332


Epoch 2/20

5/5  0s 60ms/step - accuracy: 0.5565 - loss: 0.5356 - val_accuracy: 0.6667 - val_loss: 0.5628


Epoch 3/20

5/5  0s 64ms/step - accuracy: 0.7491 - loss: 0.3989 - val_accuracy: 0.6667 - val_loss: 0.6140


Epoch 4/20

5/5  0s 63ms/step - accuracy: 0.8954 - loss: 0.3721 - val_accuracy: 0.6667 - val_loss: 0.7153


Epoch 5/20

5/5  0s 62ms/step - accuracy: 0.7130 - loss: 0.4208 - val_accuracy: 0.5000 - val_loss: 0.5793


Epoch 6/20

5/5  0s 62ms/step - accuracy: 0.7796 - loss: 0.3925 - val_accuracy: 0.6667 - val_loss: 0.9833


Epoch 7/20

5/5  0s 60ms/step - accuracy: 0.7963 - loss: 0.4835 - val_accuracy: 0.8333 - val_loss: 0.6861


Epoch 8/20

5/5  0s 63ms/step - accuracy: 0.8546 - loss: 0.2779 - val_accuracy: 0.6667 - val_loss: 0.8709


Epoch 9/20

5/5  0s 61ms/step - accuracy: 0.8593 - loss: 0.2864 - val_accuracy: 0.6667 - val_loss: 1.0154


Epoch 10/20

5/5  1s 71ms/step - accuracy: 1.0000 - loss: 0.1335 - val_accuracy: 0.8333 - val_loss: 1.4524


Epoch 11/20

5/5  0s 64ms/step - accuracy: 1.0000 - loss: 0.0782 - val_accuracy: 0.6667 - val_loss: 4.4398


Epoch 12/20

5/5  0s 62ms/step - accuracy: 0.7722 - loss: 0.6116 - val_accuracy: 0.6667 - val_loss: 0.6163


Epoch 13/20

5/5  0s 63ms/step - accuracy: 0.8398 - loss: 0.3572 - val_accuracy: 0.6667 - val_loss: 0.8430


Epoch 14/20

5/5  0s 62ms/step - accuracy: 0.9639 - loss: 0.2031 - val_accuracy: 0.6667 - val_loss: 1.2060


Epoch 15/20

5/5  0s 64ms/step - accuracy: 0.9176 - loss: 0.1421 - val_accuracy: 0.6667 - val_loss: 1.6190


Epoch 16/20

5/5  0s 66ms/step - accuracy: 1.0000 - loss: 0.0674 - val_accuracy: 0.8333 - val_loss: 2.1574


Epoch 17/20

5/5  0s 63ms/step - accuracy: 1.0000 - loss: 0.0752 - val_accuracy: 0.8333 - val_loss: 2.7059


Epoch 18/20

5/5  0s 62ms/step - accuracy: 1.0000 - loss: 0.0141 - val_accuracy: 0.8333 - val_loss: 3.2953

Epoch 19/20

5/5  0s 62ms/step - accuracy: 1.0000 - loss: 0.0052 - val_accuracy: 0.8333 - val_loss: 3.3406

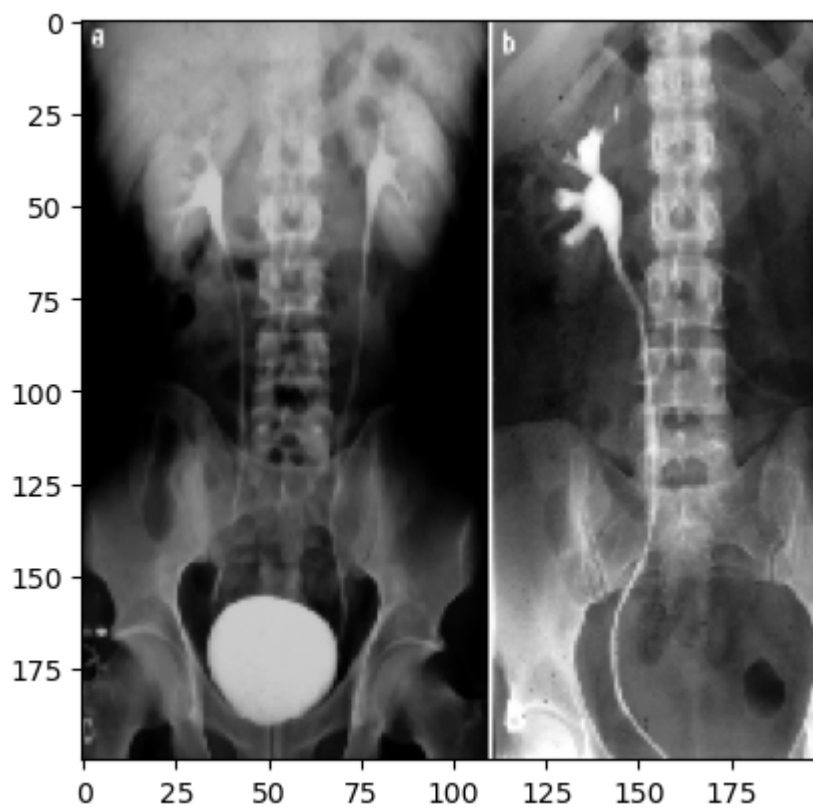
Epoch 20/20

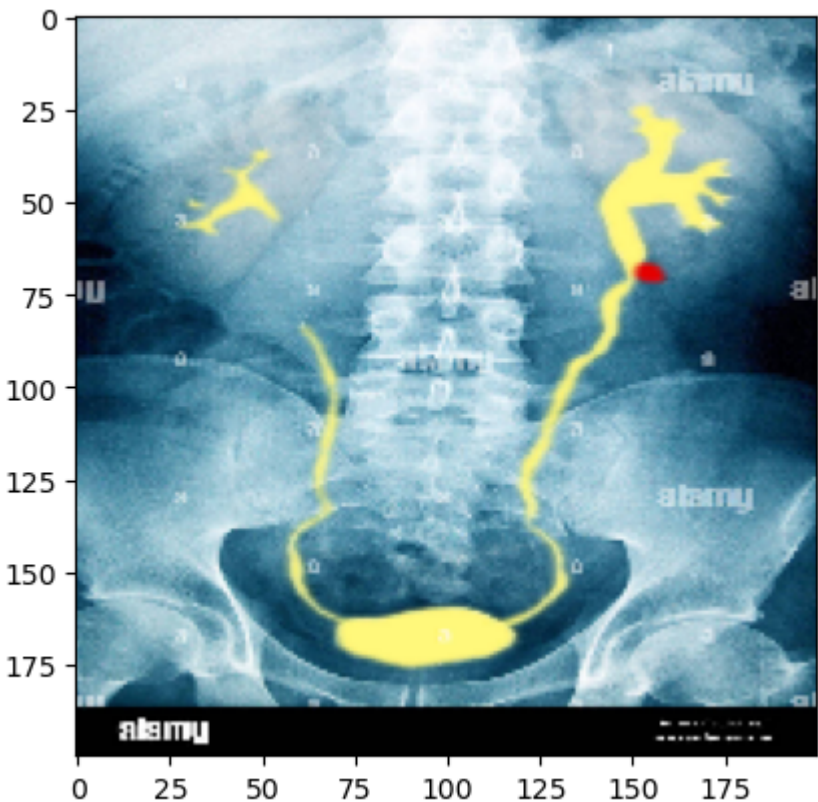
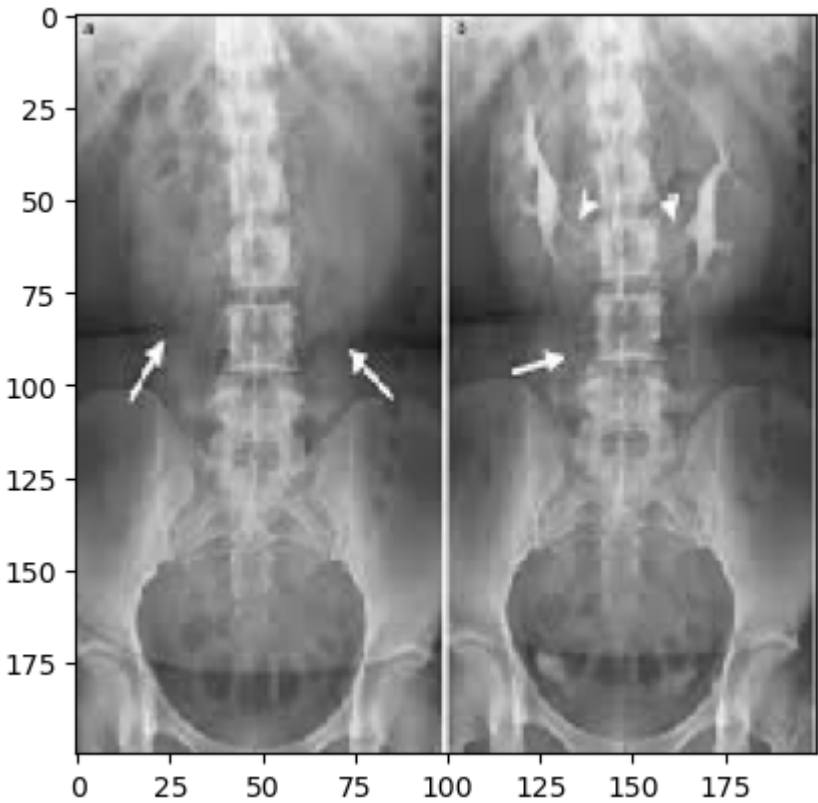
5/5  0s 64ms/step - accuracy: 1.0000 - loss: 0.0028 - val_accuracy: 0.8333 - val_loss: 3.4127

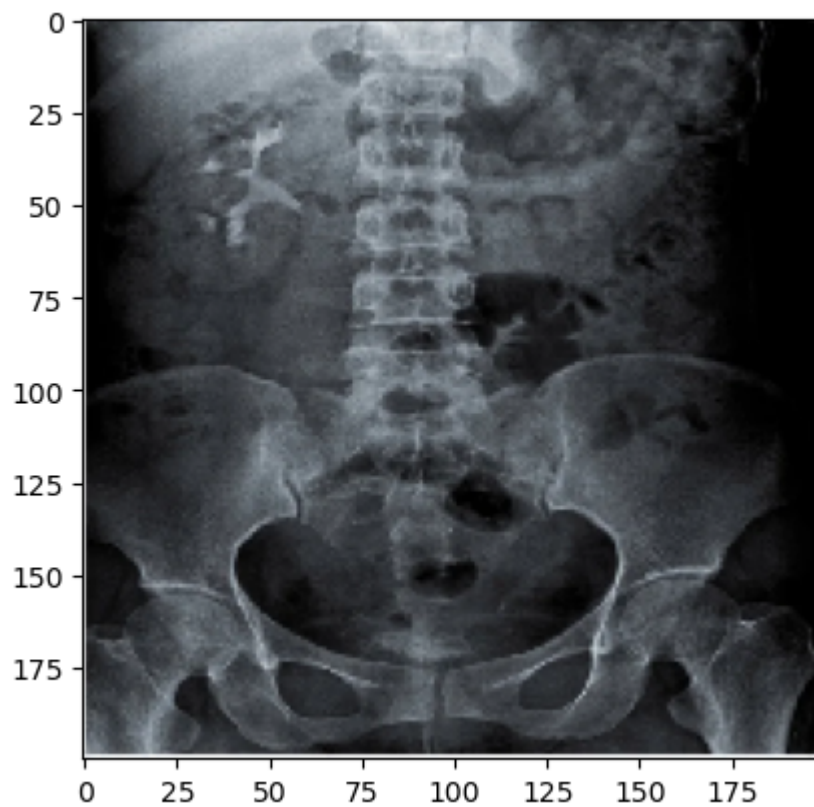
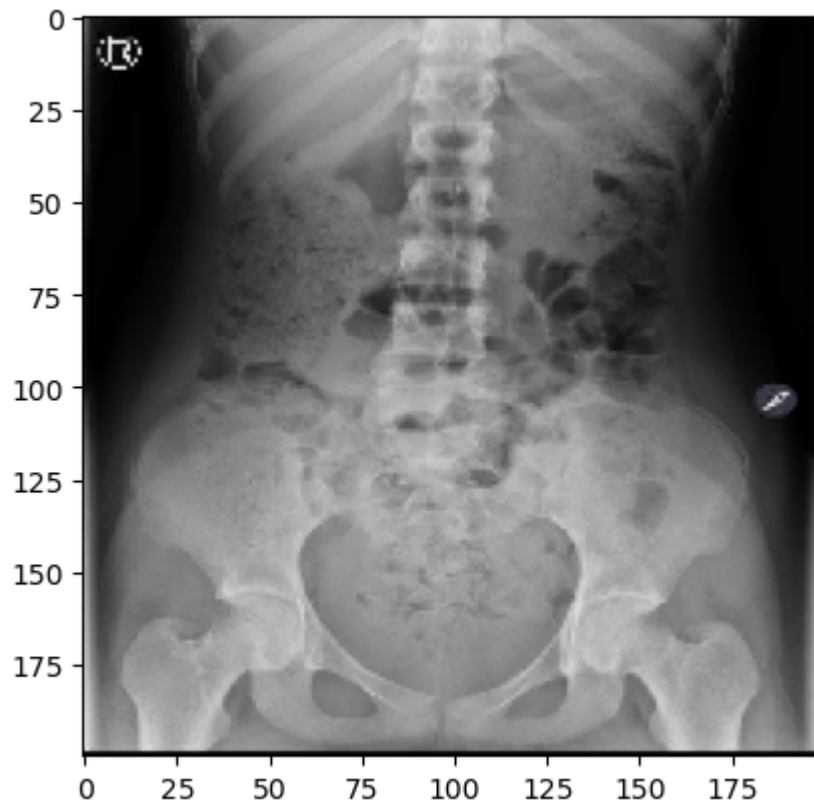
```
In [44]: dir_path=r"C:\Users\hp\OneDrive\Documents\testing"
for i in os.listdir(dir_path):
    print(i)
```

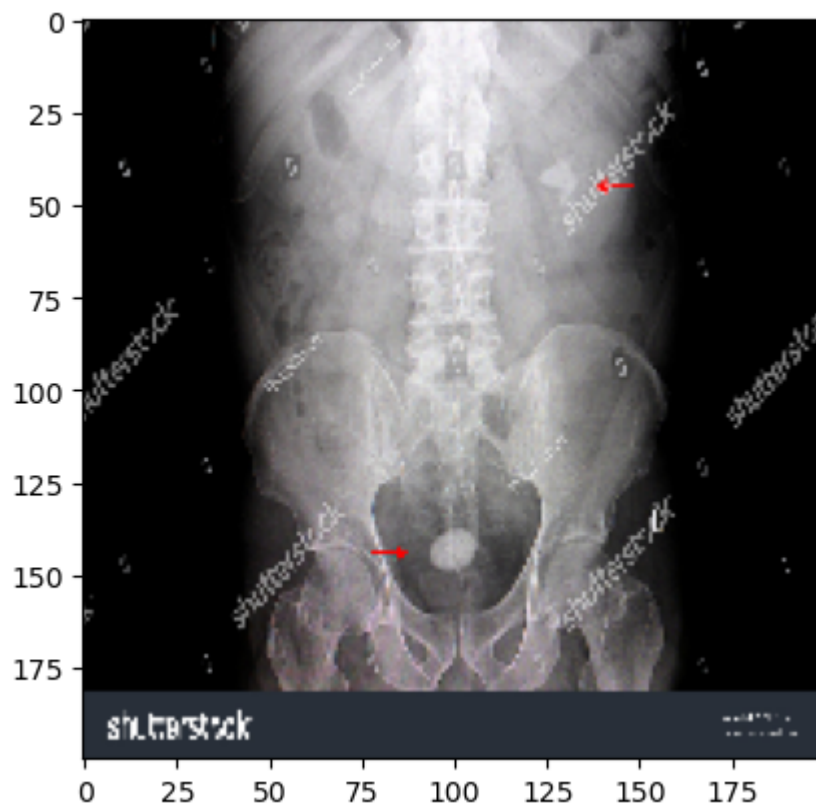
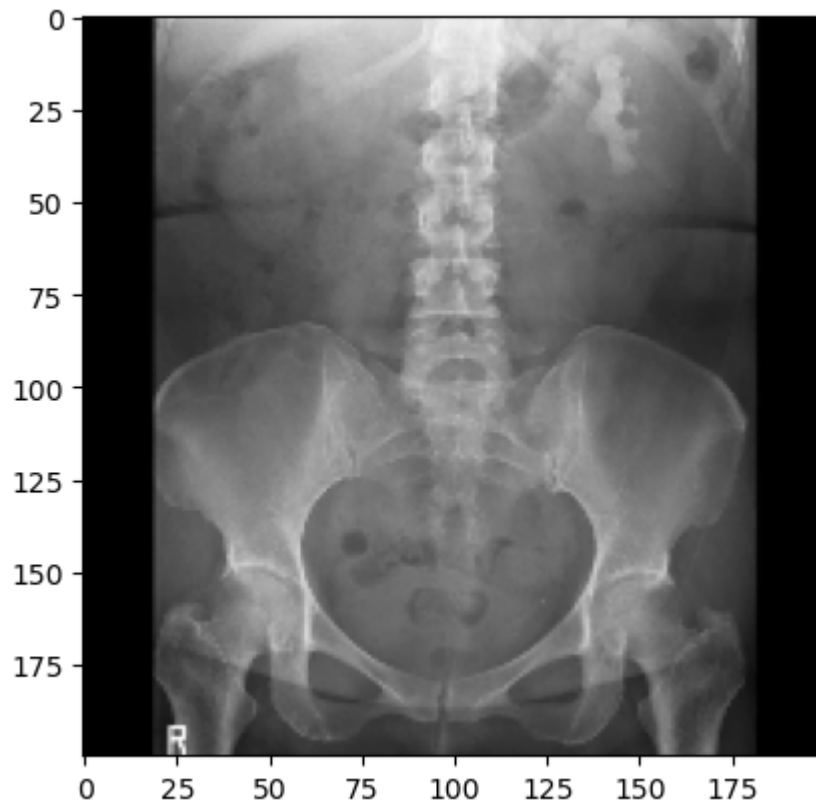
```
978-3-540-87597-0_2_Fig39_HTML.png
download (1).jpeg
download.jpeg
kidney-stone-x-ray-DWT8DT.jpg
Normal-Kidney-Ureter-and-Bladder-X-ray-with-no-radiopaque-foreign-body-see
n.png
shutterstock_632967740.webp
Staghorn_calc.gif
stock-photo-a-kub-xray-film-of-a-patient-with-a-large-urinary-bladder-stone
-and-multiple-left-kidney-stones-1924145921.jpg
x-ray-of-spinal-cord-and-normal-kidneys-HRF8GX.jpg
```

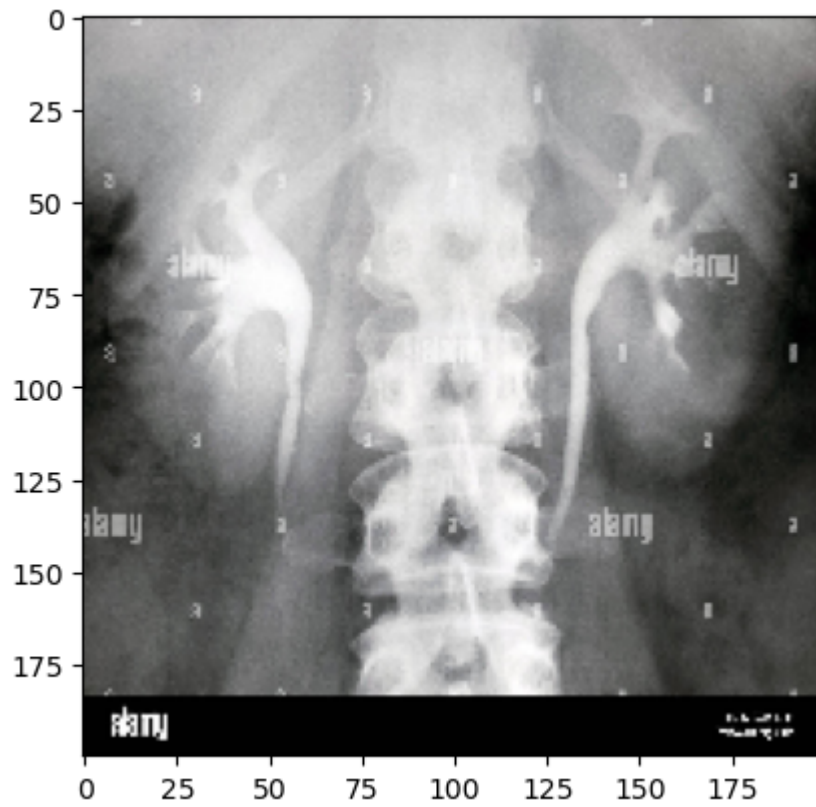
```
In [46]: dir_path=r"C:\Users\hp\OneDrive\Documents\testing"
for i in os.listdir(dir_path):
    img=image.load_img(dir_path+ '//' +i,target_size=(200,200))
    plt.imshow(img)
    plt.show()
```







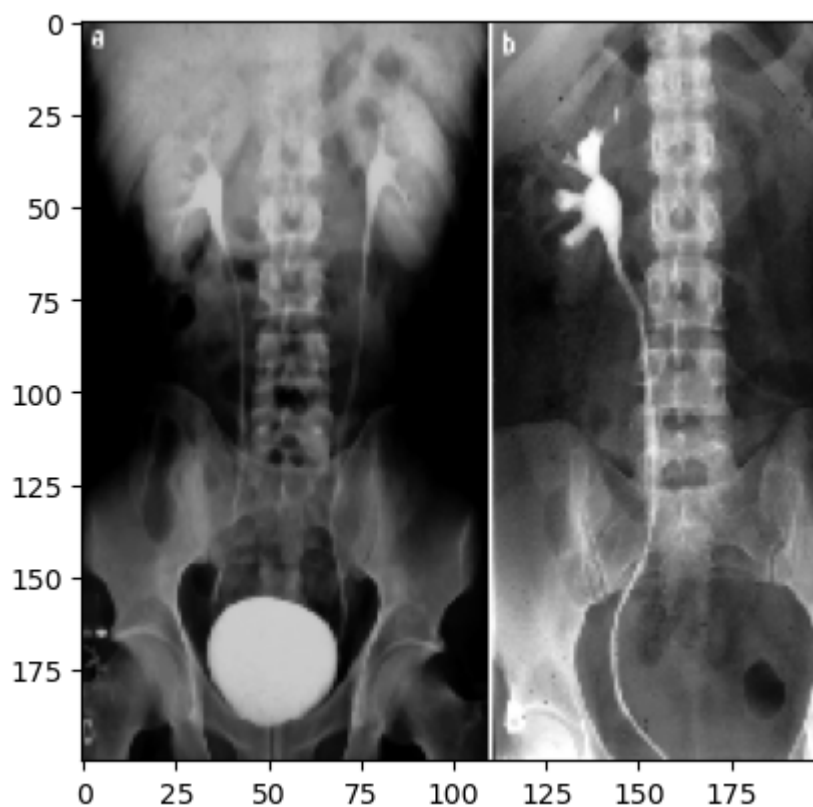




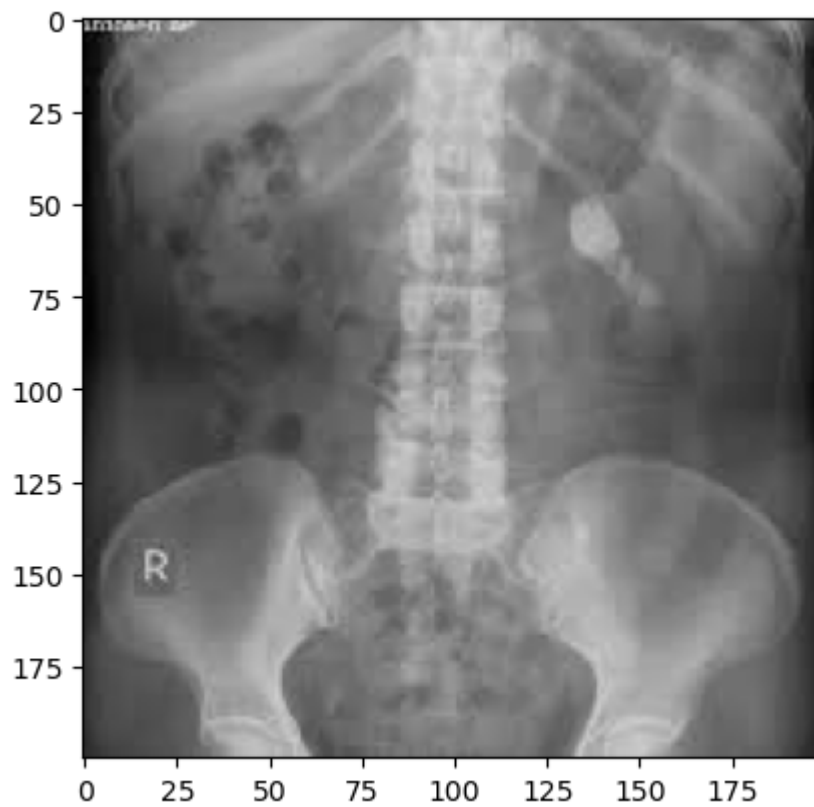
```
In [47]: dir_path=r"C:\Users\hp\OneDrive\Documents\testing"
for i in os.listdir(dir_path ):
    img = image.load_img(dir_path+ '//' +i, target_size = (200,200))
    plt.imshow(img)
    plt.show()

    x= image.img_to_array(img)
    x=np.expand_dims(x,axis = 0)
    images = np.vstack([x])

    val = model.predict(images)
    if val == 0:
        print( 'HEALTHY KIDNEY')
    else:
        print('STONE KIDNEY')
```

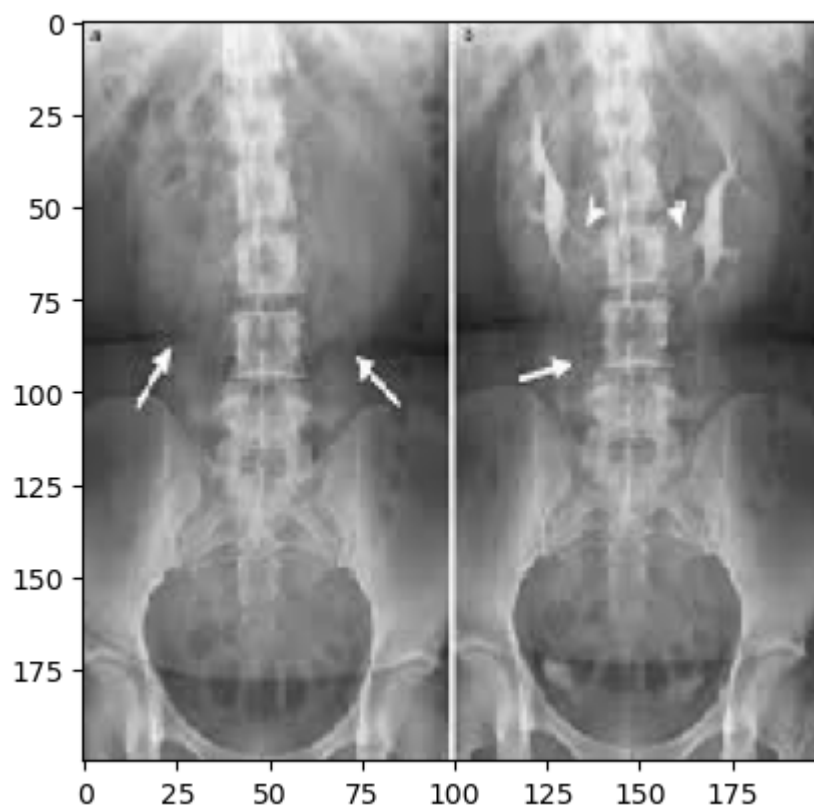


1/1 ————— 0s 114ms/step
HEALTHY KIDNEY



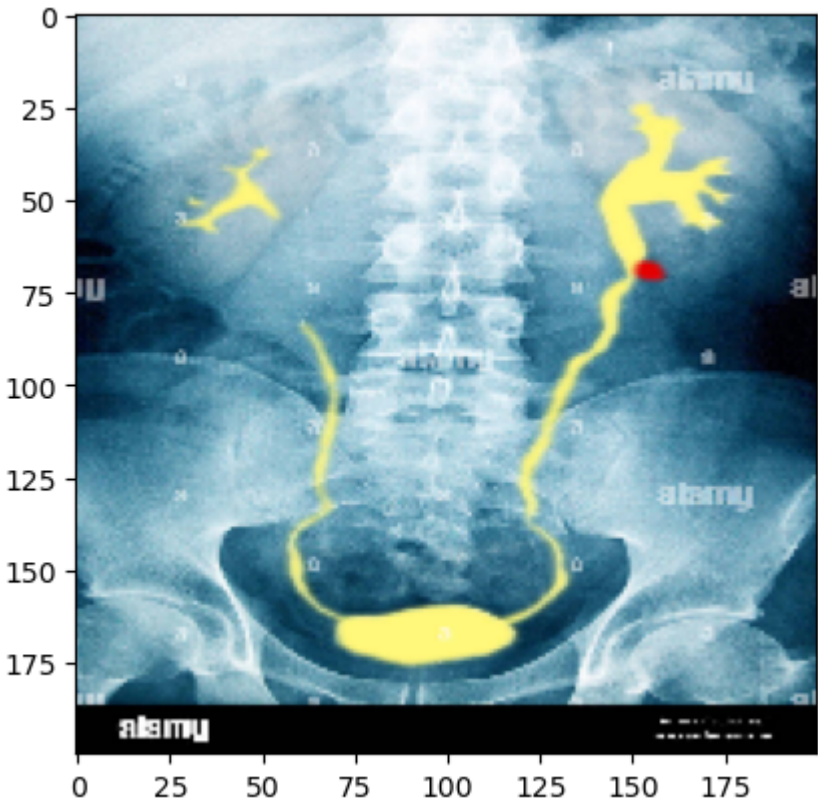
1/1 ————— 0s 26ms/step

HEALTHY KIDNEY

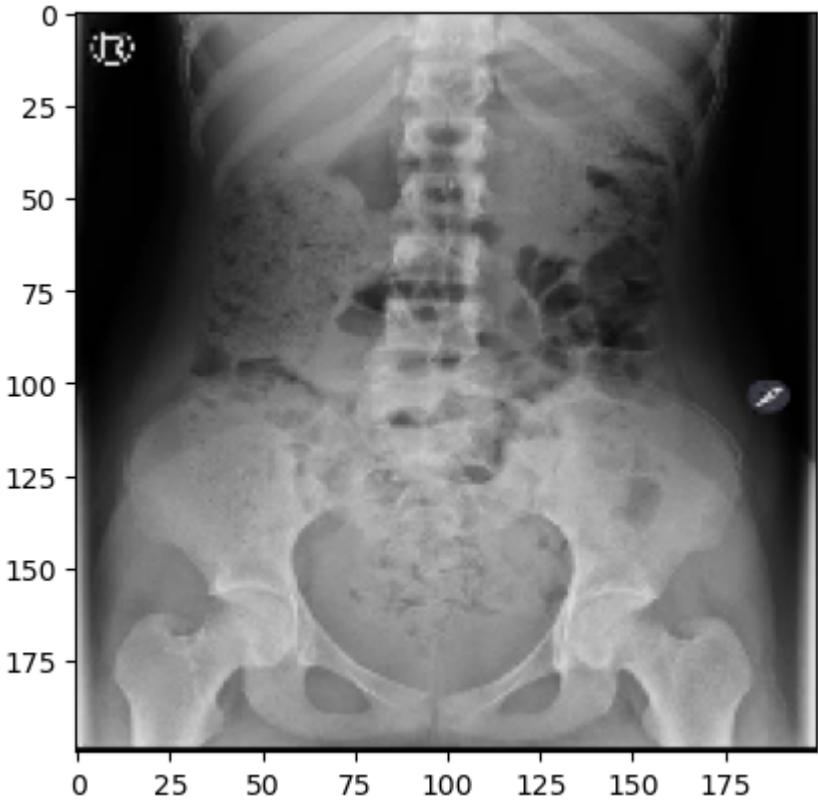


1/1 ————— 0s 26ms/step

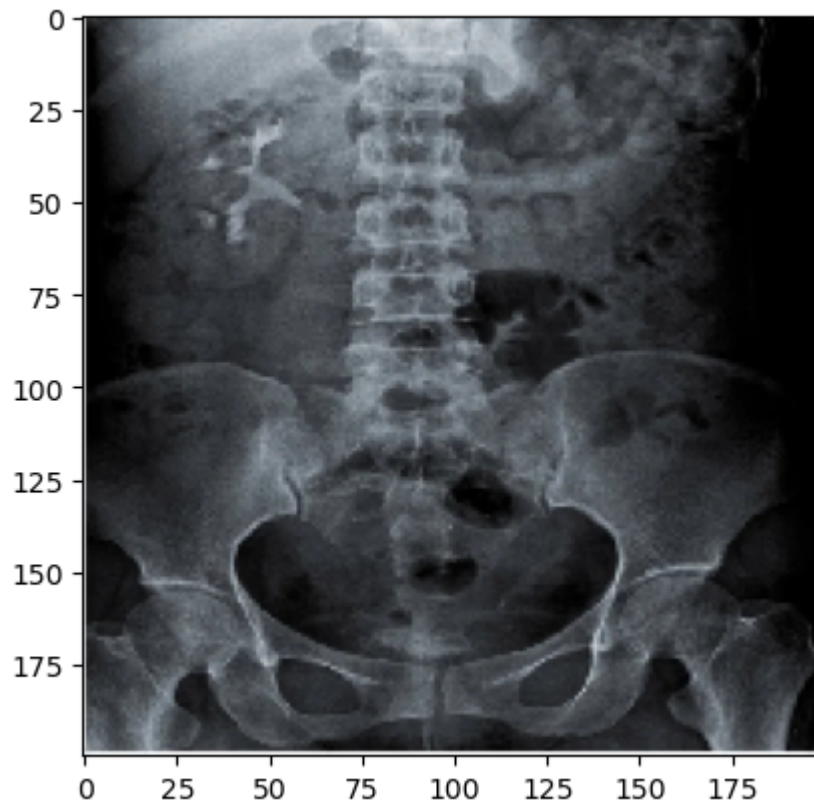
STONE KIDNEY



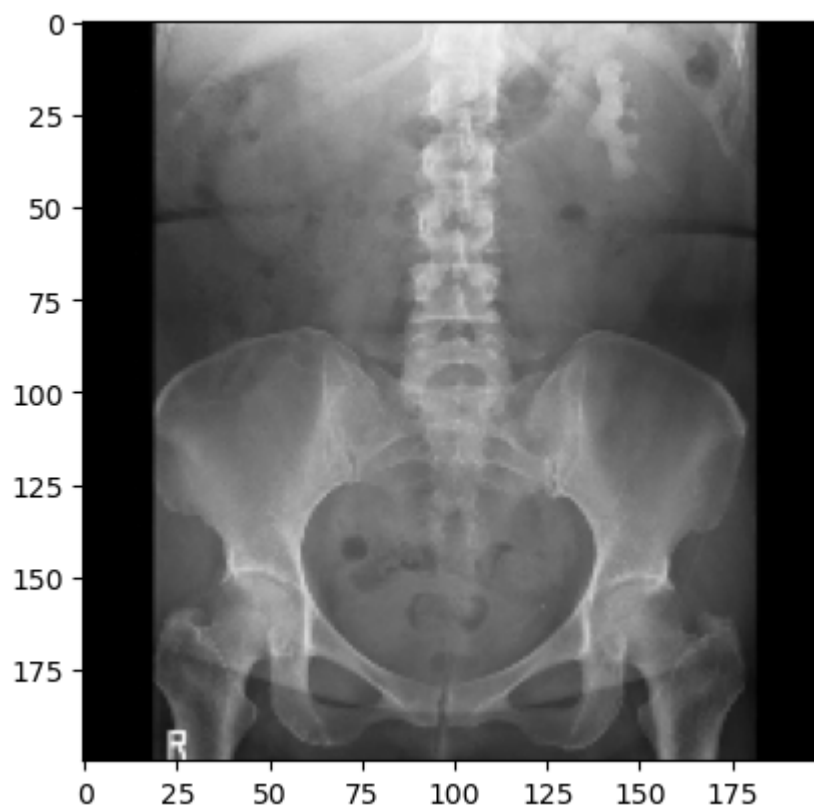
1/1 ————— 0s 33ms/step
STONE KIDNEY



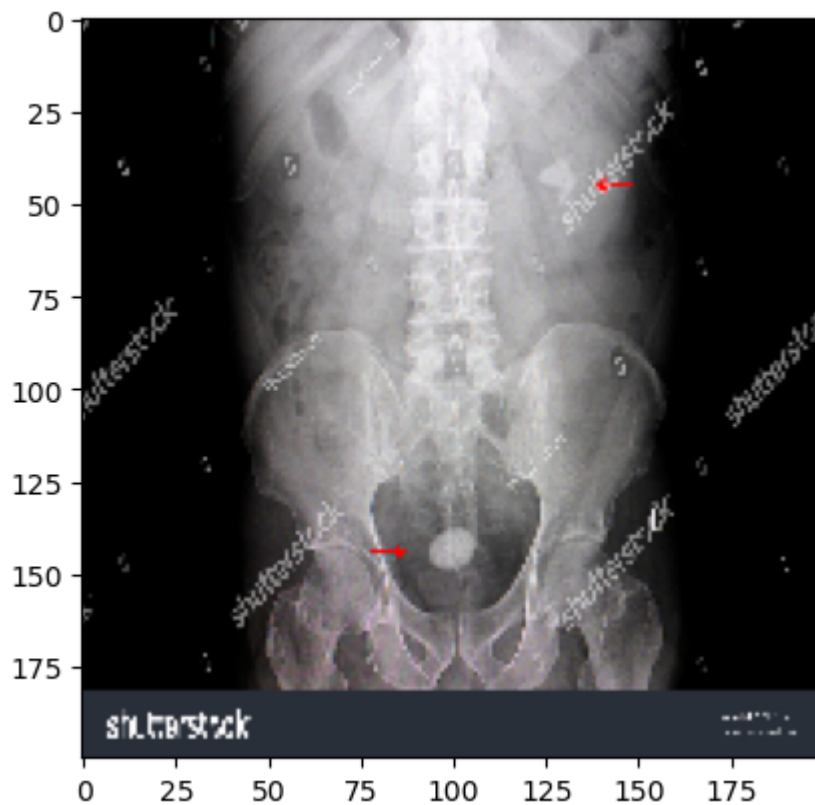
1/1 ————— 0s 41ms/step
HEALTHY KIDNEY



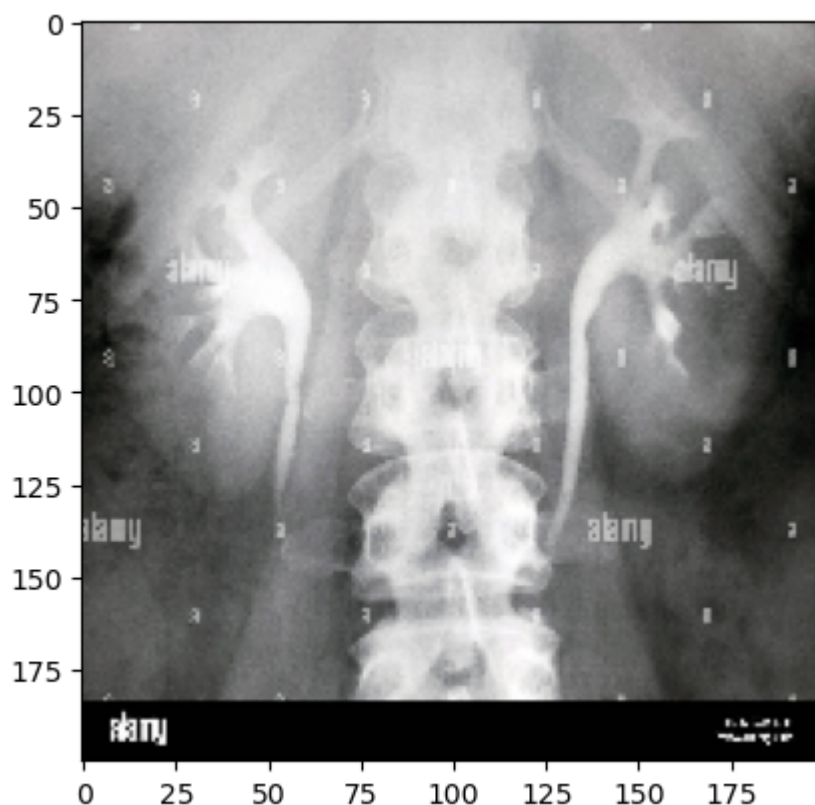
1/1 0s 29ms/step
STONE KIDNEY



1/1 0s 25ms/step
STONE KIDNEY



1/1 ————— 0s 29ms/step
STONE KIDNEY



1/1 ————— 0s 24ms/step
HEALTHY KIDNEY

training and testing of healthy kidney and stone kidneys xrays is done by binary classification of cnn, aquired accuracy 99%

In []: