

20BRS1143
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Google colab has been used to implement the program to plot facial landmarks:

```
import numpy as np
import pandas as pd
import dlib
import cv2
from google.colab.patches import cv2_imshow
import os
import random
from PIL import Image

df = pd.read_csv("/content/autism.csv")
num_classes = len(df["filepaths"].unique())
num_data = len(df)

print("Size of training data:", df.shape)
print("Number of unique classes:", num_classes)

Size of training data: (2938, 3)
Number of unique classes: 2938
```

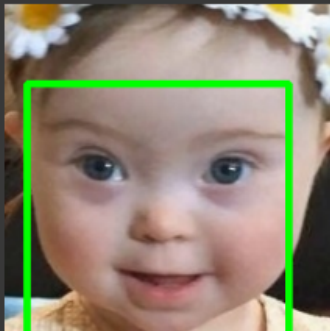
```
print(df.head(10))
```

	filepaths	labels	dataset
0	test/autistic/001.jpg	autistic	test
1	test/autistic/002.jpg	autistic	test
2	test/autistic/003.jpg	autistic	test
3	test/autistic/004.jpg	autistic	test
4	test/autistic/005.jpg	autistic	test
5	test/autistic/006.jpg	autistic	test
6	test/autistic/007.jpg	autistic	test
7	test/autistic/008.jpg	autistic	test
8	test/autistic/009.jpg	autistic	test
9	test/autistic/010.jpg	autistic	test

```
img = cv2.imread('/content/0022.jpg')
gray=cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
# Initialize dlib's face detector
detector = dlib.get_frontal_face_detector()
# Detecting faces in the grayscale image
faces = detector(gray)
print(faces)

rectangles[[15, 54] (194, 234)]

# Creating a for loop in order to extract
# specific coordinates (x1,x2,y1,y2)
for face in faces:
    x1=face.left()
    y1=face.top()
    x2=face.right()
    y2=face.bottom()
# Drawing a rectangle around the face
cv2.rectangle(img, (x1,y1), (x2,y2),(0,255,0),3)
cv2.imshow(img)
```



```

!wget http://dlib.net/files/shape_predictor_68_face_landmarks.dat.bz2
!bunzip2 "shape_predictor_68_face_landmarks.dat.bz2"

--2022-11-22 00:41:57-- http://dlib.net/files/shape_predictor_68_face_landmarks.dat.bz2
Resolving dlib.net (dlib.net)... 107.180.26.78
Connecting to dlib.net (dlib.net)|107.180.26.78|:80... connected.
HTTP request sent, awaiting response... 200 OK
Length: 64040097 (61M)
Saving to: 'shape_predictor_68_face_landmarks.dat.bz2'

shape_predictor_68_100%[=====>] 61.07M 17.7MB/s in 4.1s

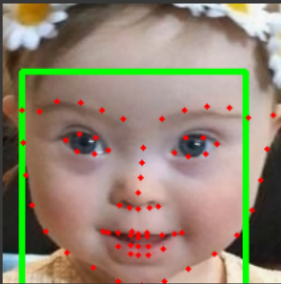
2022-11-22 00:42:01 (14.9 MB/s) - 'shape_predictor_68_face_landmarks.dat.bz2' saved [64040097/64040097]

bunzip2: Output file shape_predictor_68_face_landmarks.dat already exists.

p = "shape_predictor_68_face_landmarks.dat"
# Initialize dlib's shape predictor
predictor = dlib.shape_predictor(p)
# Get the shape using the predictor
landmarks=predictor(gray, face)

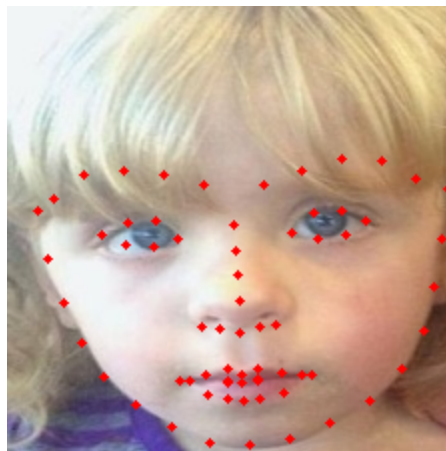
for n in range(0,68):
    x=landmarks.part(n).x
    y=landmarks.part(n).y
    cv2.circle(img, (x, y), 2, (0, 0, 255), -1)
cv2.imshow(img)

```

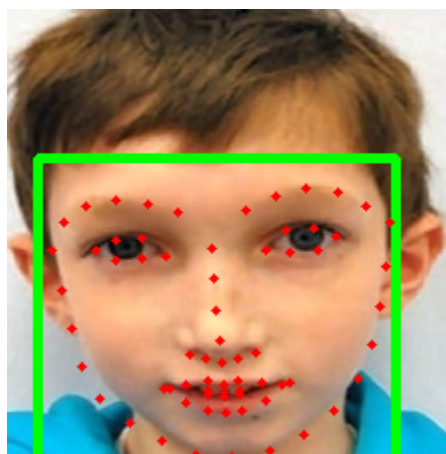


Data	Landmarks
0022	

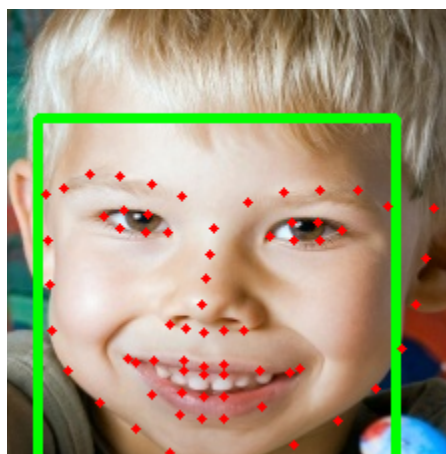
0095



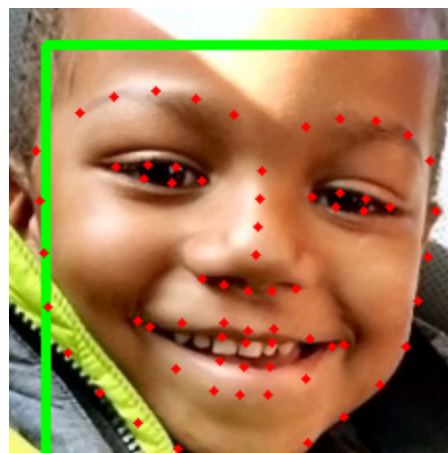
0069



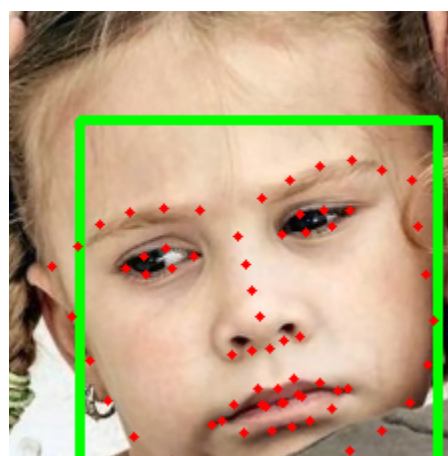
0237



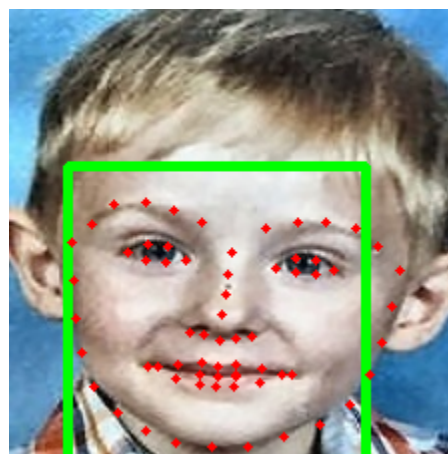
0301



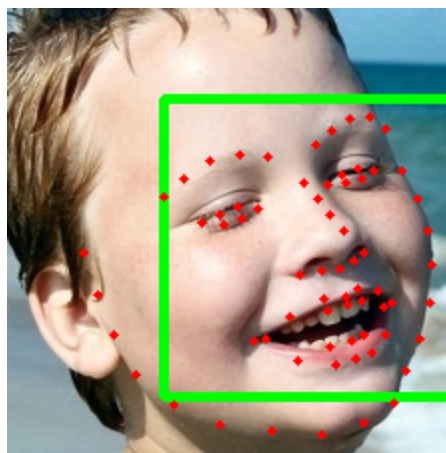
0323



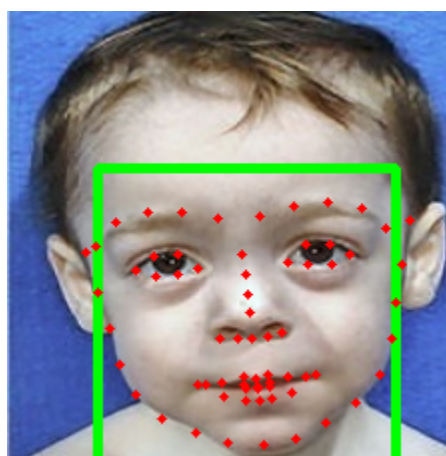
0154



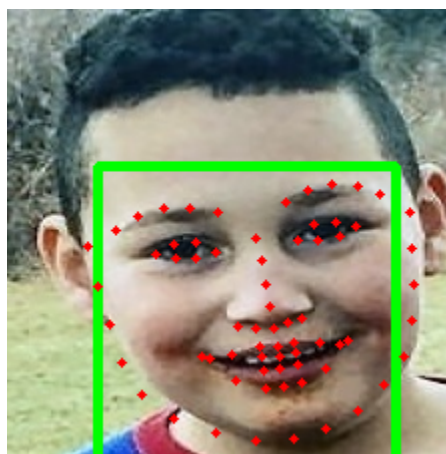
0276



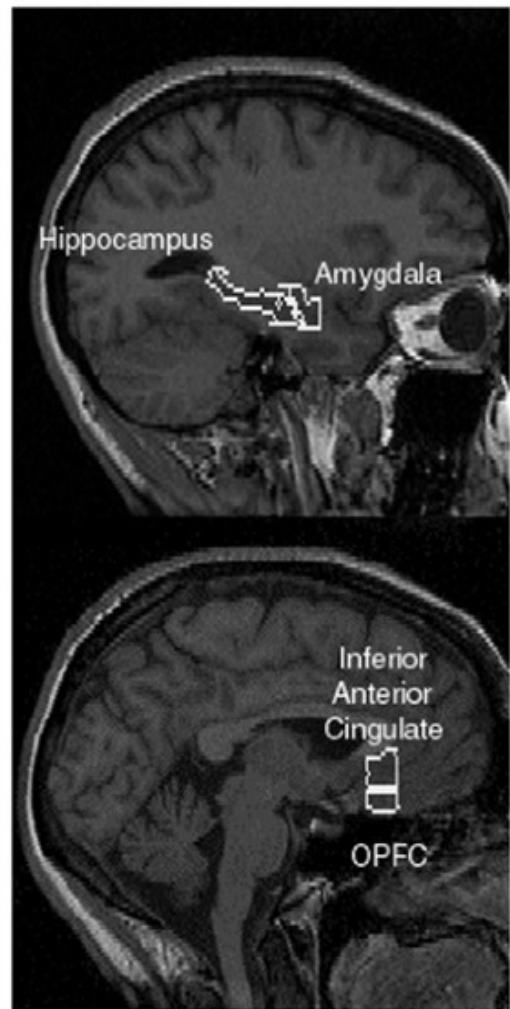
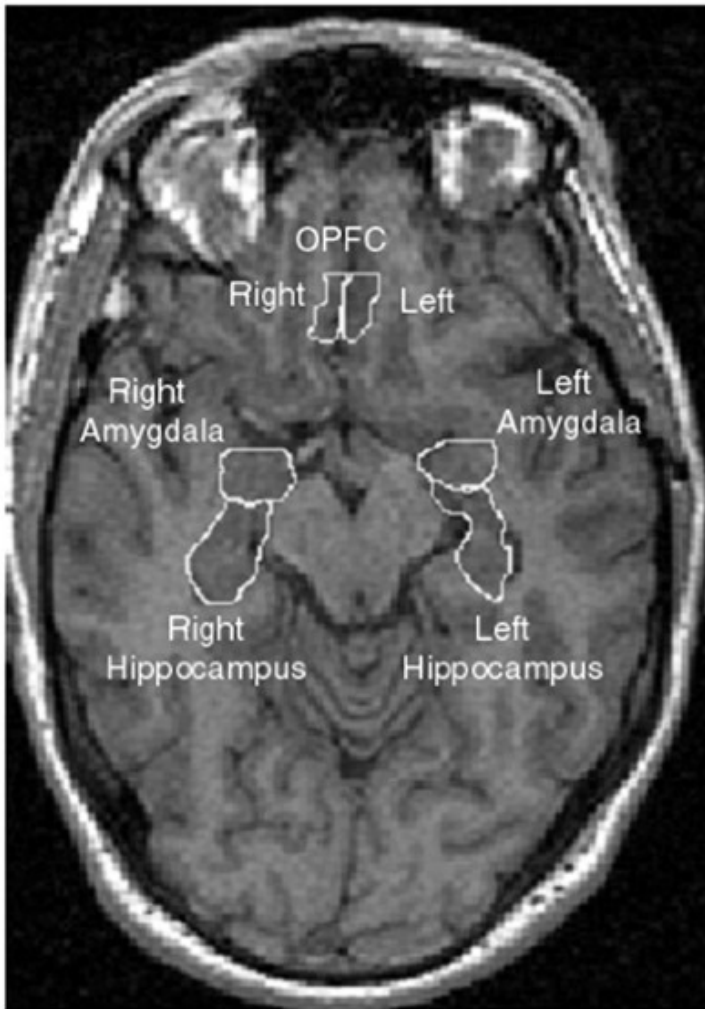
0177



0167

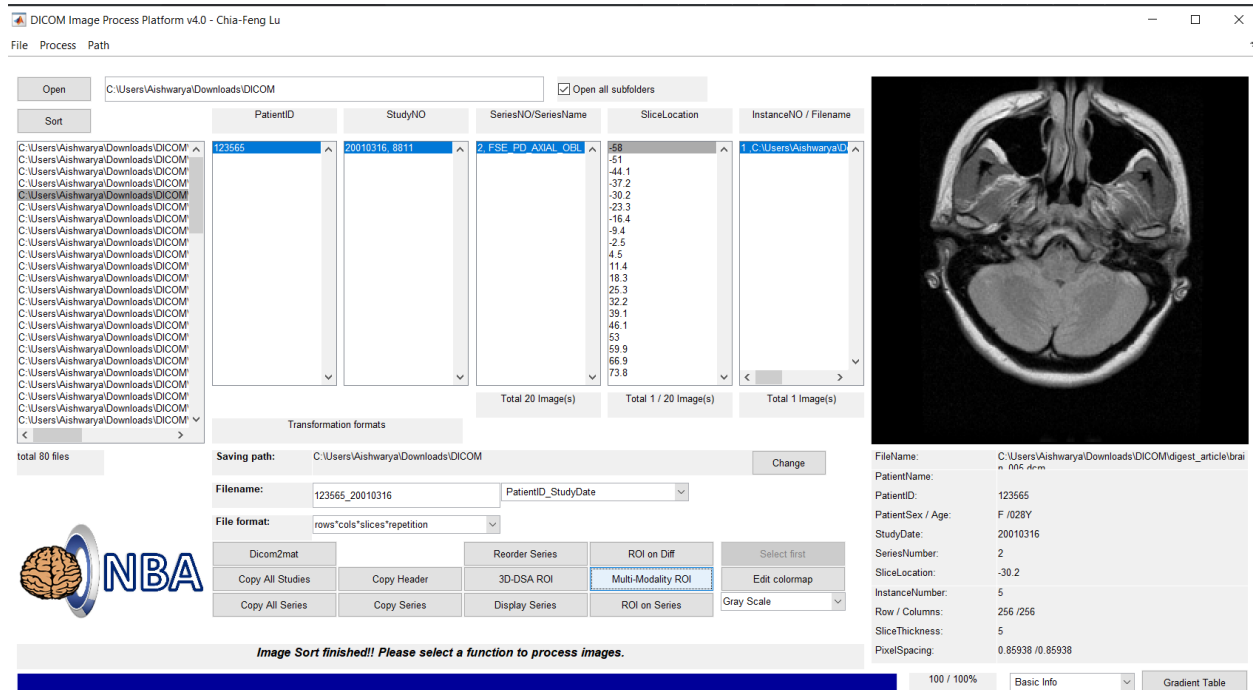


REF IMAGE FOR AMYGDALA



Multimodal Radiomics Platform (MRP) & Machine-Learning Models Platform is used to extract radiomics features.. This is supported by Matlab

PROCESS IN DICOM IMAGE PROCESSING



INTERPOLATION

Image interpolation occurs when you resize or distort your image from one pixel grid to another. Image resizing is necessary when you need to increase or decrease the total number of pixels, whereas remapping can occur when you are correcting for lens distortion or rotating an image.

ADC

Apparent Diffusion Coefficient (ADC)

The ADC value is automatically calculated by MR scanner by placing the smallest Region of Interest (ROI) on the area of interest.

Lower ADC values suggest a malignant high-grade astrocytoma, whereas higher ADCs suggest low-grade astrocytoma.

CROSS MODALITY COREGISTRATION

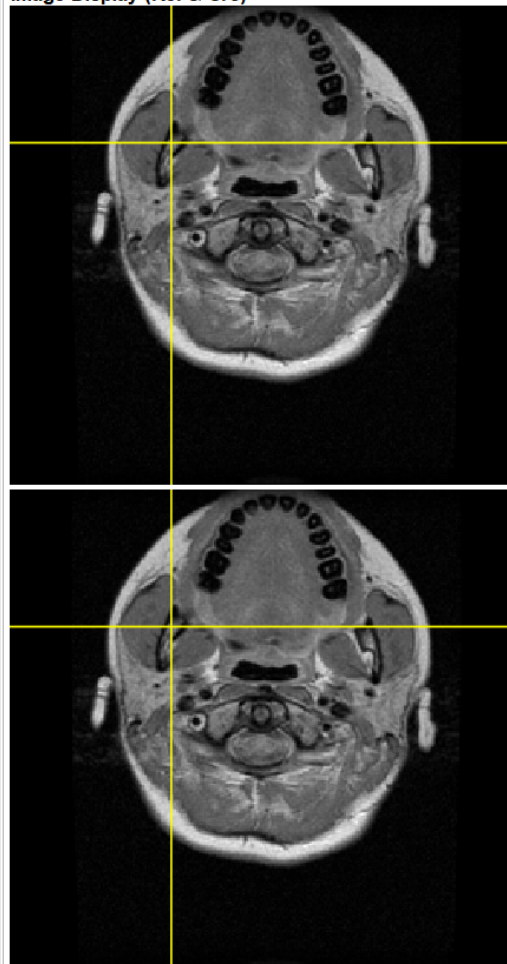
Image Selection (Patient ID: 123565)

StudyNO	SeriesNO/SeriesName
20010316_8811	2_FSE_PD_AXIAL_OBL

Reference Image	Source Images	Write Images
2_FSE_PD_AXIAL_O	2_FSE_PD_AXIAL_O	2_FSE_PD_AXIAL_O

C:\Users\Aishwarya\Downloads\DICOM\digest_article - Copy

Image Display (Ref & Src)



Control

Display

Raw

Resol. Adjust (mm)

X	0.85938
Y	0.85938
Z	2
Matrix	256

Coreg. Current

Coreg. All

Man. Adjust

X	0
Y	0
Z	0

Multi. ROI

☐ Skip Coreg.

☒ Normalize Sig.

☐ Use full Ref.

Selected Vol.

1

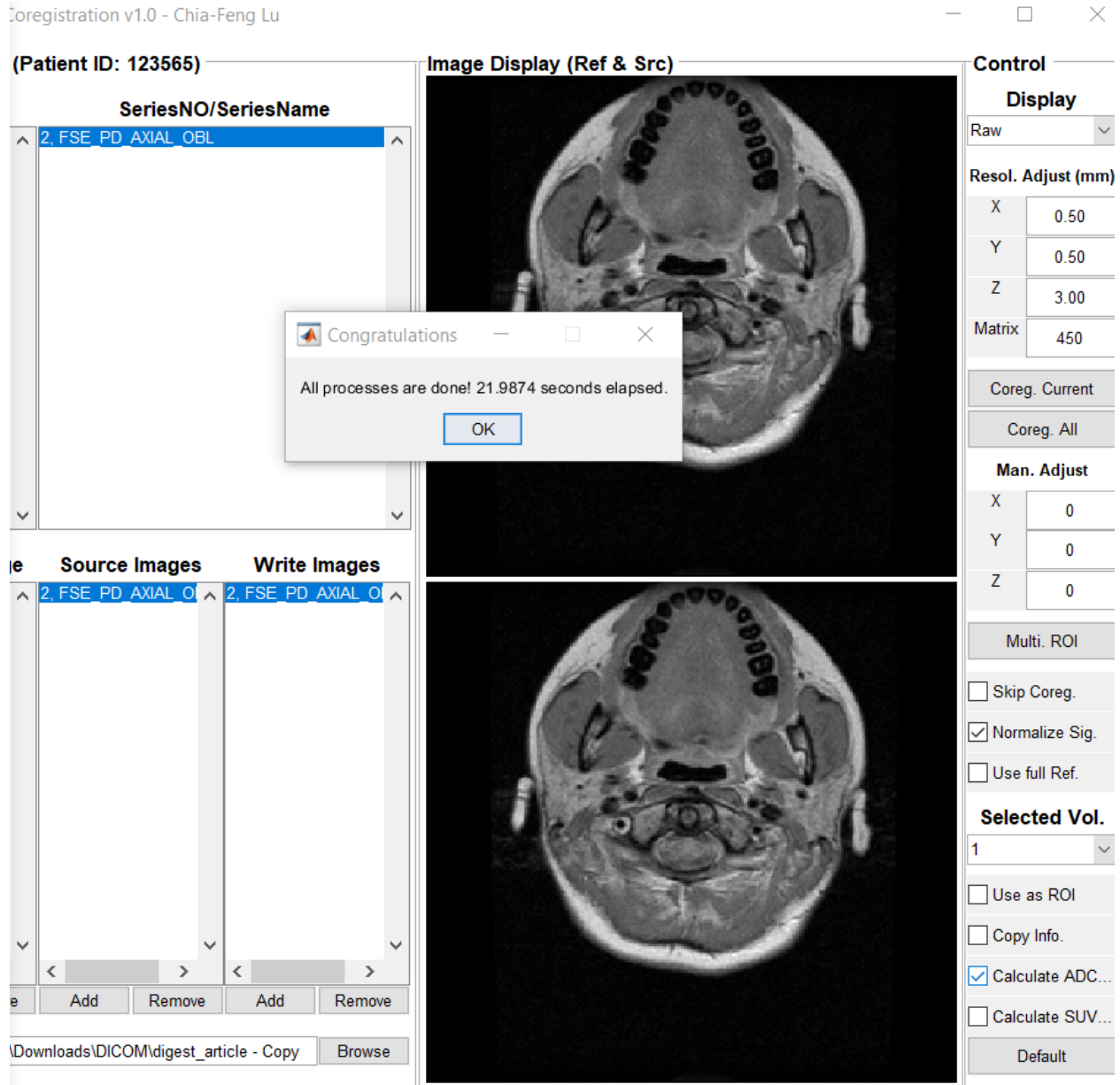
☐ Use as ROI

☐ Copy Info.

☐ Calculate ADC...

☐ Calculate SUV...

Default

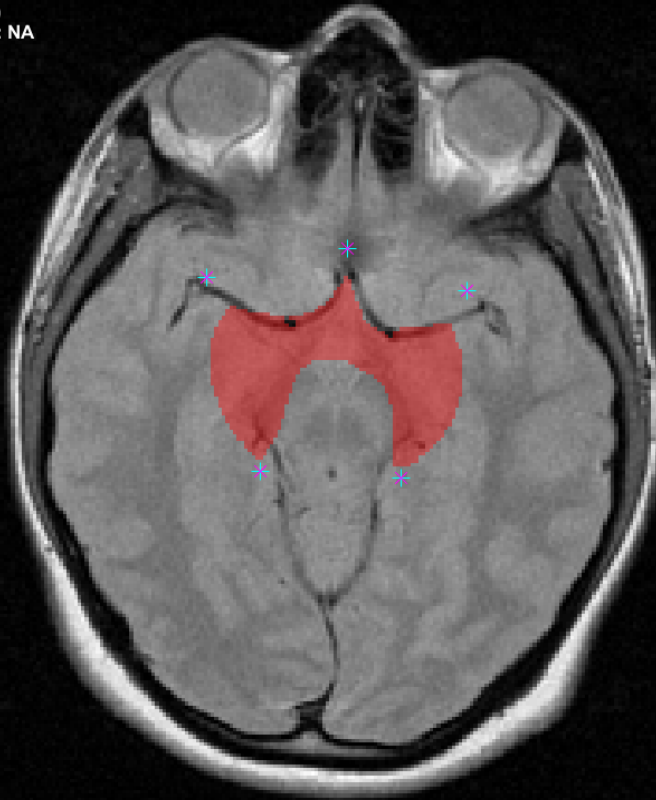


MULTIMODALITY ROI

Multimodal imaging or multiplexed imaging refers to simultaneous production of signals for more than one imaging technique. For example, one could combine using optical, magnetic, and radioactive reporters to be detected by SPECT, MRI, and PET. The goal of multimodal or multiplexed imaging is to improve early detection and localization of cancer

2_FSE_PD_AXIAL_OBL_ref.nii

Slice No.: 10/20
Slice Location: NA



Current Coordinate/ Intensity: 4,194,10/ 65.14

Selected ROI(s)
Pixel Number: 1745
Mean (Std): 1296.72 (200.47)
Area: 1288.74 mm², Volume: 2577.48 mm³

Zoom In Reset

ROI

New Remove

Slice 10, ROI 1

Brush Off 5

Copy Rename

Process

☐ Threshold Selected ...

300 Higher

Apply thres. Region Grow

+ - =

☒ Histogram Selected ...

Wavelet Decomp.

Feature Analysis

Show Results

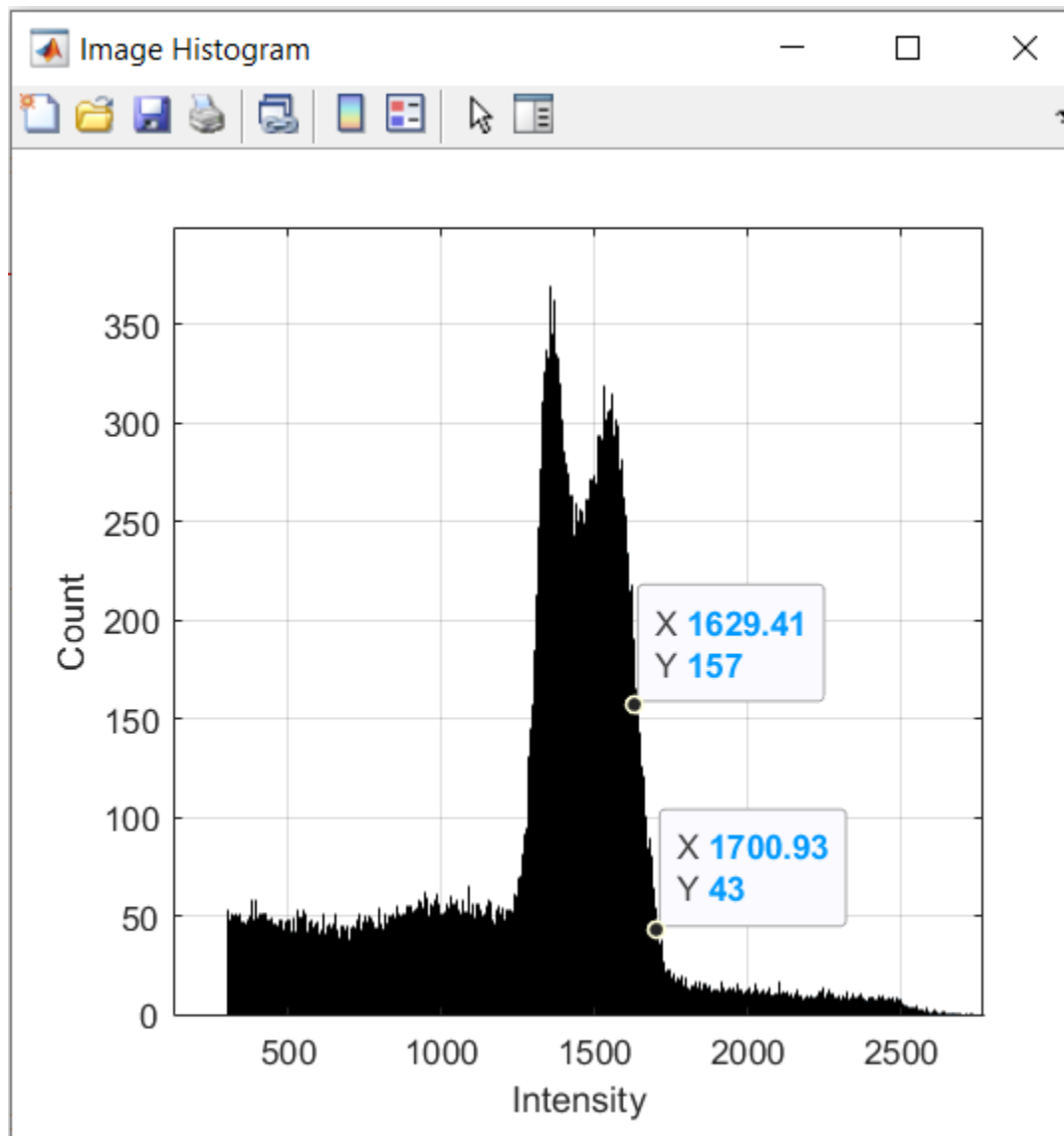
File I/O

Save Load

Export as Excel

Display Series

HISTOGRAM



RADIOMICS FEAUTURES:



Report of Radiomics Features of Lesion #1, Chia-Feng Lu

First order statistic

1. Energy:	3004270782.
2. Entropy:	7.7706
3. Kurtosis:	3.2395
4. Maximum:	1665.6298
5. Mean:	1296.7187
6. Mean absolute deviation:	147.095
7. Median:	1345.658
8. Minimum:	314.4885
9. Range:	