```
import nibabel as nib
import os
import numpy as np
import matplotlib.pyplot as plt
from scipy.ndimage import *
In [68]:
```

```
# MILESTONE 1: Loading images
nibimgs = []
nibhdrs = []
filenames = []
for file, subdir, d in os.walk(os.getcwd()):
   if 'analyze' in file:
        filenames.append(file)
        a = [x \text{ for } x \text{ in } d \text{ if 'img' in } x][0]
        imgpath = (os.path.join(file,a))
        nibimgs.append(imgpath)
        b = [x for x in d if 'hdr' in x][0]
        hdrpath = os.path.join(file,b)
        nibhdrs.append(hdrpath)
imgs = []
trainings nonsegs = []
trainings segs = []
validations nonsegs = []
validations_segs = []
testings nonsegs = []
testings segs = []
for file in nibimgs:
   img = nib.load(file)
    hdr = img.header
    data = img.get fdata()
    x, y, z = data.squeeze().shape
    middle coronal = np.rot90(data[:,:,z//2],3)
    imgs.append(middle coronal)
    if 'training' in file.lower():
        if 'segmentation' in file.lower():
            trainings segs.append(middle coronal)
        else:
            trainings nonsegs.append(middle coronal)
    elif 'testing' in file.lower():
        if 'segmentation' in file.lower():
            testings segs.append(middle coronal)
        else:
           testings nonsegs.append(middle coronal)
    elif 'validation' in file.lower():
        if 'segmentation' in file.lower():
            validations segs.append(middle coronal)
        else:
            validations_nonsegs.append(middle_coronal)
        raise ValueError ('Unidentifiable image: ',file)
```

In [69]:

```
def normalize(x):
    return x / np.max(x)

def geometric_transform(img, scale,theta,tx,ty,output_grid_shape):
    '''
    img (ndarray): 2-D img of floats
    scale (float): global expand/shrink, same positive factor applied on both axes
    theta (float): angle of rotation in degrees (ccw)
    tx (int): x-translation, assuming x is horizontal w.r.t user (right)
    ty (int): y-translation, assuming y is perpendicular to ground plane of user (downwards)
    '''

# if scale < 0:
        raise UserWarning ('Turning scale to positive')
    if np.max(img) >= 1:
        img = normalize(img)
```

```
shifted = shift(rotate(zoom(img.squeeze(),abs(scale)), theta) ,[ty,tx])
               ox, oy = shifted.shape[0]//2, shifted.shape[1]//2
               xdim = shifted.shape[0] - output_grid_shape[0]
               ydim = shifted.shape[1] - output grid shape[1]
               if xdim < 0:</pre>
                               xdim = abs(xdim)
                               shifted = np.pad(shifted,((int(np.ceil(xdim/2)),int(np.floor(xdim/2))),(0,0)),mode='constan
t',constant_values=0)
              if ydim < 0:</pre>
                               ydim = abs(ydim)
                                shifted = np.pad(shifted,((0,0),(int(np.ceil(ydim/2)),int(np.floor(ydim/2)))),mode='constan
t', constant values=0)
               \texttt{returned} = \texttt{shifted} [\texttt{max}(0, \texttt{ox} - \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2, \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid } \texttt{shape}[0] //2) : \texttt{max}(\texttt{ox} + \texttt{output} \texttt{ grid }
 _shape[0]), max(0,oy-output_grid_shape[1]//2):max(ox+output_grid_shape[1]//2,output_grid_shape[1])]
               if returned.shape[0] != output grid shape[0] or returned.shape[1] != output grid shape[1]:
                               print ('ERROR')
                               print (shifted.shape)
                              print (xdim, ydim)
                             print (returned.shape)
                              print (output_grid_shape)
               assert returned.shape[0] == output grid shape[0] and returned.shape[1] == output grid shape[1]
               return returned.shape, returned
```

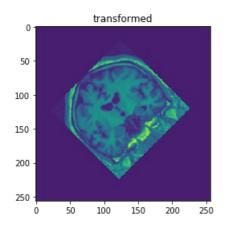
Test the Geometric Transform Works

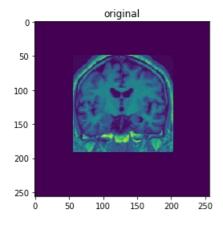
In [70]:

```
test = geometric_transform(imgs[0].squeeze(),scale=1,theta=45,tx=0,ty=0,output_grid_shape=(256,256)
)
plt.figure()
plt.title('transformed')
plt.imshow(test[1])
plt.figure()
plt.title('original')
plt.title('original')
plt.imshow(imgs[0].squeeze())
```

Out[70]:

<matplotlib.image.AxesImage at 0x259d9d08eb8>





```
In [71]:
def SSD(transform, fixed, moving):
        if not len(transform) == 4:
                raise InvalidArgumentError
        moving = moving.squeeze()
        transformed fixed =
geometric transform(fixed.squeeze(),transform[0],transform[1],transform[2],transform[3],fixed.sque
eze().shape)[1]
        return np.sum((transformed fixed - moving)**2)
from scipy.optimize import fmin,fmin_powell # uses simplex method, derivative-free
def optimize(fixed, moving, initial guess=None, **kwargs):
        if initial guess is None:
                initial guess = np.array([1.2,0.01,0.01,0.01])
        optimal = fmin powell(func=SSD,x0=initial guess,args=(fixed,moving),disp=True,maxiter=1e3,xtol=
1e-7)
           print ('\nOptimal parameters :')
        print (optimal)
        print ('='*60)
        return geometric_transform(fixed,optimal[0],optimal[1],optimal[2],optimal[3],fixed.shape)[1], o
ptimal
Test that the Registration Works
In [73]:
testscale = 1.2
testtheta = 40
testx = 15
testy = 25
testval = geometric transform(imgs[0], testscale, testtheta, testx, testy, imgs[0].shape)[1]
init = np.array([1,0,0,0])
optimal transform, params = optimize(imgs[0], testval, initial guess=init)
 \verb|C:\Users\hubh\Anaconda3\lib\site-packages\scipy\hdimage\interpolation.py:583: UserWarning: From some particles of the packages of the pac
cipy 0.13.0, the output shape of zoom() is calculated with round() instead of int() - for these in
puts the size of the returned array has changed.
    "the returned array has changed.", UserWarning)
Optimization terminated successfully.
                  Current function value: 0.000000
                  Iterations: 18
                  Function evaluations: 1740
[ 1.19730515 40.
                                        15.
                                                                         25.
   _____
In [74]:
plt.figure()
plt.xticks(np.arange(0,501,50))
plt.yticks(np.arange(0,501,50))
plt.title('Original Training Image')
plt.imshow(imgs[0].squeeze())
plt.figure()
plt.xticks(np.arange(0,501,50))
plt.yticks(np.arange(0,501,50))
plt.title('Validation Image')
plt.imshow(testval.squeeze())
plt.figure()
plt.xticks(np.arange(0,501,50))
```

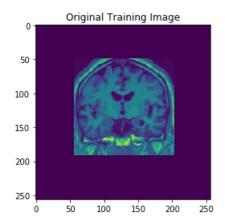
plt.yticks(np.arange(0,501,50))

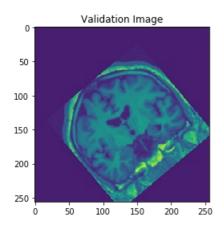
plt.imshow(optimal transform)

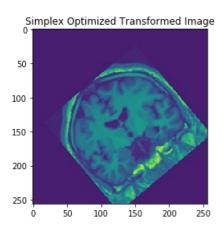
plt.title('Simplex Optimized Transformed Image')

Out[74]:

<matplotlib.image.AxesImage at 0x259d9cd36d8>







Training Image Registration

In [75]:

```
print (jaccard(np.ones((2,2)),np.ones((2,2)))) # sanity check
```

1.0

```
In [76]:
```

```
from scipy.interpolate import griddata
% matplotlib inline
init = np.array([1.5, 3, 1, 2])
mapping segmentations = {}
for i, moving in enumerate(trainings nonsegs):
         for j, fixed in enumerate(validations nonsegs):
                   optimal transform, params = optimize(normalize(moving), normalize(fixed), initial guess=init)
                    # Make gridpoints and interpolate between optimal transform and validation grid
                   xx, yy = np.meshgrid(np.arange(fixed.shape[0]),np.arange(fixed.shape[1]),indexing='ij')
                   fixedcoords = np.vstack([ xx.reshape(-1), yy.reshape(-1)]).T
                    \verb|xx|, yy = np.meshgrid(np.arange(optimal\_transform.shape[0]), np.arange(optimal\_transform.shape[0]), np.arange(optimal\_transform.sh
e[1]),indexing='ij')
                   optcoords = np.vstack([ xx.reshape(-1), yy.reshape(-1)]).T
                   a = griddata(points=optcoords, values=optimal transform.flatten(), xi=fixedcoords, method='nea
rest').reshape(fixed.squeeze().shape)
                   mapping segmentations[(i,j)] = [a, params] # store optimal transform in dict
                    # Plot results
                   f, axarr = plt.subplots(1,3,figsize=(20,20))
                   axarr[0].set title('Original Training %d' %i)
                   axarr[0].imshow(moving.squeeze())
                  axarr[1].set title('Validation %d' %j)
                  axarr[1].imshow(fixed.squeeze())
                        axarr[2].set title('Optimized Transformed')
                        axarr[2].imshow(optimal transform)
                   axarr[2].set title('Interpolated Optimized Transform of Training')
                   axarr[2].imshow(a)
                   plt.subplots_adjust(top=0.92, bottom=0.10, left=0.10, right=0.95, hspace=0.55,
                                                 wspace=0.35)
                   plt.show()
4
```

C:\Users\shubh\Anaconda3\lib\site-packages\scipy\ndimage\interpolation.py:583: UserWarning: From s cipy 0.13.0, the output shape of zoom() is calculated with round() instead of int() - for these in puts the size of the returned array has changed.

"the returned array has changed.", UserWarning)

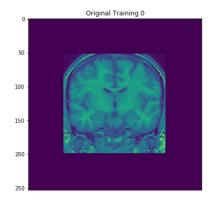
```
Optimization terminated successfully.

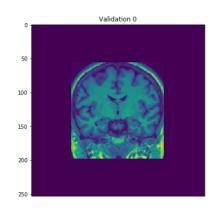
Current function value: 638.398791

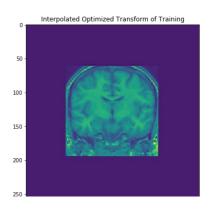
Iterations: 3

Function evaluations: 304

[-0.90732937 -0.12361573 -0.59113717 1.36386189]
```







0 50 100 150 200 250 0 50 100 150 200 250 0 50 100 150 200 2

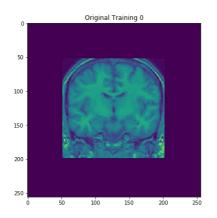
Optimization terminated successfully.

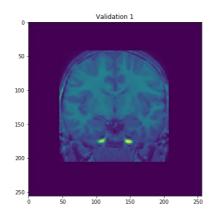
Current function value: 816.743813

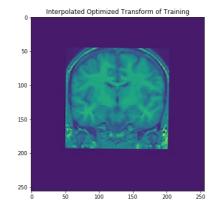
Iterations: 3

Function evaluations: 292

[-1.00918054 -0.78263784 -0.54330191 -6.43519495]







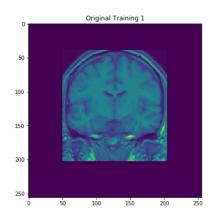
Optimization terminated successfully.

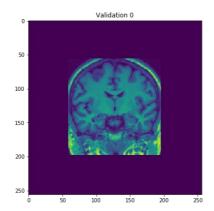
Current function value: 503.186184

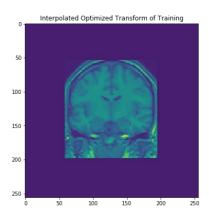
Iterations: 3

Function evaluations: 290

[-0.88387551 0.12690121 -0.11211549 3.93263445]







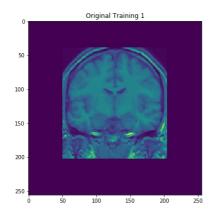
 ${\tt Optimization} \ {\tt terminated} \ {\tt successfully.}$

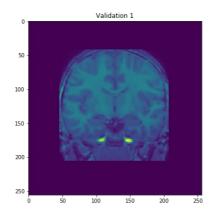
Current function value: 573.326429

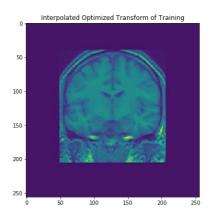
Iterations: 4

Function evaluations: 345

[-1.0109523 0.12263916 0.45808512 1.62029229]



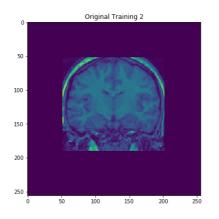


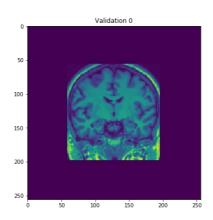


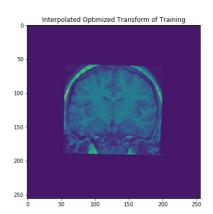
Optimization terminated successfully.

Current function value: 994.196153

Iterations: 4





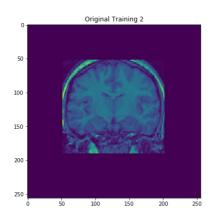


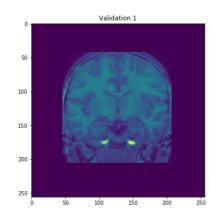
Optimization terminated successfully.

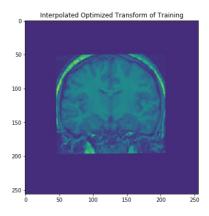
Current function value: 497.926150

Iterations: 4

Function evaluations: 363







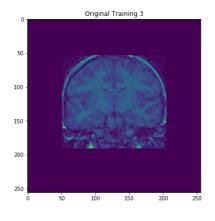
Optimization terminated successfully.

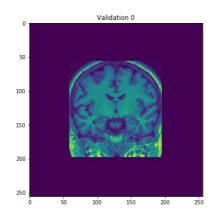
Current function value: 1100.492269

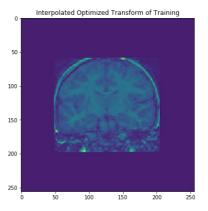
Iterations: 6

Function evaluations: 516

[1.00638212 0.4211337 -1.38065406 6.24448646]







Optimization terminated successfully.

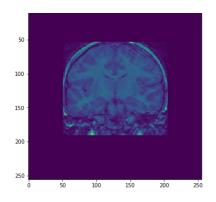
Current function value: 500.067294

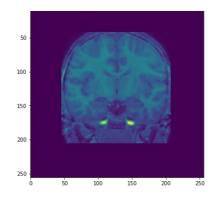
Iterations: 2

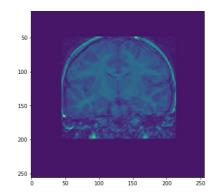
Function evaluations: 221

[1.08931193 0.10277284 1.24810472 2.440573]

Original Training 3 Validation 1







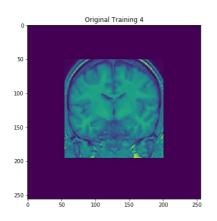
Optimization terminated successfully.

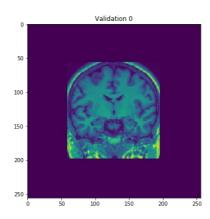
Current function value: 631.064037

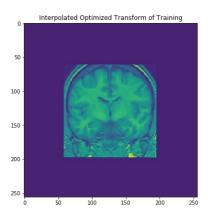
Iterations: 5

Function evaluations: 581

[-0.93857883 -0.11949087 -0.46513482 5.95412481]







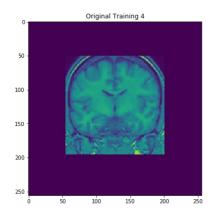
Optimization terminated successfully.

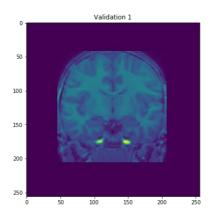
Current function value: 899.066795

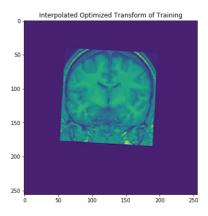
Iterations: 4

Function evaluations: 400

 $[\quad 0.93680478 \quad -3.32096431 \quad -3.47027598 \quad -11.52377725]$







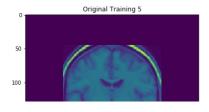
Optimization terminated successfully.

Current function value: 674.307341

Iterations: 7

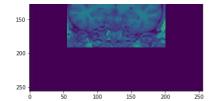
Function evaluations: 738

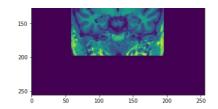
[0.94491849 -1.80440442 0.36242084 8.28426009]

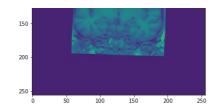












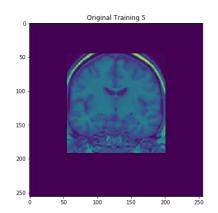
```
Optimization terminated successfully.

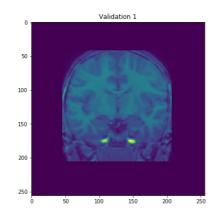
Current function value: 494.112812

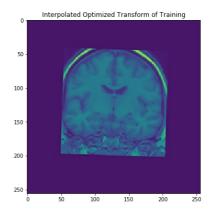
Iterations: 3

Function evaluations: 383

[ 1.06103368 -1.81969297 0.47006303 3.51593011]
```







Majority Voting Based Label Fusion

In [77]:

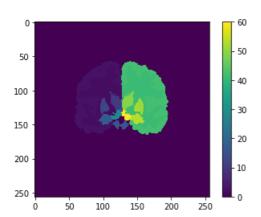
```
% matplotlib inline
# Majority based labeling
from scipy.stats import mode
# Ties broken by choosing smaller label (default scipy operation)

def majority_fusion(segs_list):
    votes = mode(np.dstack(segs_list),axis=2)
    return votes[0].squeeze()

training_majority_labs = majority_fusion(trainings_segs)
plt.imshow(training_majority_labs)
plt.colorbar()
```

Out[77]:

<matplotlib.colorbar.Colorbar at 0x259d85fbe48>



Jaccard Overlap Calculation for each Registration

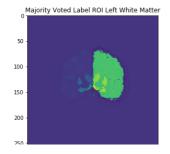
```
In [78]:
```

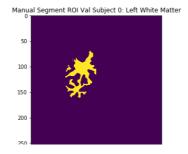
```
% matplotlib inline
table = []
label dict = {2:'Left White Matter', 3:'Left Cerebral Cortex', 41:'Right White Matter', 42:'Right Ce
rebral Cortex'}
for label in list(label dict.keys()):
   print ('Computing Jaccard Overlap for %s\n==========\n'%label dict
[label])
   for valid in range(len(validations segs)):
       manualcopy = np.copy(validations segs[valid].squeeze())
       autocopy = np.zeros like(manualcopy)
       segment transforms = []
       for mapping, automatic seg in mapping segmentations.items(): #mapping [0] is training, map
ping [1] is valid
           if mapping[1] == valid:
               params = automatic seg[1]
               p = np.max(trainings segs[mapping[0]]) # used for reversing effect of
normalization
               seg transform = p * geometric transform(trainings segs[mapping[0]].squeeze(),
                                                        params[0],
                                                        params[1].
                                                        params[2],
                                                        params[3],
                                                        trainings segs[mapping[0]].shape)[1].astype
p.float16)
                segment_transforms.append(seg_transform)
       voted = majority fusion(segment transforms).astype(np.int16)
       autocopy[np.where(voted == label)] = 1
       autocopy[np.where(autocopy != 1)] = 0
       manualcopy[np.where(manualcopy == label)] = 1
       \verb|manualcopy[np.where(manualcopy != 1)] = 0
       assert autocopy.shape == manualcopy.shape and manualcopy.shape == voted.shape
       f, axarr = plt.subplots(1,3,figsize=(20,20))
       axarr[0].set_title('Majority Voted Label ROI %s'%label_dict[label])
       s = axarr[0].imshow(voted)
       axarr[1].set title('Manual Segment ROI Val Subject %d: %s'%(valid, label dict[label]))
       axarr[1].imshow(manualcopy)
       axarr[2].set title('Automated Segment ROI %s'%label dict[label])
       axarr[2].imshow(autocopy)
       plt.subplots adjust(top=0.92, bottom=0.10, left=0.10, right=0.95, hspace=0.45,
                    wspace=0.95)
       plt.plot()
       plt.show()
        jac = jaccard(manualcopy,autocopy)
       print ('Jaccard Overlap: ', jac)
       manualcopy = np.zeros like(autocopy)
       table.append([valid,label_dict[label],jac]) # validation subjects x labels
4
```

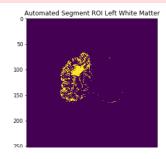
Computing Jaccard Overlap for Left White Matter

0 2

C:\Users\shubh\Anaconda3\lib\site-packages\scipy\ndimage\interpolation.py:583: UserWarning: From s
cipy 0.13.0, the output shape of zoom() is calculated with round() instead of int() - for these in
puts the size of the returned array has changed.
 "the returned array has changed.", UserWarning)
C:\Users\shubh\Anaconda3\lib\site-packages\numpy\core\fromnumeric.py:83: RuntimeWarning: overflow
encountered in reduce
 return ufunc.reduce(obj, axis, dtype, out, **passkwargs)

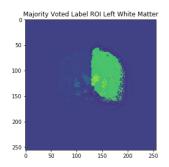


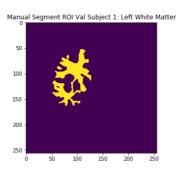


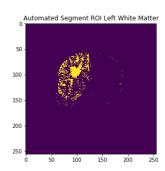


Jaccard Overlap: 0.26612605971249537

1 2



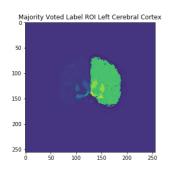


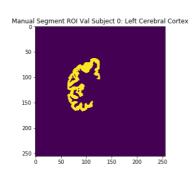


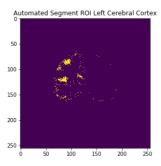
Jaccard Overlap: 0.23014668689934245

Computing Jaccard Overlap for Left Cerebral Cortex

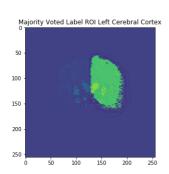
0 3

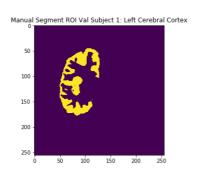


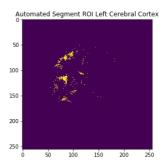




Jaccard Overlap: 0.11097708082026538

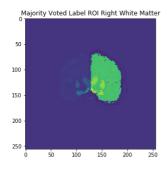


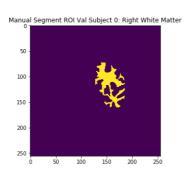


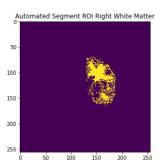


Jaccard Overlap: 0.07157569515962925
Computing Jaccard Overlap for Right White Matter

0 41

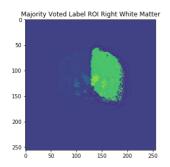


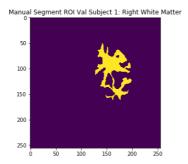


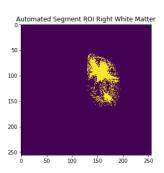


Jaccard Overlap: 0.4692113791861721

1 41



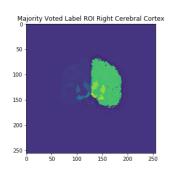


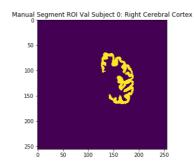


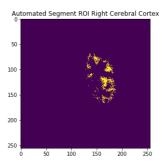
Jaccard Overlap: 0.42651673640167365

Computing Jaccard Overlap for Right Cerebral Cortex

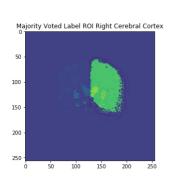
0 42

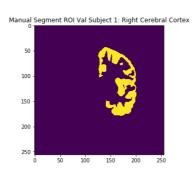


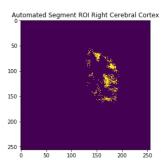




Jaccard Overlap: 0.239430724152365 1 42







Jaccard Overlap: 0.12496966755641835

In [79]:

import pandas as pd

table = pd.DataFrame(table,columns=['Validation Subject','ROI(label)','Jaccard Score'])
print (table)

Validation Subject ROI(label) Jaccard Score 0 Left White Matter 0.266126 0 Left White Matter 0.230147 1 1 Left Cerebral Cortex 2 0 0.110977 Left Cerebral Cortex 3 0.071576 1 Right White Matter Right White Matter 5 1 0.426517 0.239431 0 Right Cerebral Cortex 6 1 Right Cerebral Cortex 0.124970

In []: