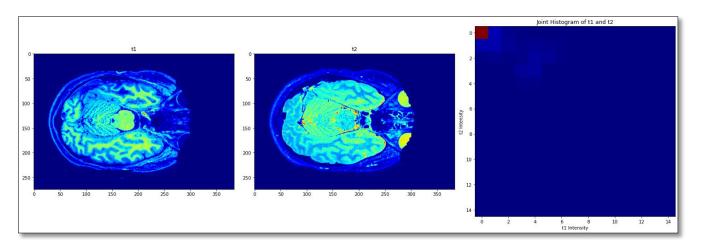
Part 1: Joint Histogram

a) Create python function JointHist(I, J, bin) calculating the joint histogram of two images of the same size. Pre-existing functions not to be used.

Solution:

In this task we have written a python function to create a joint histogram of two images of same size using histogramdd of numpy and plotting using matplotlib

Result:



b) For images of size n x p, verify that:

$$\textstyle\sum_{i,j} H_{I,J}(i,j) = n \cdot p$$

Solution:

Calculated sum of joint histogram created from two images t1 and t2 using np.sum(). Also calculated the product of each dimension of one of the image used. Both the result came out to be same.

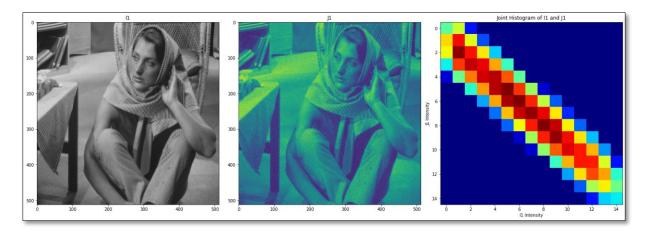
Note: In the below implementation the joint histogram was created out of two images of size (n,p,r). Hence the above formula is proved to be true for n*p*r as well.

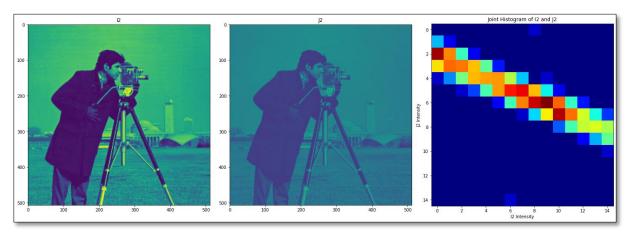
```
In [314]: runfile('C:/Users/scorp/OneDrive/Documents/Bishops Course/Sem2/
VolumetricImageAnalsis/Assignment2/Part1.py', wdir='C:/Users/scorp/OneDrive/
Documents/Bishops Course/Sem2/VolumetricImageAnalsis/Assignment2')
The shape of the t1 is (274, 384, 384)
The shape of the t2 is (274, 384, 384)
Sum of joint histogram of t1 and t2 is 40402944
Product of each dimension (274* 384* 384) of t1 is 40402944
Sum of joint histogram of two images is equal to product of each dimension of the size of an image used
In [315]:
```

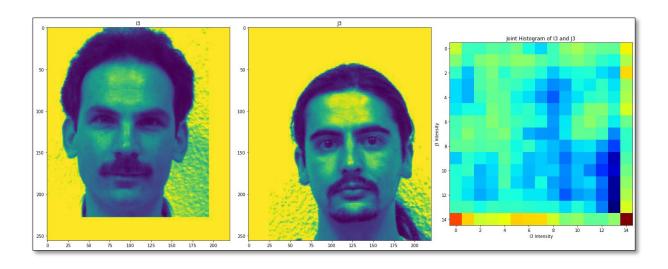
c) Calculate and show the joint histogram of different pairs of images given (I1,J1), (I2,J2)...etc. briefly describing the observation.

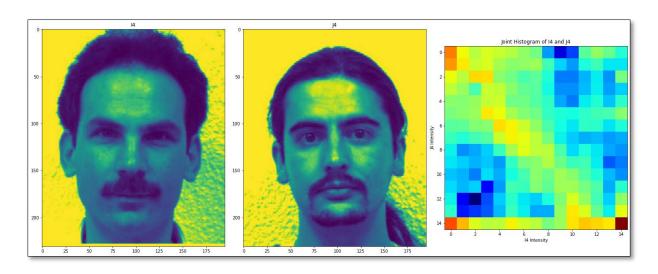
Solution:

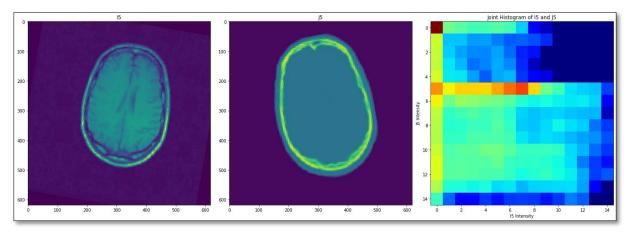
We have used the function created in part a to display the joint histogram of each pair of images I and J. We have used the logarithmic scale too to visualize the histogram more clearly.

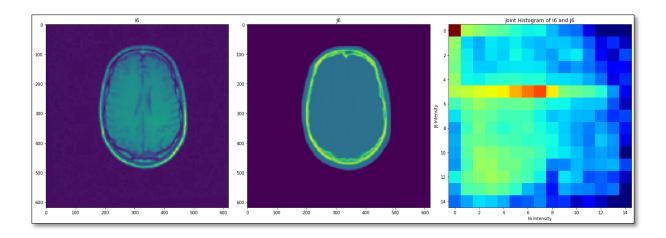












Part 2: Similarity Criteria

For two images I and J of the same size:

- a) Create python function SSD(I,J) calculating the sum squared difference
- b) Create python function CORR(I,j) calculating the pearson correlation coefficient
- c) Create python function MI(I,J) calculating the mutual information
- d) Compare results of the above three functions for different pair of images providing along with observations.

Observation:

```
In [260]: runfile('C:/Users/scorp/OneDrive/Documents/Bishops Course/Sem2/VolumetricImageAnalsis/Assignment2')

Calculating the sum squared difference between two images I and J:
Images II and JI are not of same shape

SSO of Image I2 and Image J2 is 30364219

SSO of Image I3 and Image J3 is 4401503

SSO of Image I3 and Image J4 is 3848301

SSO of Image I6 and Image J6 is 42510000

Calculating the pearson correlation coefficient between two images I and J:
Images II and JI are not of same shape

CORR of Image I2 and Image J2 is [[2.49969168 0.99621727]

[0.99621727 0.40005242]]

CORR of Image I3 and Image J3 is [[1.0212648 0.14339325]

[0.14339325 0.97921275]]

CORR of Image I3 and Image J4 is [[0.96004816 0.56404675]

[0.56404675 1.04166066]]

[0.5643844 1.18589841]]

CORR of Image I3 and Image J6 is [[0.86555368 0.78025159]]

[0.78025159 1.1553359]]

Calculating the mutual information between two images I and J:
Images II and JI are not of same shape

MI of Image I2 and Image J3 is 0.6687586245620836

MI of Image I3 and Image J3 is 0.06687586245620836

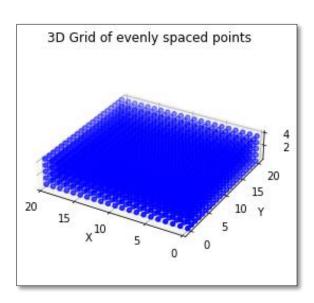
MI of Image I5 and Image J5 is 0.55488206712046

MI of Image I6 and Image J5 is 0.5748563215819839

In [270]:
```

Part 3: Spatial Transforms

a) Generate a 3d grid of evenly spaced points

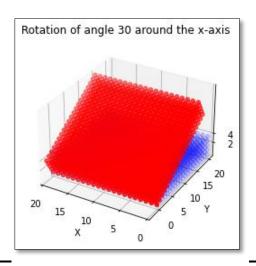


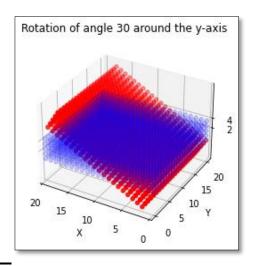
- b) Create a python function rigid_transform(theta, omega, phi, p, q, r) that returns the matrix (in homogenous coordinates) of the rigid transform corresponding to:
 - i. Rotation of angle theta around the x-axis
 - ii. Rotation of angle omega around the y-axis

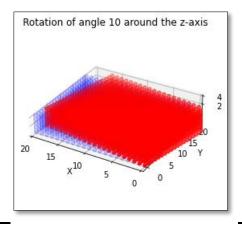
- iii. Rotation of angle phi around the z-axis
- iv. Translation of vector t=(p,q,r) test your function on the 3d point cloud from (a) and show the result.

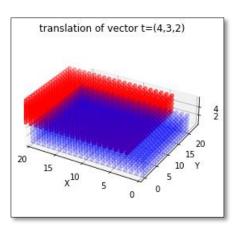
Result:

Note: Blue 3D point cloud is original and Red is the transformed one





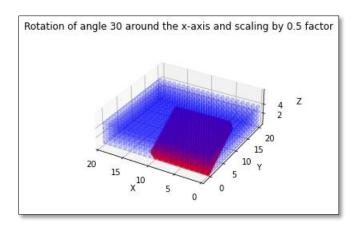


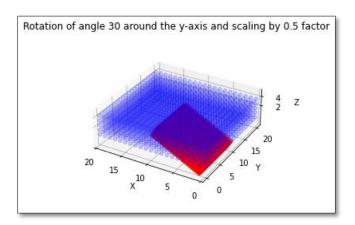


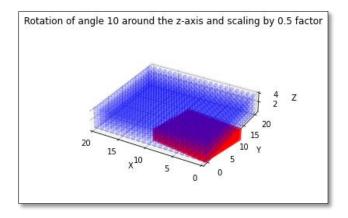
c) Write a function affine_transform(s, theta, omega, phi, p, q, r) that does the same as above (b) and adds a scaling factor s. test and show this function, as in (b)

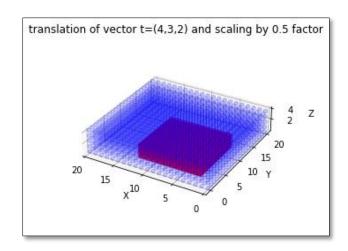
Result:

Note: Blue 3D point cloud is original and Red is the transformed one









d) Given the 3 following matrices M1, M2, M3, determine the type of transformation corresponding to each matrix. Justify.

Result:

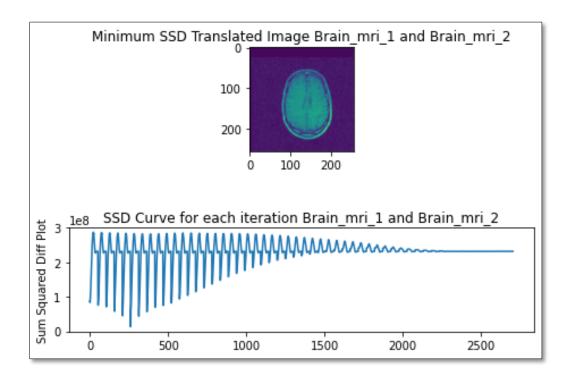
Part 4: Simple 2d registration

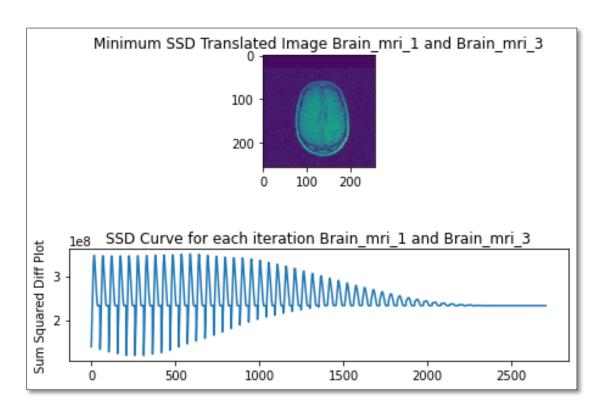
a) Translate an image by given vector t=(p,q) in x and y direction using pre-existing interpolation python function incase p and q values used are floats

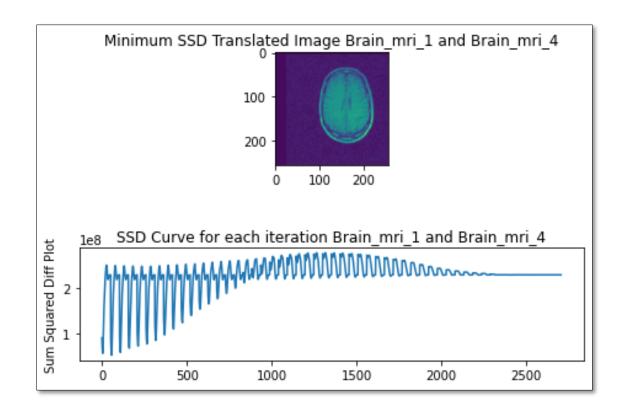
Result:



b) 2D registration minimizing SSD and considering only translations. The function written should search for the alignment, SSD for each iteration to be saved and plotted. The function to be tested on the 3 different translations provided for image Brain_MRI_1.png (BrainMRI_2,3,4).



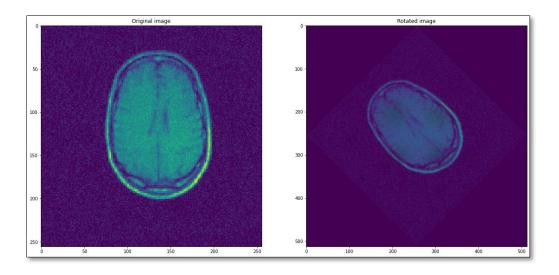




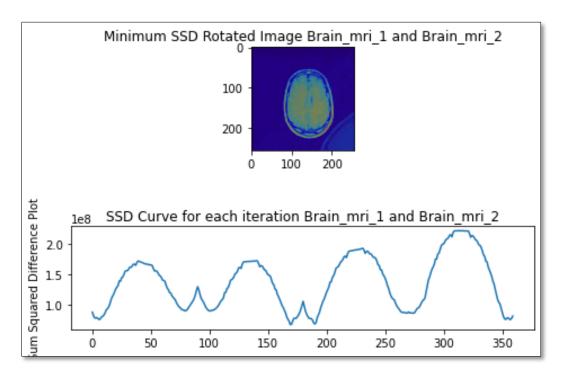
c) Rotate an image by an angle theta, around the top left of the image. Pre-existing python rotate functions is not to be used. Pre-existing interpolation python function can be used.

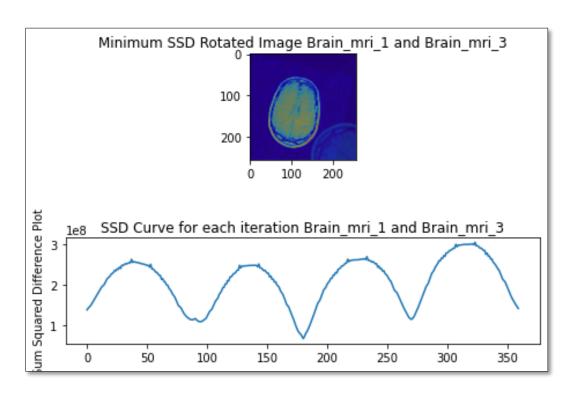
Result:

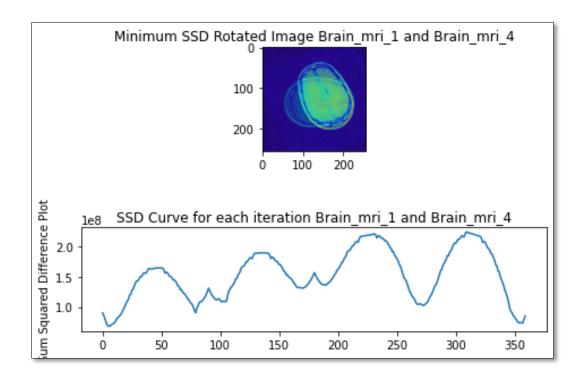
Rotate image by 45 degrees around the top left of the image. Used math and numpy libraries for implementation. Pre-existing interpolation python function is used.



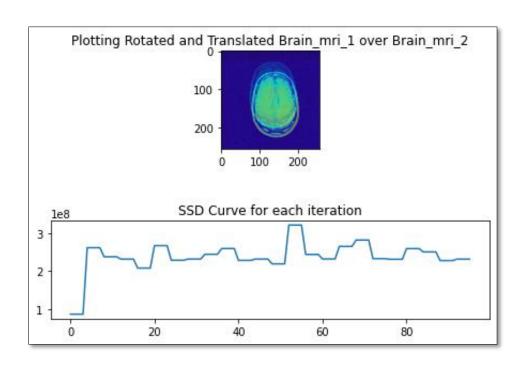
d) 2d registration of an image minimizing SSD and considering only rotations. For each iteration, SSD values are to be saved and plotted. The function to be tested on the 3 different translations provided for image Brain_MRI_1.png (BrainMRI_2,3,4).

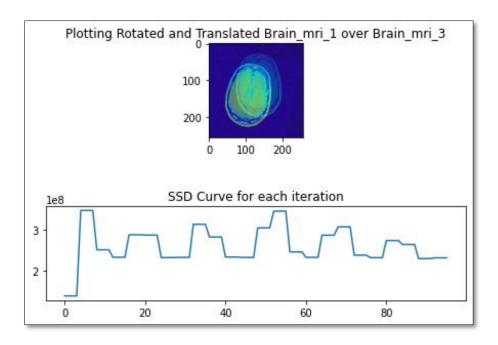


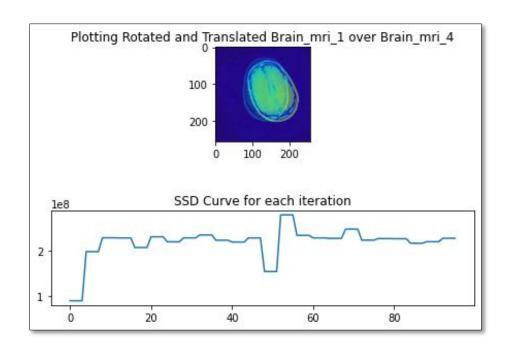




- e) Implement a gradient descent for minimizing SSD, considering both translation and rotation. Register BrainMRI_2,3,4 onto Brain_MRI_1. Which registrations converge? Which do not converge? Why do you think some fail to converge?
- i. To improve the performance of gradient descent, a more advanced optimization technique is needed. Improve your rigid registration with a better optimization technique (of your choice). Test for the 3 cases of rigid transformations given (BrainMRI $_2$ 3,4)



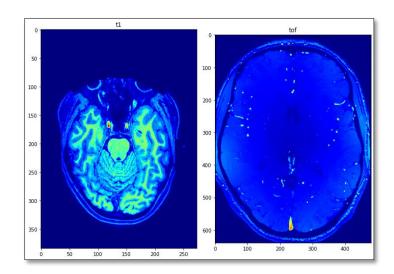




Part 4: Practical Application

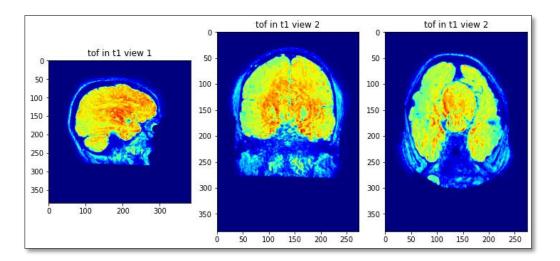
a) Align tof image on t1 image

- Plotting images t1 and tof



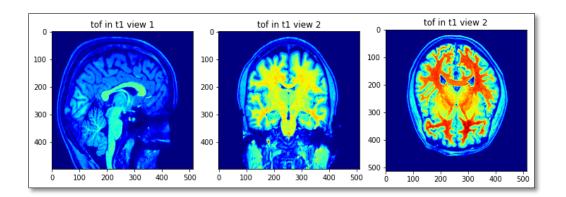
Command 1:

flirt -searchrx -180 180 -searchry -180 180 -searchrz -180 180 -in t2.nii -ref t1.nii -out t2_in_t1_aligned.nii.qz



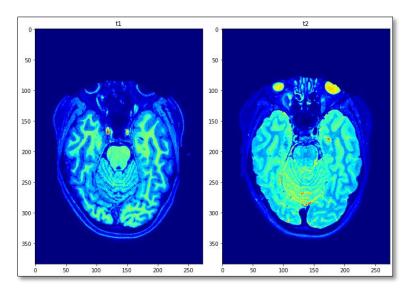
Command 2:

flirt -searchrx -180 180 -searchry -180 180 -searchrz -180 180 -cost mutualinfo -in t1.nii -ref tof.nii -out t2_in_t1_aligned_mi.nii.gz



b) Align t2 image on t1 image

- Plotting images t1 and t2



Command1:

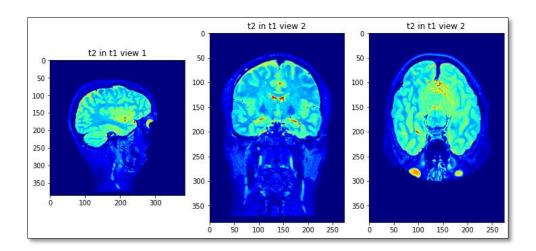
flirt -search
rx -180 180 -searchry -180 180 -searchrz -180 180 -in t
2.nii -ref t1.nii -out t2_in_t1_aligned.nii

Result:

```
shweta@LAPTOP-310F2A8A:/mnt/c/Users/scorp/Downloads/images$ flirt -searchrx -180 180 -searchry -180 180 -searchrz -180 180 -in t2.nii -ref t1.nii -o ut t2_in_t1_aligned.nii

Final result:
1.002575 0.002773 -0.007889 0.371326
0.000935 0.986044 0.005607 -0.545707
0.002766 0.021569 0.934856 4.455892
0.000000 0.000000 0.000000 0.000000 1.0000000

shweta@LAPTOP-310F2A8A:/mnt/c/Users/scorp/Downloads/images$
```



Command2:

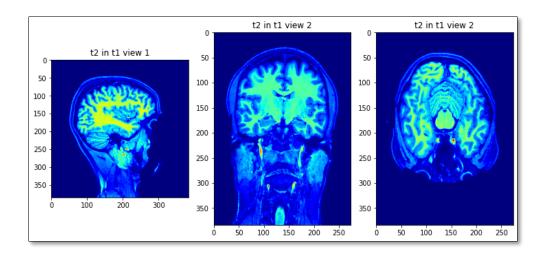
flirt -searchrx -180 180 -searchry -180 180 -searchrz -180 180 -cost mutualinfo -in t1.nii -ref t2.nii -out t2_in_t1_aligned_mi.nii

Result:

```
Shweta@LAPTOP-310F2A8A:/mnt/c/Users/scorp/Downloads/images$ flirt -searchrx -180 180 -searchry -180 180 -searchrz -180 180 -cost mutualinfo -in tl.n ii -ref t2.nii -out t2_in_t1_aligned_mi.nii

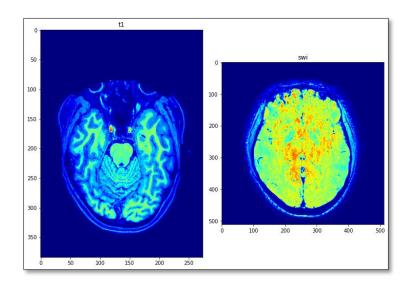
Final result:
1.001019 -0.003405 0.003635 -0.261618
0.003181 0.001203 -0.011500 1.151275
-0.004590 0.003408 1.000390 -0.926505
0.000000 0.0000000 0.0000000 1.0000000

shweta@LAPTOP-310F2A8A:/mnt/c/Users/scorp/Downloads/images$
```



c) Align swi image on t1 image

- Plotting images t1 and swi



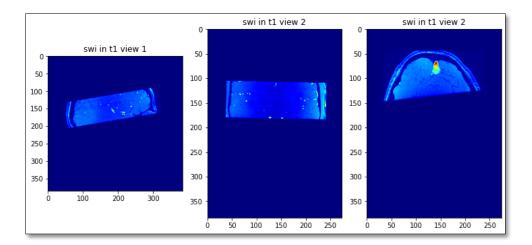
Command1:

flirt -searchrx -180 180 -searchry -180 180 -searchrz -180 180 -in swi.nii -ref t1.nii -out swi_in_t1_aligned.nii

Result:

```
Shweta@LAPTOP-310F2A8A:/mnt/c/Users/scorp/Downloads/images$ flirt -searchrx -180 180 -searchry -180 180 -searchrz -180 180 -in swi.nii -ref tl.nii -
out swi_in_tl_aligned.nii

Final result:
1.0000715 0.025382 0.025492 -17.587788
-0.026411 1.000367 0.029355 18.168634
-0.026411 1.000367 0.029355 18.168634
-0.026488 -0.024846 0.000000 0.0000000 1.0000000
```



Command2:

flirt -searchrx -180 180 -searchry -180 180 -searchrz -180 180 -cost mutualinfo -in t1.nii -ref swi.nii -out swi_in_t1_aligned_mi.nii

```
ShwetagLAPTOP-310F2A8A:/mmt/c/Users/scorp/Downloads/images$ flirt -searchrx -180 180 -searchry -180 180 -searchrz -180 180 -cost mutualinfo -in t1.n ii -ref swi.nii -out swi_in_t1_aligned_mi.nii

Final result:
1.001252 -0.021938 -0.026944 19.498028
0.021031 1.001724 -0.036457 -14.677683
0.026310 0.024546 1.001597 -75.630932
0.000000 0.000000 0.000000 1.0000000
shwetagLAPTOP-310F2A8A:/mmt/c/Users/scorp/Downloads/images$
```

