

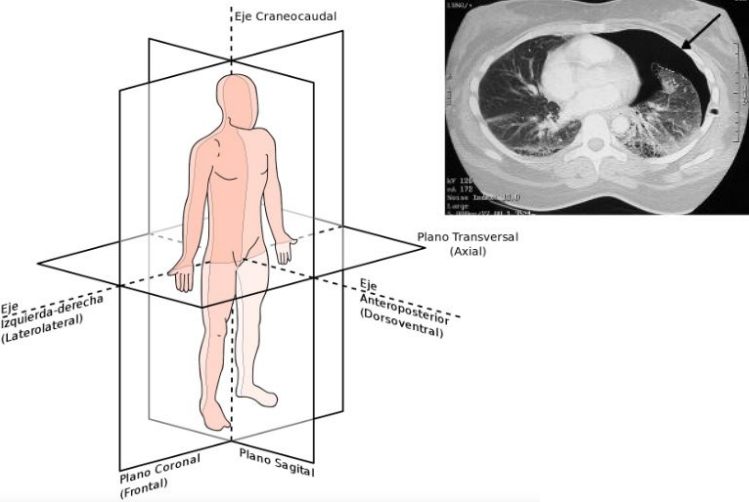
Final Lab

11763 - Procesamiento de Imágenes Médicas

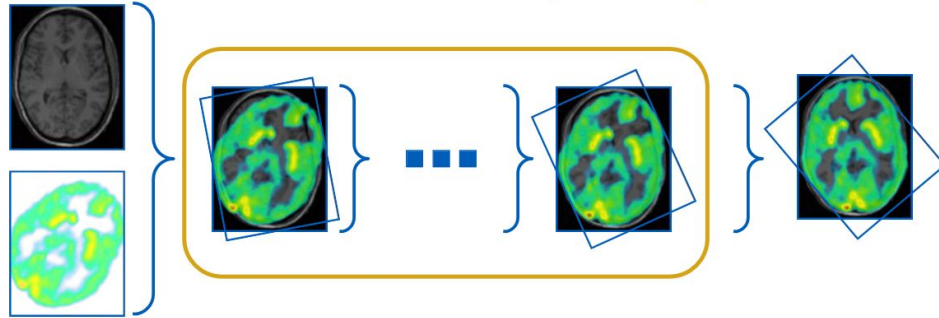
Martí Gelabert Gómez

What we have to do?

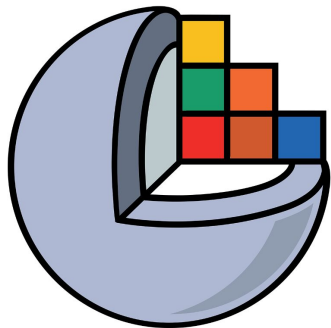
Task 1



Task 2



Dicom Files



3D Slicer

Pixel Array
Slice Index
Position Patient

```
Expression:
ds['PerFrameFunctionalGroupsSequence']

Result:
> parent = (ReferenceType) <weakref at 0x7f50b92c5ee0; to 'FileDataset' at 0x7f50b904c8b0>
  Protected Attributes
  > _abc_impl = (_abc_data) <_abc._abc_data object at 0x7f50ba56b480>
    _is_protocol = (bool) False
  > _list = (list: 356) [(0008, 9124) Derivation Image Sequence 1 item(s) ---- \n (0008, 2112) Source Image Sequence 1 item(s) ---- \n
    > 000 = (Dataset: 4) (0008, 9124) Derivation Image Sequence 1 item(s) ---- \n (0008, 2112) Source Image Sequence 1 item(s) ---- \n
      > DerivationImageSequence = {Sequence: 1} <Sequence, length 1>
      > FrameContentSequence = {Sequence: 1} <Sequence, length 1>
      > PlanePositionSequence = {Sequence: 1} <Sequence, length 1>
      > SegmentIdentificationSequence = {Sequence: 1} <Sequence, length 1>
        default_element_format = {str} '%(tag)s %(name)-35.35s %(VR)s: %(repval)s'
        default_sequence_element_format = {str} '%(tag)s %(name)-35.35s %(VR)s: %(repval)s'
        file_tell = (int) 14080
        indent_chars = {str} ' '
        is_decompressed = (bool) False
        is_implicit_VR = (NoneType) None
        is_little_endian = (NoneType) None
        is_original_encoding = (bool) False
```

Task 1

Task 1: Procedure

1. Get information from DICOM files.
2. Order them with slice information.
3. Process Segmentation masks
4. Apply transformations to all.
5. Generate GIF

Task 1: Get the data

```
slices = []
for i in paths:
    ds = pydicom.dcmread(i)
    if hasattr(ds, 'SliceLocation'):
        slices.append(ds)
slices = sorted(slices, key=lambda s: s.SliceLocation)

# stack the slices to obtain a 89xIMG_SIZExIMG_SIZE
slides3d = np.array([s.pixel_array for s in slices])
```

```
def sortMaks(mask):
    lista_ordenada = sorted(mask, key=lambda tupla: tupla[1])
    lista_A = [tupla[0] for tupla in lista_ordenada]
    return lista_A
```

```
ds = pydicom.dcmread('HCC_005/01-23-1999-NA-ABDPVLVIS-36548/300.000000-Segmentation-06660/1-1.dcm')
segmentation = []
mask1 = []
mask2 = []
mask3 = []
mask4 = []

for index, i in enumerate(ds['PerFrameFunctionalGroupsSequence']):
    image_position = i.PlanePositionSequence[0].ImagePositionPatient # última componente

    segment_seq = i.SegmentIdentificationSequence
    if segment_seq is not None:
        segment_number = segment_seq[0].ReferencedSegmentNumber # esto te dice en que segmentación estas

        if segment_number == 1:
            mask1.append((ds.pixel_array[index], image_position[2]))

        if segment_number == 2:
            mask2.append((ds.pixel_array[index], image_position[2]))

        if segment_number == 3:
            mask3.append((ds.pixel_array[index], image_position[2]))

        if segment_number == 4:
            mask4.append((ds.pixel_array[index], image_position[2]))
```

Task 1: Apply Transformations

```
5 usages  🧑 Martí Gelabert Gómez
def MIP_sagittal_plane(img_dcm: np.ndarray) -> np.ndarray:
    """ Compute the maximum intensity projection on the sagittal orientation. """
    return np.max(img_dcm, axis=2)

4 usages  🧑 Martí Gelabert Gómez
def MIP_coronal_plane(img_dcm: np.ndarray) -> np.ndarray:
    """ Compute the maximum intensity projection on the coronal orientation. """
    return np.max(img_dcm, axis=1)

5 usages  🧑 Martí Gelabert Gómez
def rotate_on_axial_plane(img_dcm: np.ndarray, angle_in_degrees: float) -> np.ndarray:
    """ Rotate the image on the axial plane. """
    return scipy.ndimage.rotate(img_dcm, angle_in_degrees, axes=(1, 2), reshape=False)
```

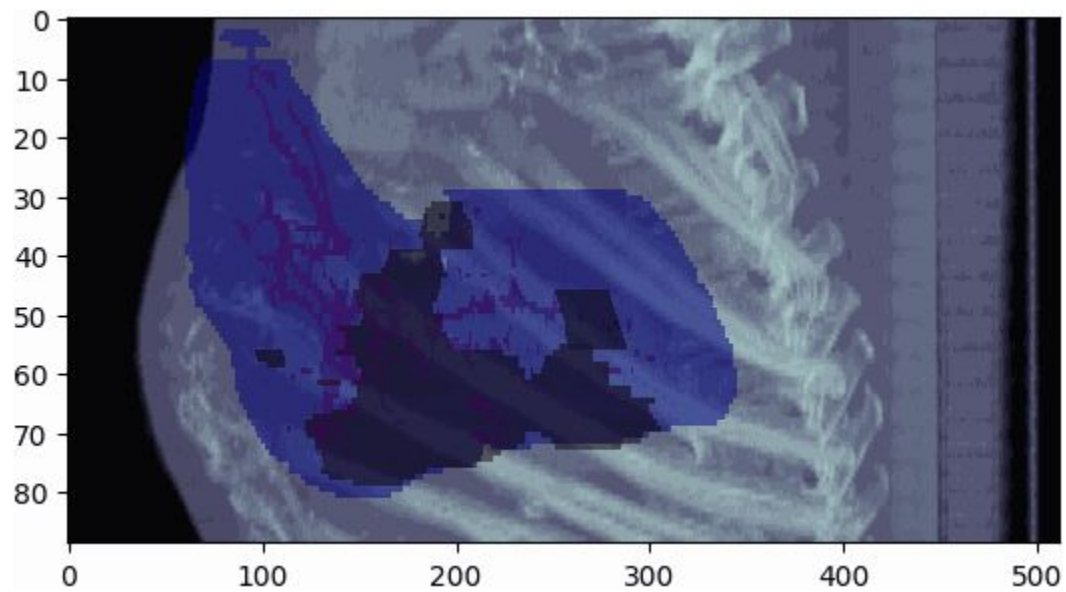
```
n = 24
projections = []
for idx, alpha in enumerate(np.linspace(0, 360 * (n - 1) / n, num=n)):
    rotated_img = rotate_on_axial_plane(img_dcm, alpha)
    projection = MIP_sagittal_plane(rotated_img)

    rotated_mask1 = rotate_on_axial_plane(mask1, alpha)
    rotated_mask2 = rotate_on_axial_plane(mask2, alpha)
    rotated_mask3 = rotate_on_axial_plane(mask3, alpha)
    rotated_mask4 = rotate_on_axial_plane(mask4, alpha)
```

```
mask1_projection = MIP_sagittal_plane(rotated_mask1)
mask2_projection = MIP_sagittal_plane(rotated_mask2)
mask3_projection = MIP_sagittal_plane(rotated_mask3)
mask4_projection = MIP_sagittal_plane(rotated_mask4)
```

Task 1: Show your projections

```
plt.imshow(projection, cmap=cm, vmin=img_min, vmax=img_max, aspect=pixel_len_mm[0] / pixel_len_mm[1])
plt.imshow(mask1_projection, cmap="jet", aspect=pixel_len_mm[0] / pixel_len_mm[1], alpha=0.5)
plt.imshow(mask2_projection, cmap="hot", aspect=pixel_len_mm[0] / pixel_len_mm[1], alpha=0.5)
plt.imshow(mask3_projection, cmap="viridis", aspect=pixel_len_mm[0] / pixel_len_mm[1], alpha=0.5)
plt.imshow(mask4_projection, cmap="inferno", aspect=pixel_len_mm[0] / pixel_len_mm[1], alpha=0.5)
```

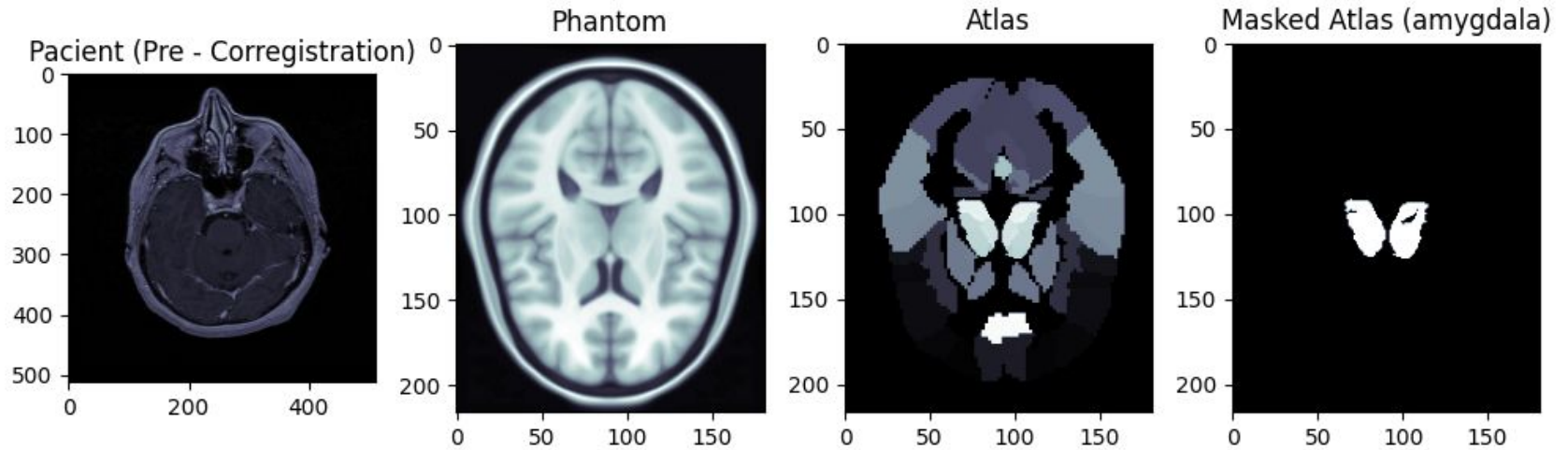



Task 2

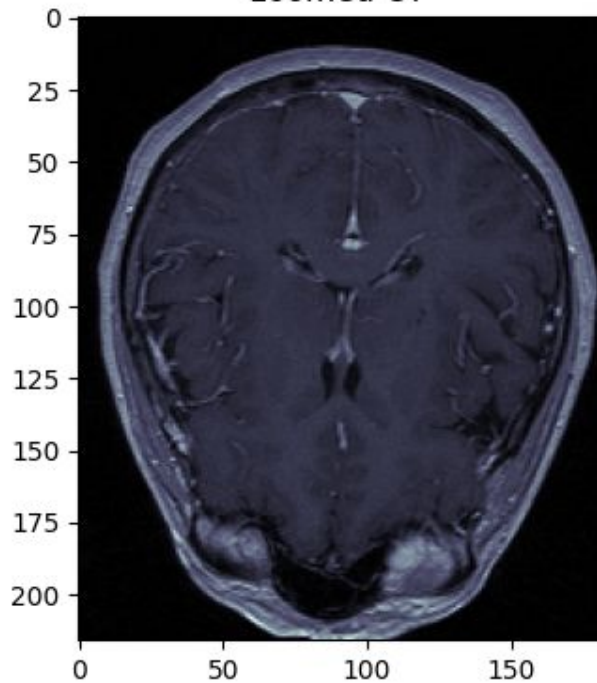
Task 2: Procedure

1. Get information from DICOM files.
2. Order them with slice information.
3. Fix CT proportions to match reference.
4. Find landmarks.
5. Use an optimization algorithm to find optimal parameters for transformation
6. Transform and apply with your optimal parameters.

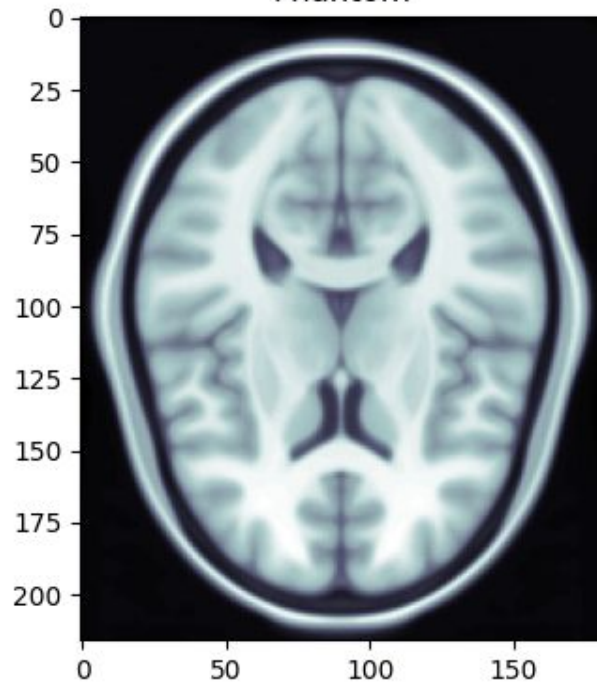
Task 2: Fix proportions



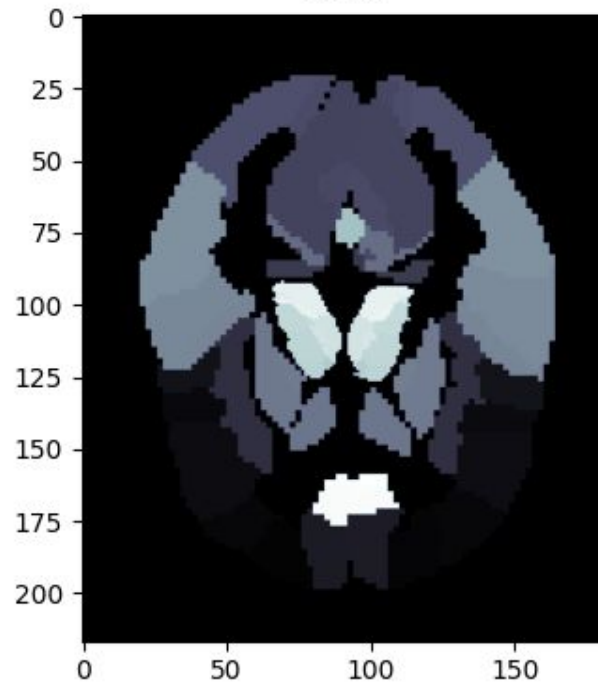
zoomed CT



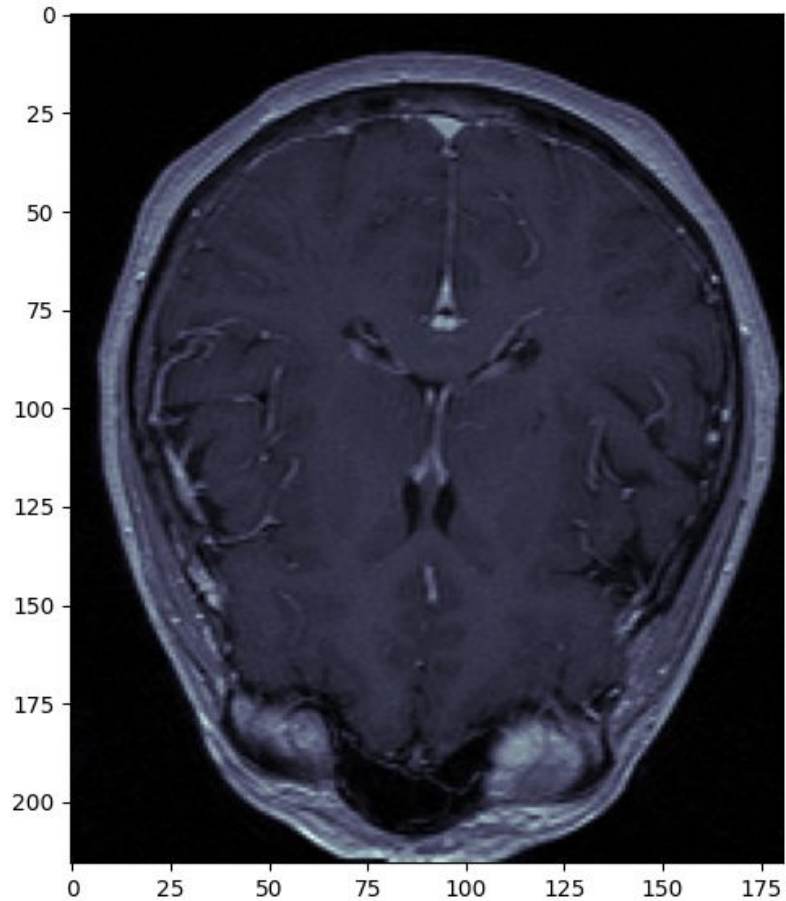
Phantom



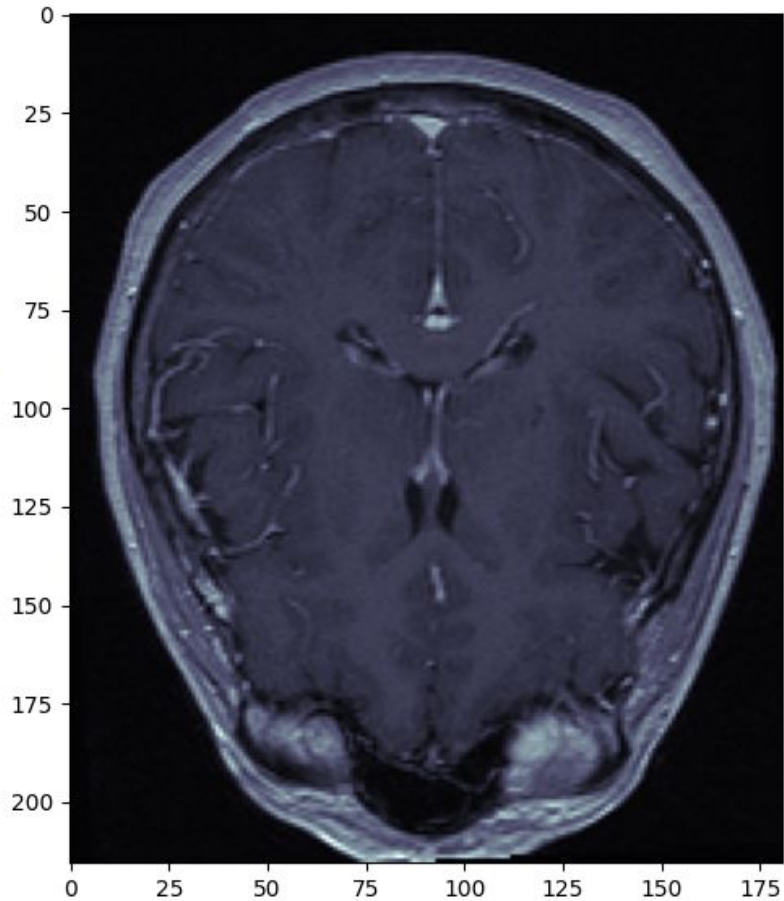
Atlas



zoomed CT

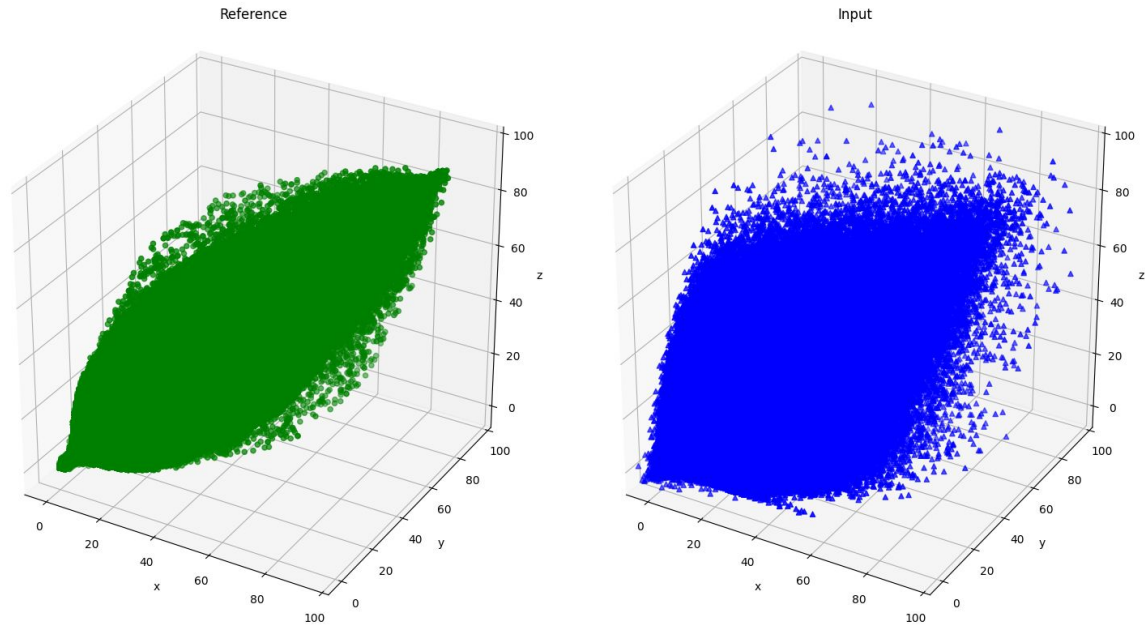


Rotated Zoomed CT



```
# Computing The landmarks (https://gist.github.com/ttschreiner/8f971bbbd40606e58f1e4fb1852e8b8e)
landmarks_ref = images_phantom[:, :, :15].reshape(-1, 3)
landmarks_input = processedCT[:, :, :15].reshape(-1, 3)
```

Landmark Before coregistration




```
def coregister_landmarks(ref_landmarks: np.ndarray, inp_landmarks: np.ndarray):
    """ Coregister two sets of landmarks using a rigid transformation. """
    initial_parameters = [
        0, 0, 0, # Translation vector
        0, # Angle in rads
        1, 0, 0, # Axis of rotation
    ]

    # Find better initial parameters
    centroid_ref = np.mean(ref_landmarks, axis=0)
    centroid_inp = np.mean(inp_landmarks, axis=0)
    # Your code here:
    # ...
    initial_parameters[0] = centroid_ref[0] - centroid_inp[0]
    initial_parameters[1] = centroid_ref[1] - centroid_inp[1]
    initial_parameters[2] = centroid_ref[2] - centroid_inp[2]
```

👤 Martí Gelabert Gómez

```
def function_to_minimize(parameters):
    """ Transform input landmarks, then compare with reference landmarks. """
    # Your code here:
    # ...
    inp_landmarks_transf = np.asarray(
        [translation_then_axialrotation(point, parameters) for point in inp_landmarks])
    return vector_of_residuals(ref_landmarks, inp_landmarks_transf)

# Apply least squares optimization
result = least_squares(
    function_to_minimize,
    x0=initial_parameters,
    verbose=1)
return result
```

>> Mean residual value: 24.97651824358491. (before corregistration)

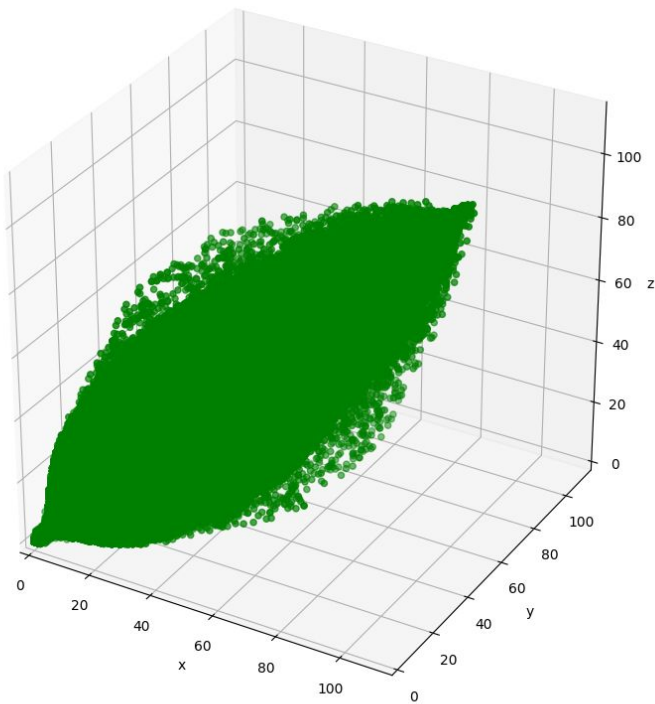
>> Best parameters: ([7.76592284 8.37130049 10.05798676 -0.25040872
0.59583415 0.44332766 0.71613931]

mean_absolute_error (centroid_idx)>> 17.34841054272961

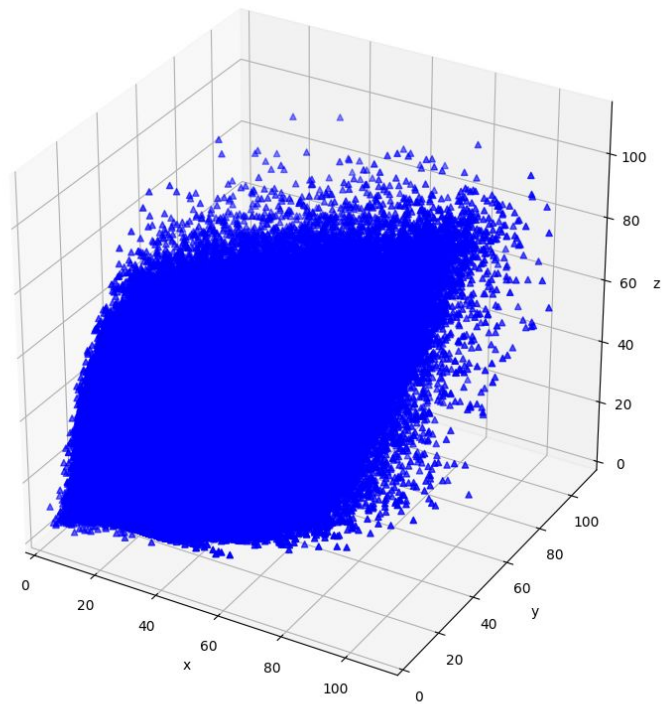
mean_squared_error (centroid_idx)>> 433.00956087370577

mutual_information (centroid_idx)>> 0.6591496641902275

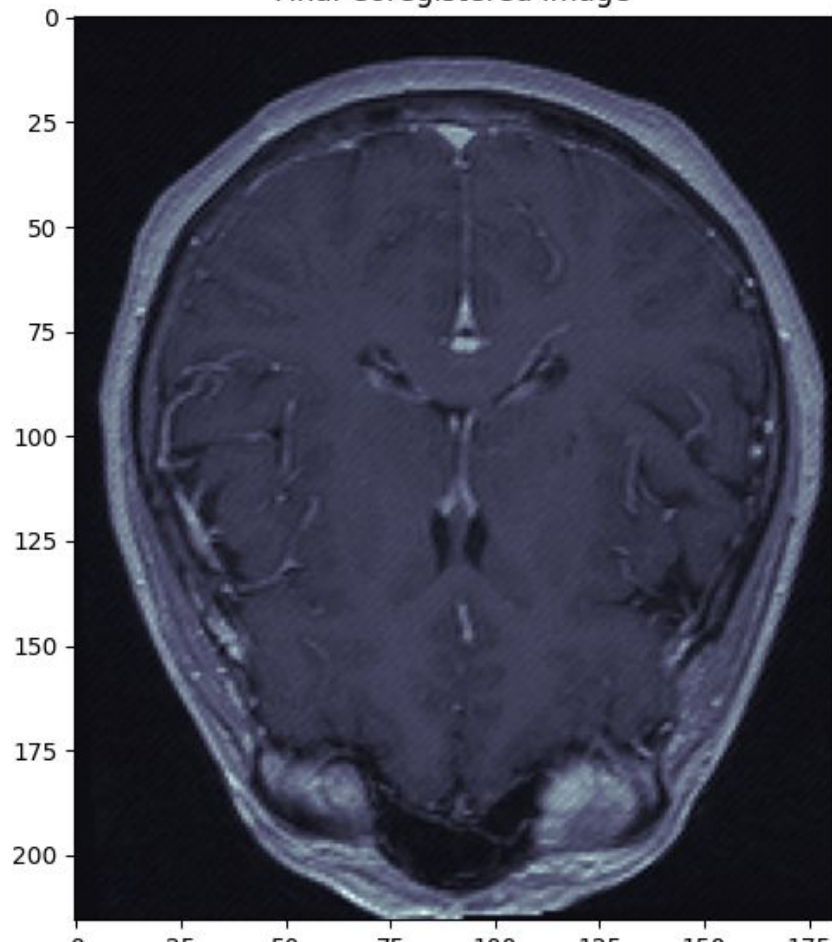
Reference



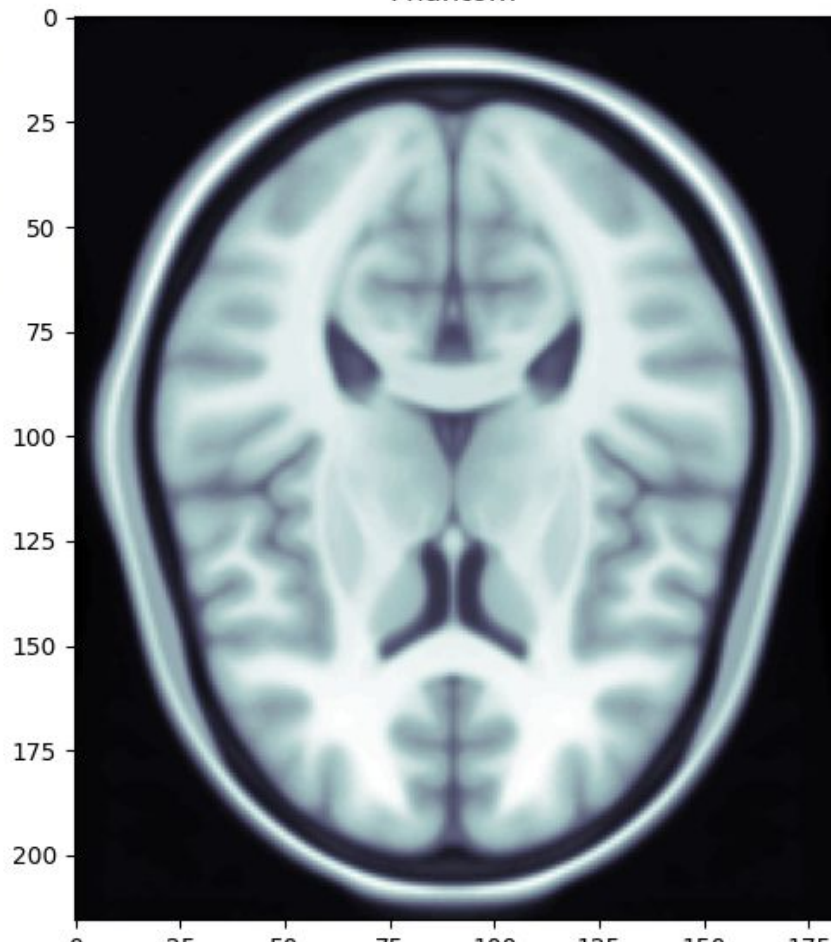
Transformed Input

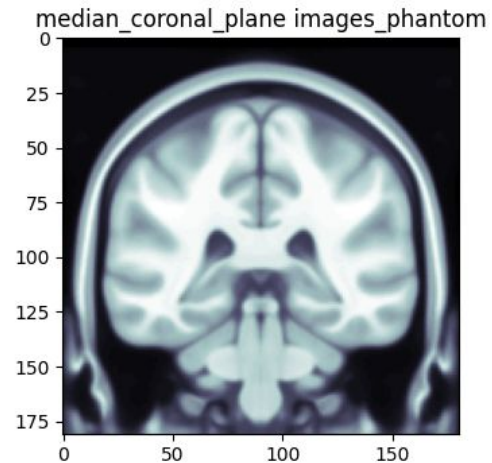
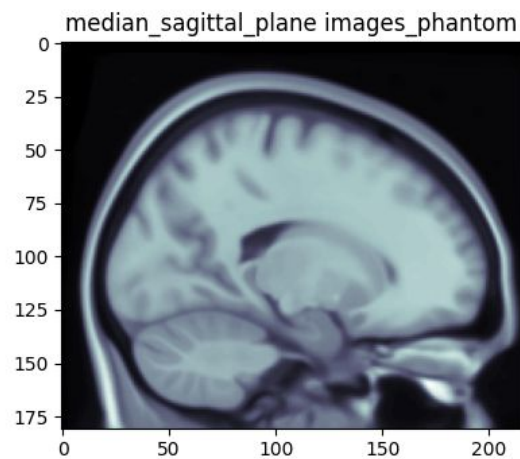
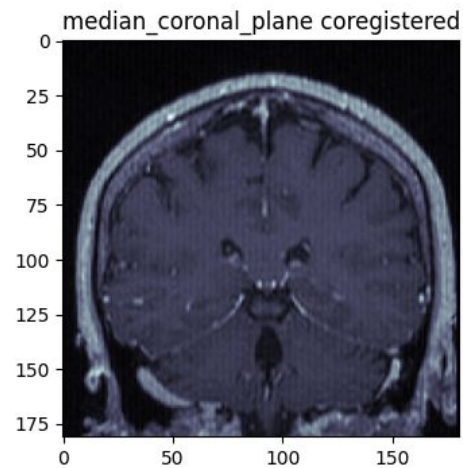
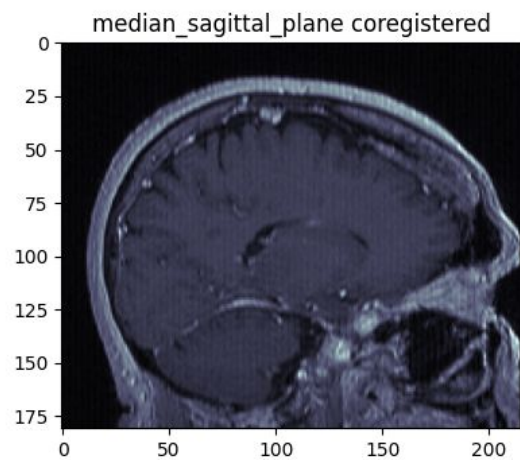


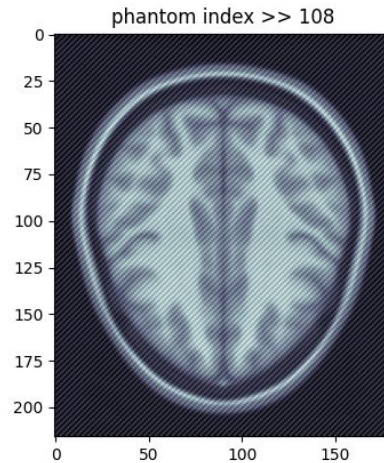
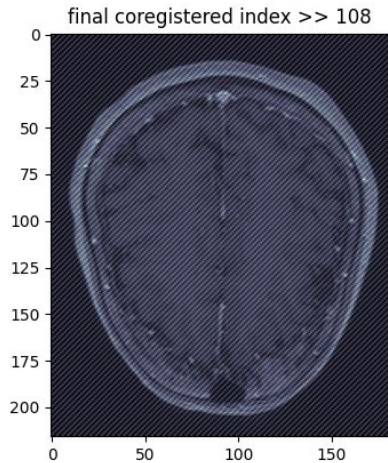
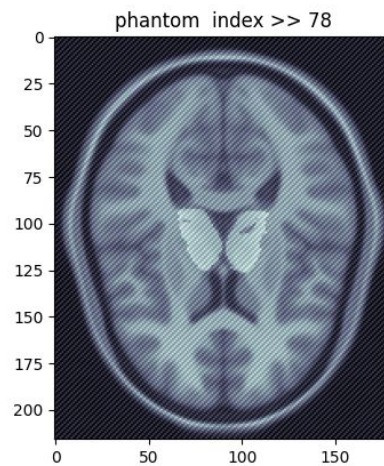
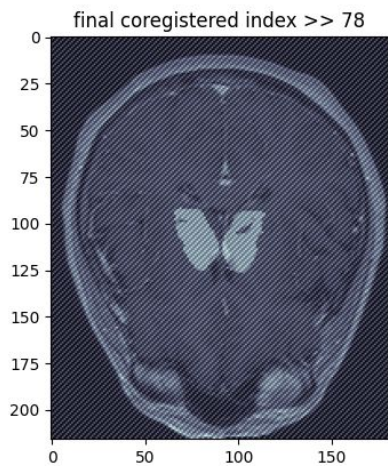
Final Coregistered image



Phantom







Q&A

Thank you for your time!