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Homework 3: Brain Cancer Image Segmentation Pipeline on Seven Bridges

Question 1) Write a short summary about the (a) why segmentation is important, (b) various segmentation tools you learnt about, and (c) possible challenges of such workflows (6 points)

- a. Segmentation is the process of dividing digital images into subgroups called image segments. Labels are then assigned to pixels to identify important objects or elements of a given image. Image segmentation is important because it allows investigators to identify different regions of an image, reduces the complexity of images for easy analysis, and is a significant step for computer vision technologies and algorithms.
- b. One segmentation tool used for skull stripping or brain extraction in order to get a clear image of the brain while also preserving patient confidentiality (proper de-identification) is the Brain Mask Generator (BrainMaGe). This tool uses deep learning segmentation architecture to strip brain images of bone, marrow, and blood vessels that are not needed for analysis. This is challenging as some non-brain material may be hard to remove computationally and result in affected segmentation outputs. Another tool used during the brain segmentation process is called SUSAN or Smallest Univalve Segment Assimilating Nucleurs. SUSAN uses nonlinear filtering to reduce the noise of images while preserving the structure and quality of the image contents. This tool is helpful for smoothing all volumes and reduces the possibility of high frequency intensity variations in an image. Additionally, Seven Bridges uses the tool 'FAST' with FSL brain segmentation that relies on the intensity of in the radio frequency to identify subregions of the brain which can be useful for highlighting cerebrospinal fluid (CSF) and other portions of the brain.
- c. The main challenge of segmentation workflows is maintaining the integrity and quality of a given image while removing and refining certain aspects for proper analysis. If there is non-brain elements left over from the brain extraction step or too much/too little noise left over from the denoising step, segmentation outputs suffer and may not be as useful to gain insight from. Imaging technology can introduce a lot of noise and when handling the data, there is risk for noise to be added and error introduced. With Statistical Parametric Mapping (SPM), which produces similar outputs to FSL, the brain extraction and skull-stripping step is not used. Instead, SPM uses probability maps to understand and identify subregions of the brain such as grey matter, white matter, CSF, etc. However, predicting the boundaries of each voxel is challenging.

Question 2) Write a short summary in simple words what the FSL brain segmentation workflow in Seven Bridges does (1 point)

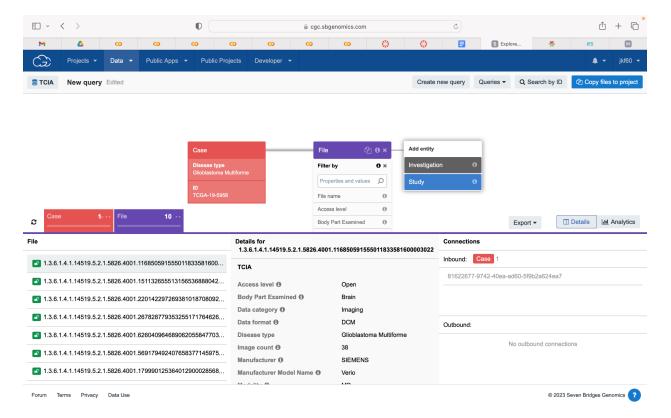
a. The FSL brain segmentation workflow works through five steps of image segmentation processing. First, images are re-oriented using LPS (left posterior superior) coordinate system (a requirement for GLISTRboost) so that they can be properly read. Then, images are registered using FSL FLIRT and skull stripped using FSL BET to de-noise and clear out all non-brain elements (image normalization). Finally images are segmented by FSL FAST and de-noised using SUSAN as discussed above. Files (HTML) are outputted with grey matter, white matter, CSF, and other tissue (FLAIR images only).

Question 3) Can you think of any real world use cases/applications that would use such an image segmentation workflow ? (2 point)

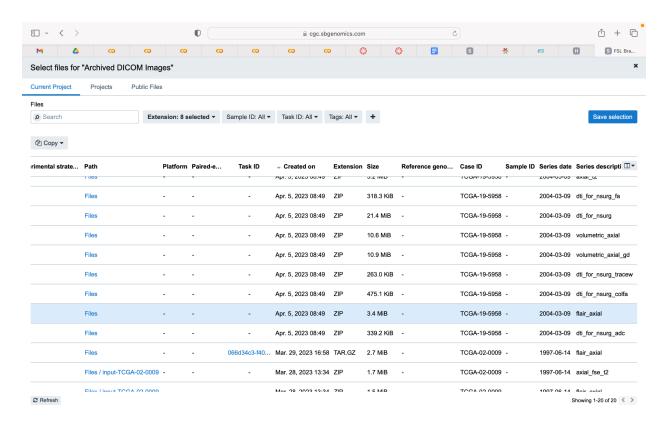
a. Some real-world applications for brain image segmentation include measuring and visualizing the brain's anatomical features for research purposes or education, analyzing brain changes, delineating pathological regions of the brain, and surgical preparation and image-guided procedures. Additionally, as we discussed in class, tumor size and location is easy to analyze and predict using image segmentation which can help clinical teams in providing personalized patient care and decision making.

Question 4) (6 points) Run the FSL Brain segmentation workflow on a TCGA-GBM patient # TCGA-19-5958.

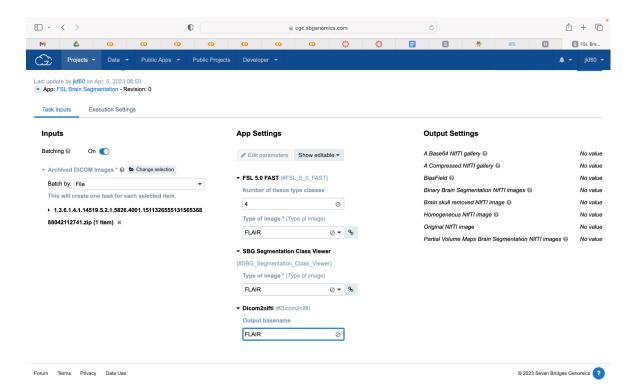
1. Selecting image files for patient #TCGA-19-5958 in Data Browser. Copy to project.



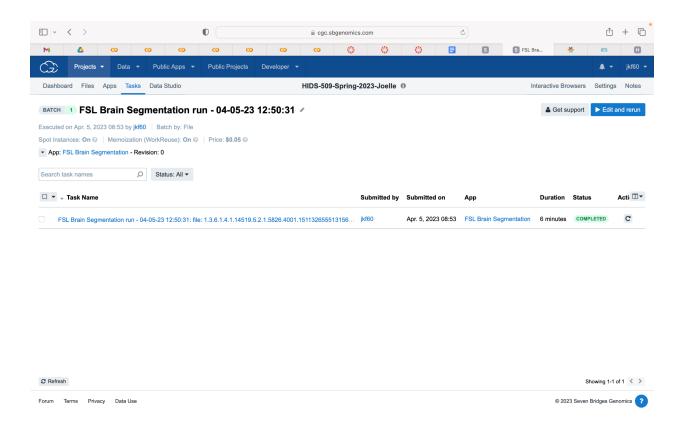
2. Selecting FLAIR image files for the patient to be used in FSL workflow.



3. Number of tissue type classes = 4. Image Type = FLAIR.



4. Workflow complete. This workflow took 6 minutes and cost \$0.05.



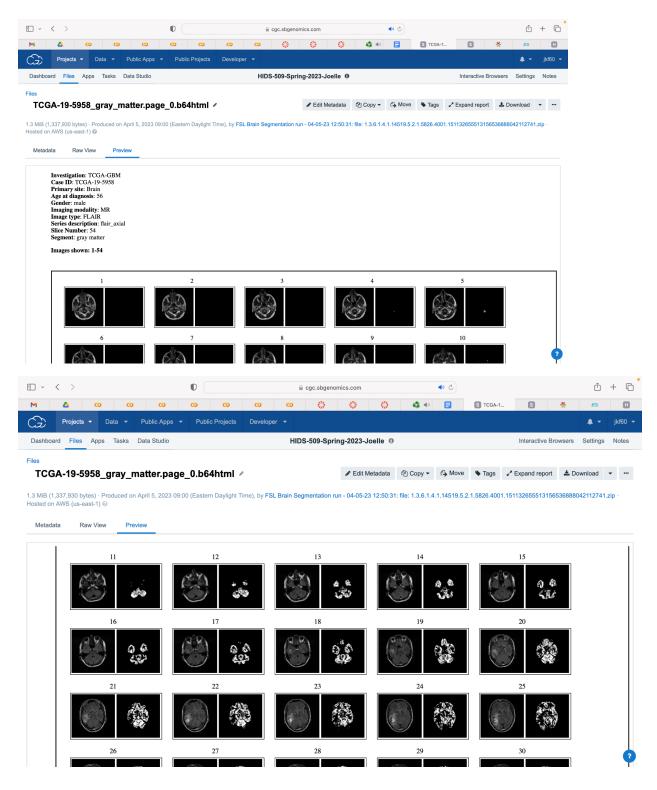
Output Settings 🝃

- ▼ A Base64 NIfTI gallery ② ►
 - TCGA-19-5958_csf.page_0.b64html
 - TCGA-19-5958_white_matter.page_0.b64html
 - TCGA-19-5958_gray_matter.page_0.b64html
 - TCGA-19-5958_other_tissue_0.page_0.b64html

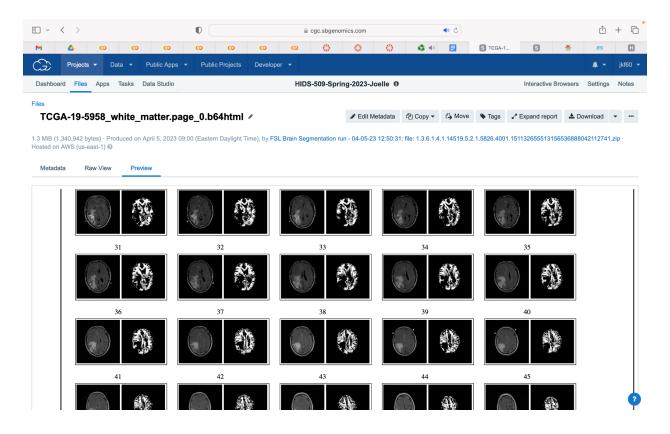
The files outputted from running the FSL workflow with FLAIR images include cerebrospinal fluid, white matter, grey matter, and other tissue.

Selected File Outputs:

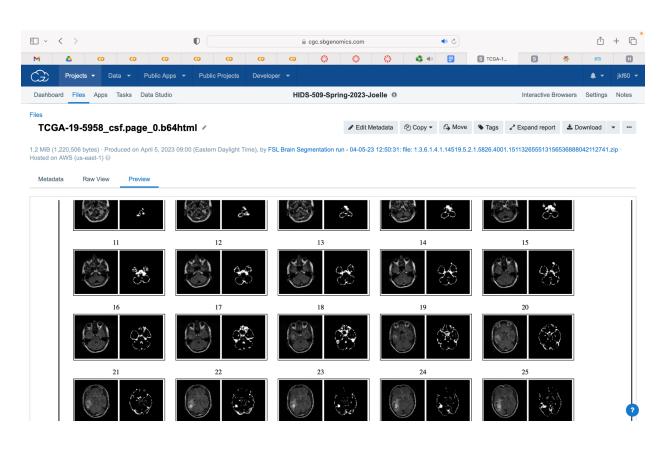
Grey Matter:



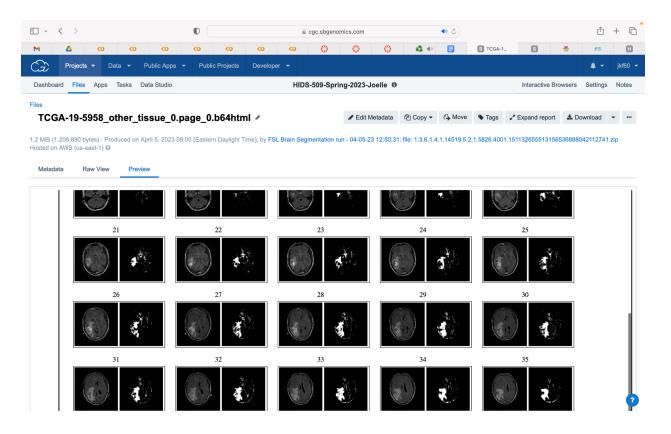
White Matter:



CSF:



Other Tissue:



From these screenshots above, you can see that the workflow was able to identify regions of the brain (grey matter, white matter, and CSF) as well as other tissue which could be a tumor. Other tissue elements are only produced with FLAIR images. In the second screenshot on the above page, the FSL workflow identified the region of tumor growth in this patient (big white blob in image).

Question 5) This TCGA Gliobastoma dataset we used in class today contains not only imaging data but also genomics data. For example, there is gene expression data from the same patients. This is represented as a matrix with features (genes) as rows and samples as columns. The numerical data are normalized gene expression values. As a data scientist, how would you integrate/connect the two data types (imaging and gene expression data)? (Bonus 1 point)

As a data scientist, I would be interested in understanding which genes are differentially expressed in certain regions of the brain and then studying patient images where abnormalities are produced and linking which genes may be responsible (over or under expressed) for that abnormality. Connecting gene expression data with imaging data may be useful for

understanding why a tumor is growing in a certain region of the brain and highlighting patterns in cancer growth for certain patients.