

# Project Plan: Brain Tumor Segmentation Pipeline

## 1. Project Overview & Goal

Our mission is to develop a robust Computer Vision pipeline that can automatically segment (outline) a tumor from a brain MRI scan. This is a high-impact project that demonstrates a deep understanding of multi-stage image processing, which is far more impressive than a simple classification task.

Our final goal is to create an application that:

1. Takes a brain MRI as input.
2. Processes it through a 6-module pipeline.
3. Outputs the original image with the predicted tumor region clearly highlighted.
4. Provides a quantitative score of our model's accuracy.

## 2. The Dataset: Figshare MRI Dataset

We will be using the publicly available dataset we found on Figshare.

- **Contents:** 3,064 T1-weighted MRI images from 233 patients.
- **Key Feature:** Crucially, it includes **ground truth tumor masks**. A mask is a perfect, pixel-by-pixel outline of the tumor, which is essential for training and, more importantly, for evaluating the accuracy of our final result.
- **Format:** The data is in .mat files, which we will read using the scipy library in Python.

## 3. The 6-Module Pipeline: Our Blueprint for Success

This project is perfectly suited for our six-person team structure. Each team member will own one sequential module. **The success of the entire project depends on each module owner strictly adhering to the defined Input and Output contracts.**

| Module & Owner   | Objective                                 | Input from Previous Module  | Output to Next Module   |
|------------------|---|-----------------------------|---|
| 1. Data Handler  | Load and parse the complex .mat files.    | File path to a .mat file.   | 1. Original MRI (NumPy array)2. Ground Truth Mask (NumPy array) |
| 2. Pre-processor | Clean the raw MRI to improve segmentation | Original MRI (NumPy array). | Cleaned MRI (NumPy array).                                      |

|                                   |   |   |  |
|-----------------------------------|---|---|--|
|                                   | accuracy.   |   |  |
| <b>3. Segmenter (Core Logic)</b>  | Identify and isolate the pixels that belong to the tumor. | Cleaned MRI (NumPy array).                            | Predicted Tumor Mask (Binary NumPy array).               |
| <b>4. Post-processor</b>          | Refine the predicted mask to remove noise and errors.     | Predicted Tumor Mask (NumPy array).                   | Final Predicted Mask (Cleaned NumPy array).              |
| <b>5. Analyst &amp; Evaluator</b> | Quantify the tumor's properties and measure our accuracy. | 1. Final Predicted Mask 2. Ground Truth Mask          | 1. Tumor Area & Centroid 2. <b>Dice Score (Accuracy)</b> |
| <b>6. UI / Visualizer</b>         | Display the results in a user-friendly interface.         | 1. Original MRI 2. Final Predicted Mask 3. Dice Score | A visual application for demonstration.                  |

## 4. Measuring Success: The Dice Coefficient

How do we know if our model is good? We can't just say "it looks okay." We need a number.

Our primary metric for success will be the **Dice Similarity Coefficient (DSC)**.

- **What it is:** A score from 0.0 to 1.0 that measures the overlap between our predicted tumor mask and the perfect ground truth mask.
- **Our Goal:** To maximize this score. A score of **0.85 or higher** would be an excellent result for a mini-project.

The Analyst (Module 5) will be responsible for implementing the function to calculate this score.

## 5. Key Challenges & Our Strategy

1. **Reading .mat files:** This is not a standard image format.
  - **Strategy:** The Data Handler (Module 1) will use the `scipy.io.loadmat` function. This is the first technical task to solve.
2. **Getting good segmentation results:** This is the hardest part of the project.
  - **Strategy:** The Segmenter (Module 3) will start with a simple, classical algorithm like **Otsu's Thresholding**. We will test it, see the results, and then try a more complex method like **K-Means Clustering** to improve our Dice Score. This iterative approach

is key.

## 6. Tech Stack

- **Language:** Python
- **Core Libraries:**
  - OpenCV (for all image processing tasks)
  - NumPy (for handling image arrays)
  - SciPy (specifically for loading .mat files)
  - Streamlit (for building the UI in Module 6)

## 7. Immediate Next Steps

1. **Assign Owners:** Assign one team member to each of the 6 modules.
2. **Set Up Environment:** Everyone should install the required Python libraries.
3. **Module 1 Starts:** The owner of Module 1 begins work immediately on the script to load and display an MRI and its mask from a .mat file. This is our first milestone.