

# Research Papers & References for Brain Tumor Segmentation

This document provides key academic papers to guide the implementation of the 6-module brain tumor segmentation pipeline. The focus is on papers that explain the classical, non-deep-learning methods relevant to the project's scope, as well as significant modern approaches for context.

## Category 1: Foundational Papers (Core Implementation)

These papers cover the classical pipeline steps that form the basis of this project's implementation.

### 1. "Brain Tumor Segmentation Using Thresholding, Morphological Operations and Extraction of Features from MRI Images"

- **Relevance to Project:** This paper serves as a practical blueprint. It details the use of thresholding (e.g., Otsu's method) and morphological operations for cleaning up segmented masks. This directly informs the implementation for **Module 3 (Segmenter)** and **Module 4 (Post-processor)**.
- **Key Takeaway:** The methodology section demonstrates how fundamental image processing operations, when properly sequenced, can achieve effective segmentation.
- **Source Link:** <https://ieeexplore.ieee.org/stamp/stamp.jsp?tp=&arnumber=7002427>

### 2. "An efficient brain MR image segmentation and tumor detection using K-Means and Fuzzy C-Means"

- **Relevance to Project:** This paper explains the use of clustering algorithms for segmentation, which is a more advanced alternative to simple thresholding. It is a primary resource for the **Module 3 (Segmenter)** owner to implement K-Means clustering.
- **Key Takeaway:** The paper clarifies how K-Means can partition an image into distinct regions based on pixel intensity, thereby isolating the tumor from surrounding tissues.
- **Source Link:** <https://ieeexplore.ieee.org/stamp/stamp.jsp?tp=&arnumber=8864326>

## Category 2: Advanced Model (Context & Future Work)

This paper describes the modern, deep-learning-based state-of-the-art, which is essential context for the project report.

### 3. "U-Net: Convolutional Networks for Biomedical Image Segmentation" by Ronneberger et al.

- **Significance:** This is a seminal paper in the field of medical image analysis, introducing the U-Net architecture. U-Net is the current industry standard for biomedical segmentation. While not implemented in this project, it must be cited in the final report as a state-of-the-art alternative and for the "Future Work" section.

- **Key Takeaway:** The model's encoder-decoder structure is a key concept, enabling it to both understand the image context and achieve precise localization of the target region.
- **Source Link:** <https://arxiv.org/pdf/1505.04597>

### Category 3: Survey & Review Papers (Project Framing)

This paper provides a high-level overview of the field, which is valuable for the project's introduction and literature review sections.

#### 4. "A comprehensive review on brain tumor segmentation and classification using deep learning"

- **Relevance to Project:** This review summarizes a wide array of techniques, both classical and modern. It is an excellent resource for the **Project Lead** and those responsible for writing the report's introduction. It helps to properly position the project's classical approach within the broader research landscape.
- **Key Takeaway:** The paper is useful for understanding the historical progression of segmentation techniques and for sourcing relevant terminology.
- **Source Link:** <https://ieeexplore.ieee.org/stamp/stamp.jsp?tp=&arnumber=10940343>

### Recommended Reading Strategy:

1. **All Team Members:** Should read the abstracts of all four papers to understand the project's context.
2. **Module 3 & 4 Owners:** Should perform a detailed study of papers **#1 and #2**, as their implementation will be directly based on the described methodologies.
3. **Project Lead / Report Writers:** Should use paper **#4** to frame the introduction and literature review, and paper **#3** to discuss state-of-the-art methods and future work.