

Tuning into Tumors: a 2D Fourier Frequency Filter Approach

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Abstract

Brain tumor segmentation from MRI is a critical step in diagnosis, treatment planning, and disease monitoring, yet it remains challenging due to tumor heterogeneity, low-contrast boundaries, and noise in medical images. Current manual and simple intensity-based automated methods are time-consuming and often struggle to capture complex tumor textures and irregular edges. Motivated by the need for more robust, signal-informed segmentation tools, we developed an image-processing pipeline that treats each MRI slice as a two-dimensional signal and analyzes it in both the spatial and frequency domains. Using high-pass and band-pass Fast Fourier Transform (FFT) filters alongside spatial-domain Canny edge detection, we aim to enhance tumor visibility and improve boundary detection and regional isolation of tumors. We benchmark these approaches against Otsu-based intensity thresholding baselines and evaluate segmentation performance using Dice coefficient and Intersection over Union (IoU) metrics.

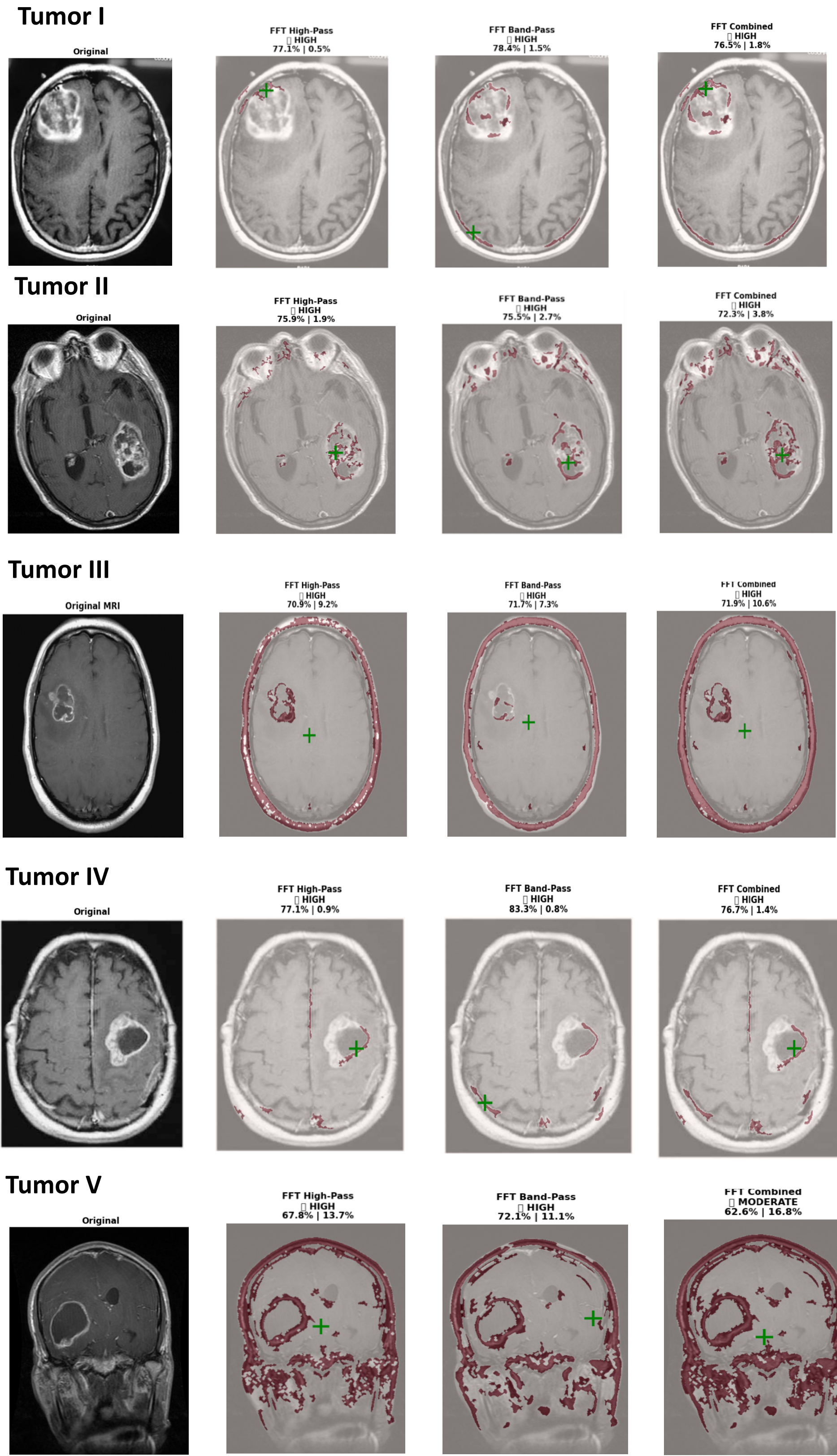
Objectives

- Develop an automated signal-processing-based tumor segmentation pipeline for MRI/CT images
- Implement and evaluate frequency-domain filtering approaches using 2D Fast Fourier Transform (FFT) with high-pass and band-pass masks
- Implement spatial-domain segmentation methods, including Canny edge detection and Otsu thresholding with morphological post-processing
- Establish and compare against intensity-based baseline methods (raw and Gaussian-smoothed Otsu thresholding)
- Quantitatively evaluate segmentation performance using: dice similarity coefficient (DSC); intersection over Union (IoU / Jaccard Index); and boundary-based accuracy metrics
- Statistically compare segmentation performance across methods to assess the impact of frequency-domain vs. spatial-domain processing

Methods

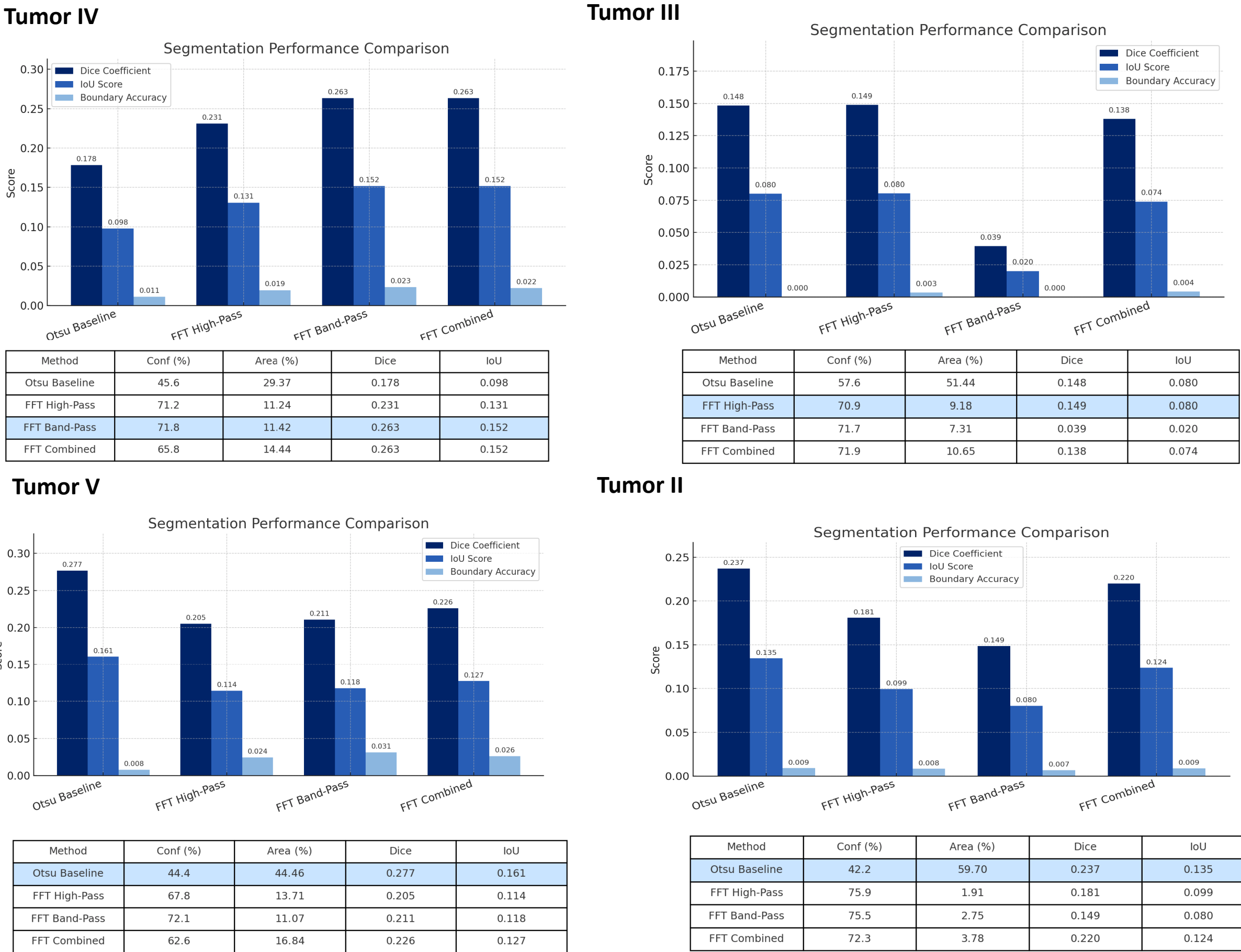
- Image Acquisition & Preprocessing:** MRI slices were loaded into Python, converted to grayscale, and normalized to [0, 1] intensity; ground truth masks were then loaded
- Frequency-Domain Transformation:** compute a 2D Fast Fourier Transform (FFT)
- Filter Design & Application:** *High-pass filters*: to suppress low-frequency, and enhance edge features; *Band-pass filters*: to isolate mid-frequency texture
- Segmentation Pipeline:** FFT & Otsu thresholding, Canny edge detection, and baseline raw/smoothed Otsu. Morphological cleanup & large connected component selection were used to generate final tumor masks
- Quantitative Evaluation:** predicted segmentation masks were compared to ground truth using DSC, IoU, & boundary accuracy metrics to assess overlap accuracy

Results



Left panel shows the original axial MRI slice. Middle panels show frequency-filtered reconstructions highlighting tumor-associated edges and texture patterns. Right panels display the corresponding binary segmentation masks produced by the algorithm, where highlighted regions with green crosses represent the predicted tumor area.

Quantitative Evaluation



Future Directions

- Enhance segmentation agreement with ground truth:**
 - Optimize frequency-domain filter parameters and thresholding strategies to better align predicted masks with expert-annotated tumor boundaries
 - Investigate post-processing techniques to reduce false positives/negatives while improving Dice and IoU scores
- Implement tumor volume estimation:**
 - Incorporate pixel dimensions and slice thicknesses to enable physically meaningful measurements & extend current 2D segmentation outputs to multi-slice 3D volumetric calculations to quantify total tumor burden
- Expand signal processing features:**
 - Implement multi-scale frequency filtering to better capture tumor texture and boundary heterogeneity
 - Integrate spatial-frequency hybrid features to improve robustness across varying image contrast and noise levels
- Scale & benchmark the pipeline:**
 - Evaluate performance across larger and more diverse imaging datasets
 - Standardize segmentation accuracy metrics to support reproducibility and cross-method comparison
- Integrate skull-stripping to suppress non-brain signal**
 - Implement a skull-stripping step to remove bone and extracranial tissue, reducing high-intensity artifacts and improving the reliability of downstream segmentation

References

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- AI Assistance: ChatGPT and Claude were used for writing and coding support as permitted*



GitHub Repository