

Image Processing Brain Tumors

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- **What is the problem you want to solve?**
 - To make an image-processing brain tumor predictive model to automate on scale.
 - The accuracy has to be above 95% to be considered successful. This will be completed within the next 3 months.
 - Stakeholders would be hospitals and doctors.
 - The solution space will most likely be creating the model using Python, Pandas, and CNN.
 - Constraints would be the limit amount of image data to train and the quality of the image data.
- **Who is your client and why do they care about this problem? In other words, what will your client do or decide based on your analysis?**
 - My client would be hospitals wanting to
 - scan through their existing history of MRI brain scans, but the treatment
 - Or quickly scan through a new patient's brain MRI scan.
 - Yes a human can do it but if you have 10,000 images, then you would want to automate this tedious process and save manpower, time, and money.
- **What data are you using? How will you acquire the data?**
 - I will be using this Dataset from Kaggle.
 - <https://www.kaggle.com/datasets/masoudnickparvar/brain-tumor-mri-data-set>
 - which was provided by the UCI Machine Learning Repository.
 - <https://ieee-dataport.org/documents/brain-tumor-mri-dataset>
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- **Briefly outline how you'll solve this problem. Your approach may change later, but this is a good first step to get you thinking about a method and solution.**
 - **Data Wrangling:**
 - Preprocess images (resize, normalize, augment) and split into training, validation, and testing sets.
 - **EDA:**
 - Count the number of images, classes, and check image dimensions, file sizes, class imbalance.
 - **Feature Engineering:**
 - Then standardize the data to make it compatible to be inputted into the predictive model.
 - Will split the data set into a training dataset (80% of the total) and a testing dataset (20% of the total). I will design and train the model til I get the desired accuracy.
 - **Modeling:**
 - Use CNNs for classification (e.g., ResNet), U-Net for segmentation, or YOLO/Faster R-CNN for detection.
 - Train with appropriate loss functions, optimizers, and learning rate schedules; monitor metrics to avoid overfitting.
 - Assess performance using metrics like accuracy, F1-score, Dice coefficient, IoU, or mAP based on the task.
 - Fine-tune hyperparameters, use transfer learning, and apply regularization techniques for improved results.
 - **Documentation:**
 - Present results with confusion matrices, ROC curves, segmented overlays, or detection bounding boxes.
 - Summarize methodology, results, and insights with suggestions for future improvements.
- **What are your deliverables?**
 - I will deliver a GitHub link to my code, slide deck, and project report.