Radiogenomic Analysis of Glioblastoma with Deep Learning Techniques

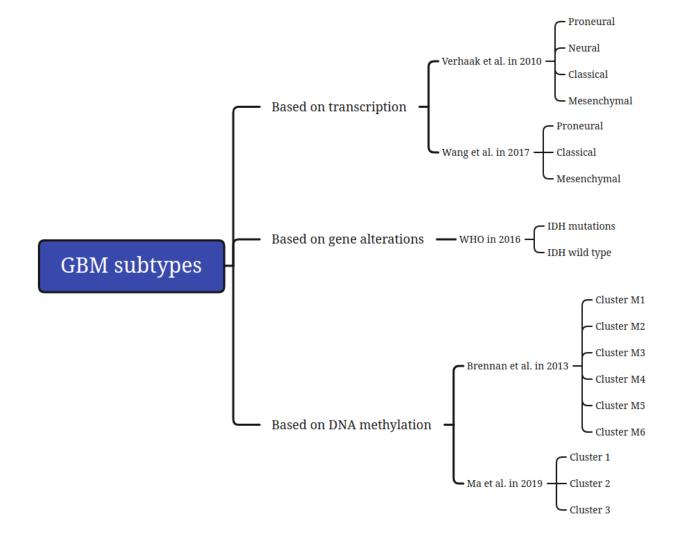
01/17/2021

Yichong Wang

Brief Introduction

Glioblastoma

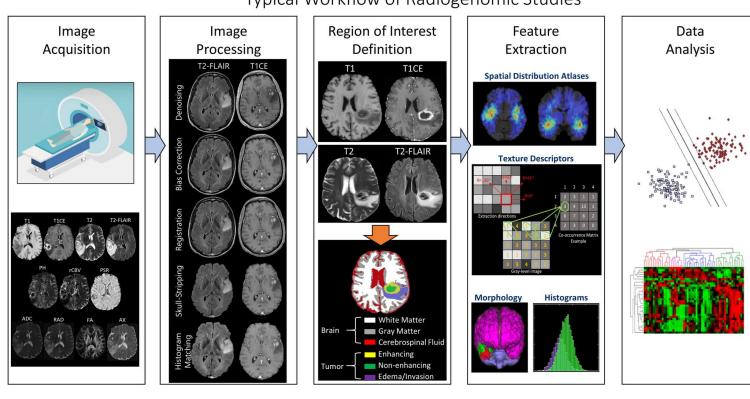
- Glioblastoma (GBM) is a highly aggressive adult brain tumor and the deadliest brain tumor that originates in glial cells.
- A high degree of heterogeneity in both space and time
 - Do not respond well to prescribed therapy because therapy targets the entire tumor rather than specific genetic sub-regions.
- Three main criteria to divide GBM subtypes
 - The most common subtype we use in lab is Varhaak et al. research in 2010



Radiogenomics

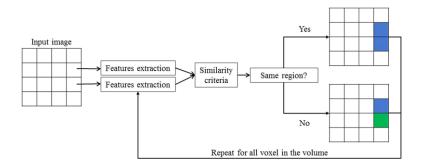
- Image acquisition;
- Image processing, including noise/artifact reduction, intensity and/or orientation standardization, coregistration of the multiparametric MRI scans
- **ROI definition** using manual annotation or automatic segmentation
- Feature extraction based on humanengineered (conventional radiomics) or deeplearning approaches
- Data analysis, involving machine/deeplearning methods for feature selection, classification, and cross-validation

Typical Workflow of Radiogenomic Studies

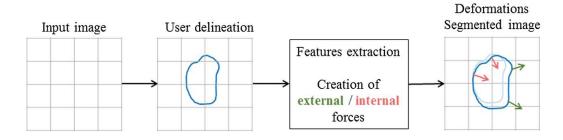


Progress made to-date

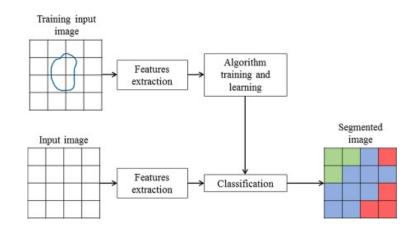
Segmentation Result

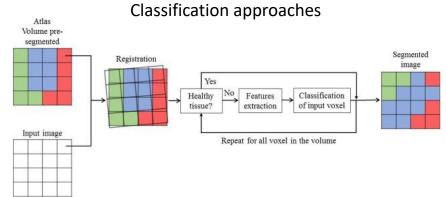


Region-based Approaches



Edge-Based Approaches

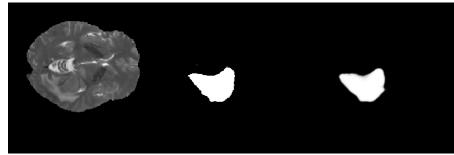




Altas-based approcahes

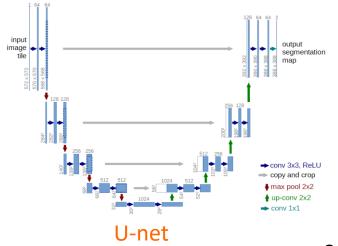
Segmentation Result

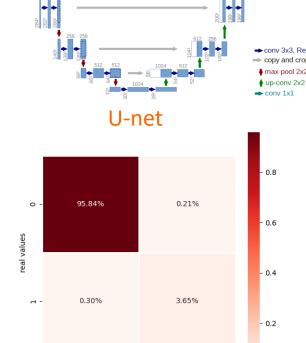






Left: MRI Image, Middle: Ground Truth, Right: Semgentation Results





predictions

Dataset: 39 Patients

MRI Series: T1,T2,FLAIR

Plane: AX, COR, SAG

Totally 18528 Images

Train: Test = 4:1

Train times: 225248

Sensitivity and Specificity:





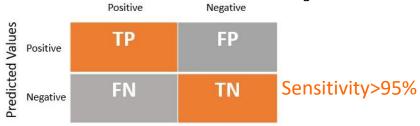


Figure 1

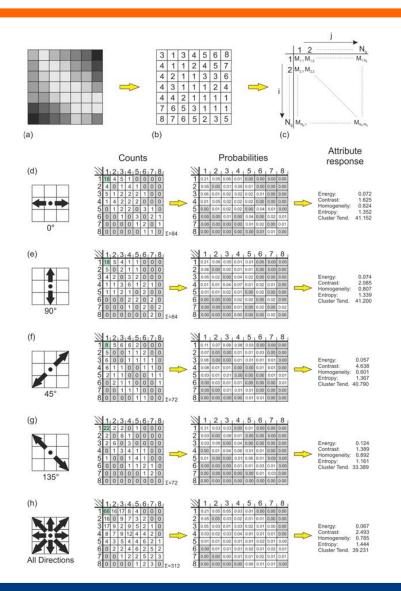
True Positives $Sensitivity = \frac{}{True\ Positives + False\ Negatives}$

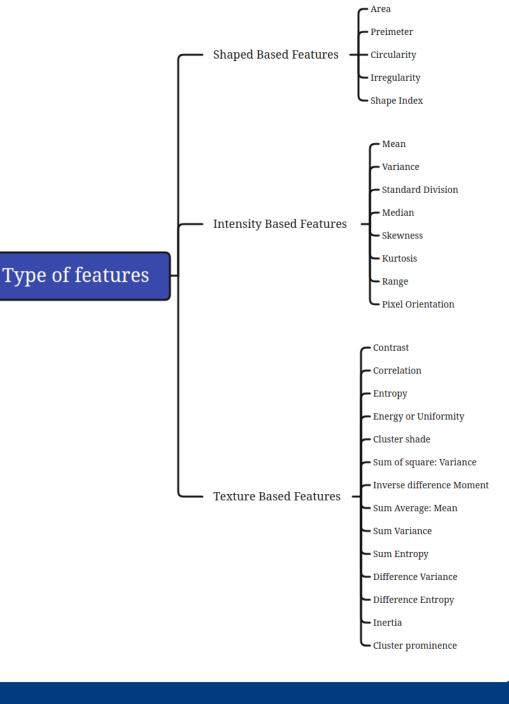
True Negatives $Specificity = \frac{1}{True\ Negatives + False\ Positives}$

Confusion Matrix of all image segmentation results

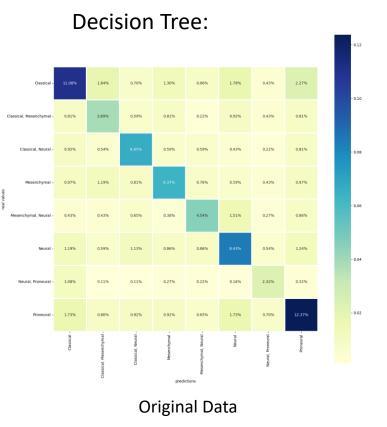
Feature Extraction

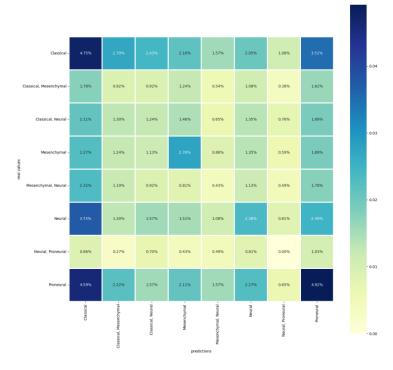
- In the feature Extraction procedure, we apply conventional CV method, using specific equation to calculate features to put image information from 2D to 1D
- At first, I use my own code to calculate features, but the speed is slow. Right now, we use pyradiomics module to extract features from ROI
- Pyradiomics have 120 features, most of them are texture features.
- The most classical texture feature design is using GLCM

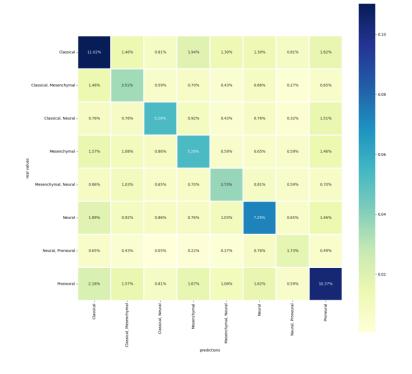




Classify GBM Subtypes







PCA Data (99% information)

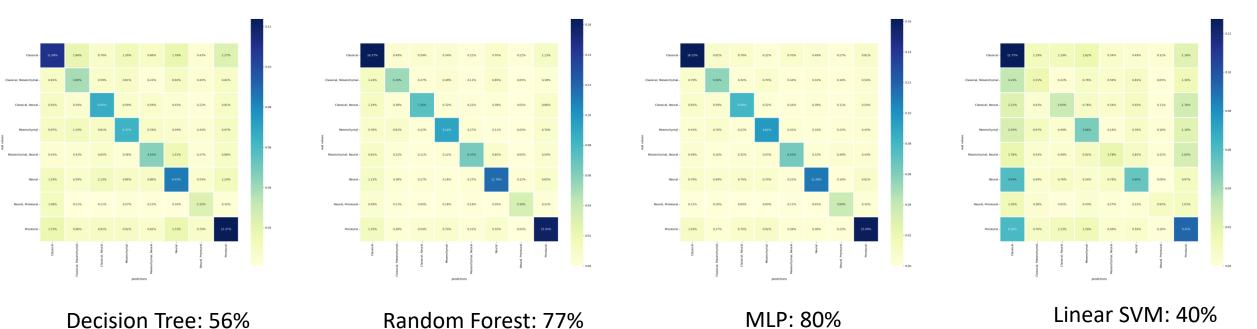
PCA Normalized Data

Need Normalize before PCA

Classify GBM Subtypes

Overview:

Decision Tree: 56%

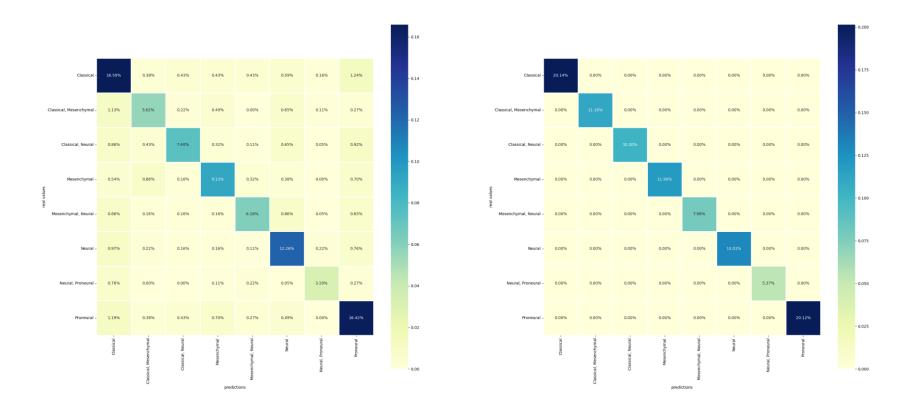


MLP have the best preference in classifying this dataset This dataset is not linear relationship

Random Forest: 77%

Classify GBM Subtypes Problems

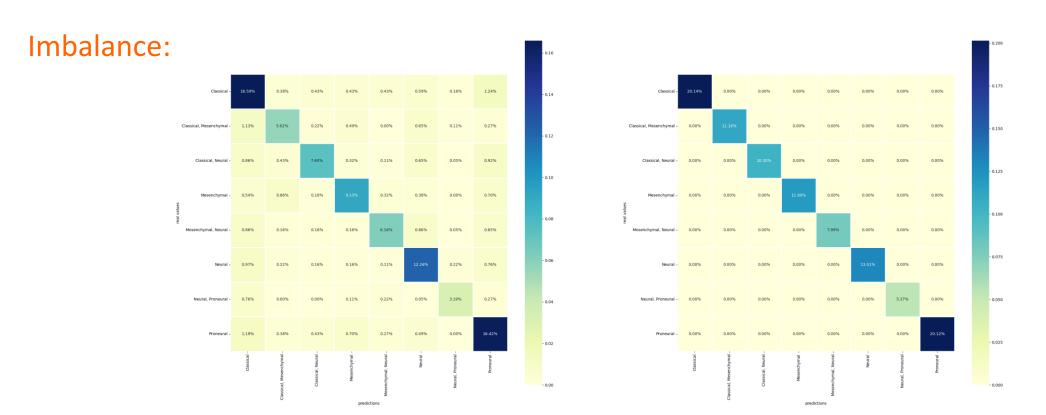




Even add random forest tree numbers into 500, the overfit question still exist (Train accuracy = 100%)

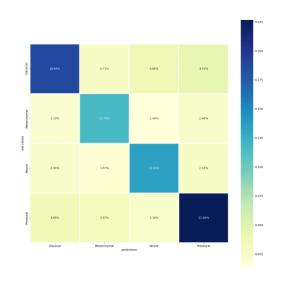
This problem exist in every algorithms I have ever tried

Classify GBM Subtypes Problems

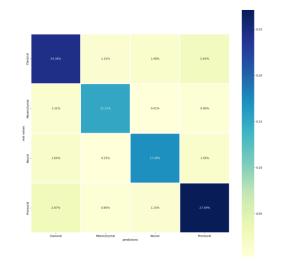


Data Imbalance, the Classical and Proneural is the most subtypes of GBM, leading to data imbalance, which may lead classical and proneural type prediction sensitivity is higher, but others will be lower

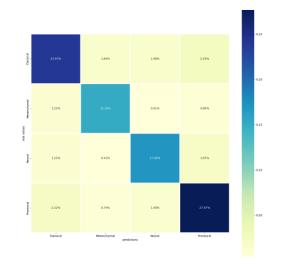
Classify GBM Subtypes Problems



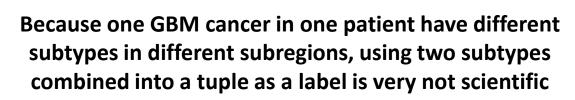


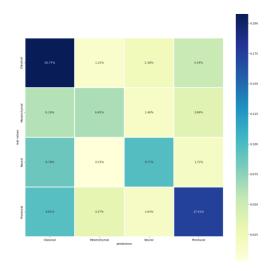


Random Forest: 84%



MLP: 88%



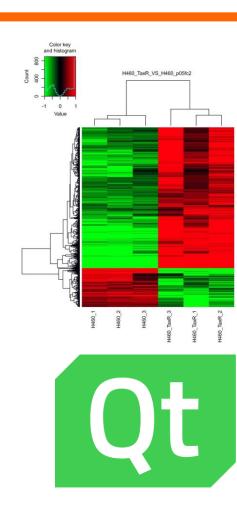


Linear SVM: 54%

Future Plan

Improvements and Other direction

- Using Transformer Network replace U-net (CNN) and Pyradiomics to do feature extraction to do classification
- Using gene data to map radiomic features with gene data.
- Predict key gene mutation or not.
- Learn Qt to make UI



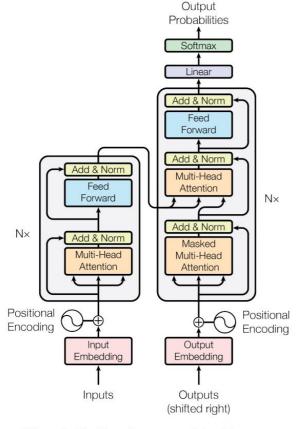


Figure 1: The Transformer - model architecture.