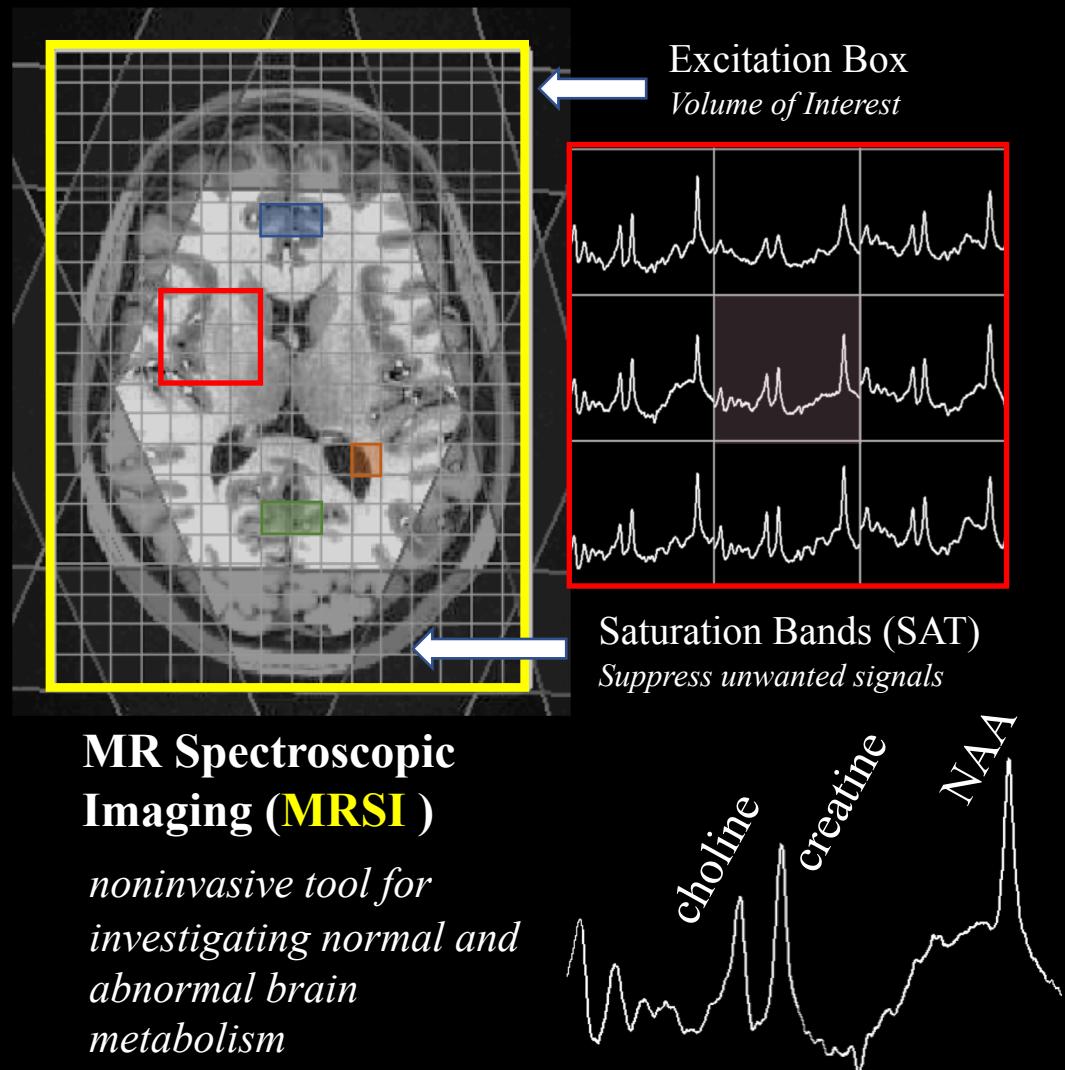


Deep Learning based Automatic Quality Control of MRSI



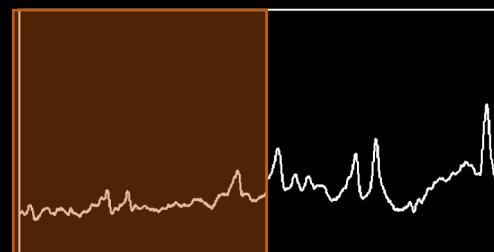
Emily Xie, Yan Li
2020 Radiology and ci² Summer Symposium
7/31/2020

Introduction

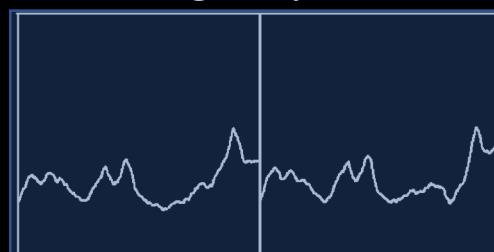


Artifacts in MRSI

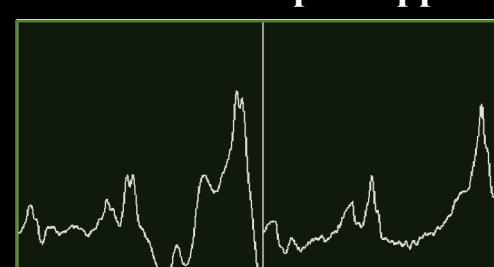
- **Tissue boundary**



- **Inhomogeneity**



- **Insufficient Lipid suppression**



- Water suppression
- Low SNR
- Post-processing errors ...

The **purpose** of this study is to **develop a deep learning model** to automatically **filter spectra with artifacts** in healthy controls and patients, greatly decreasing the time to review spectra quality.

Methods

Population	# Scans	Scanner	# Bad ¹	# Usable	Brain Atlas?
Healthy Controls	10	7T	2051	2330	yes
Major Depressive Disorder	10	7T	1698	2374	yes
Multiple Sclerosis	10	7T	2433	1547	yes
Parkinson's Disease	10	7T	2641	1689	yes
Brain Tumor	10	3T	2418	2557	no

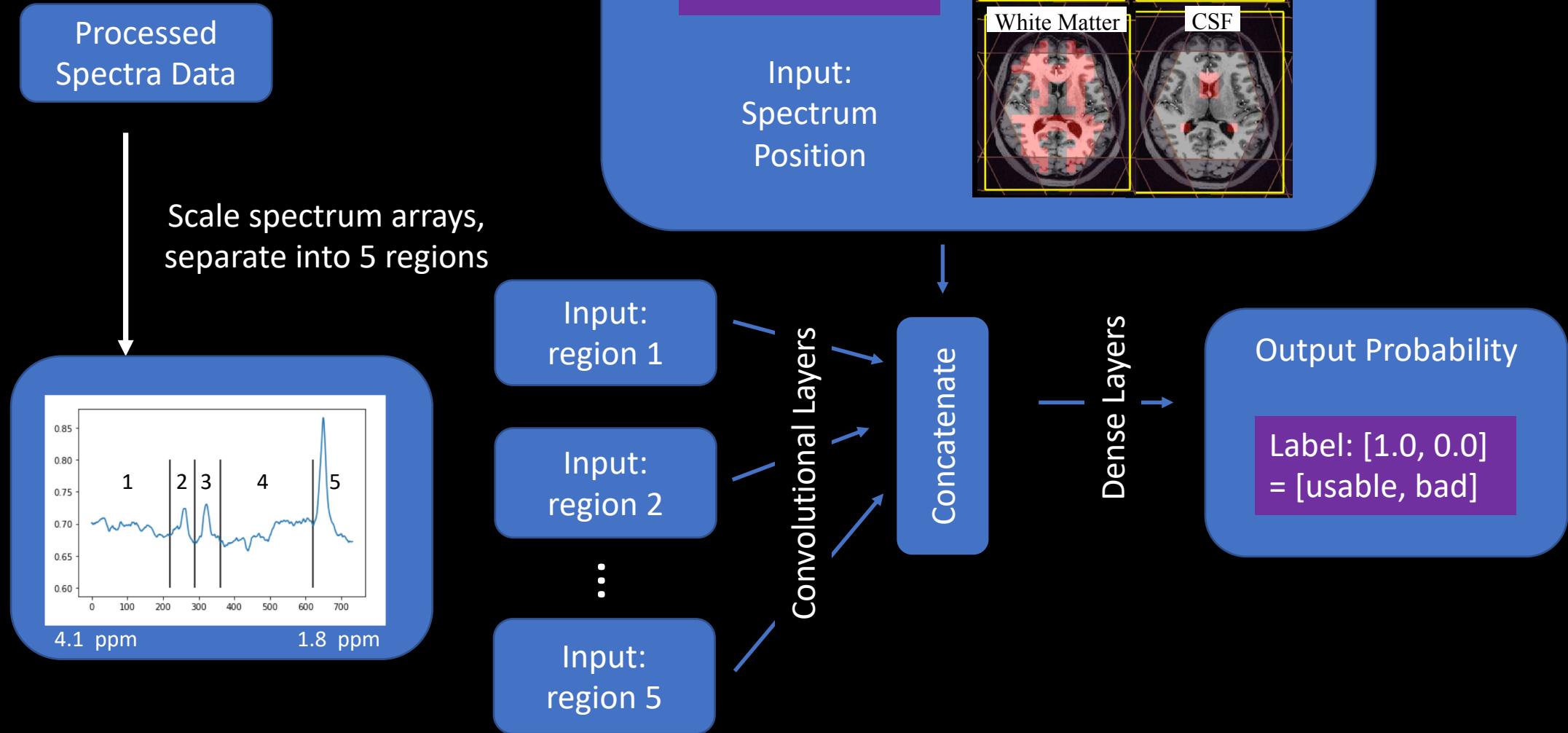
usable voxels: 10497

bad voxels: 11241

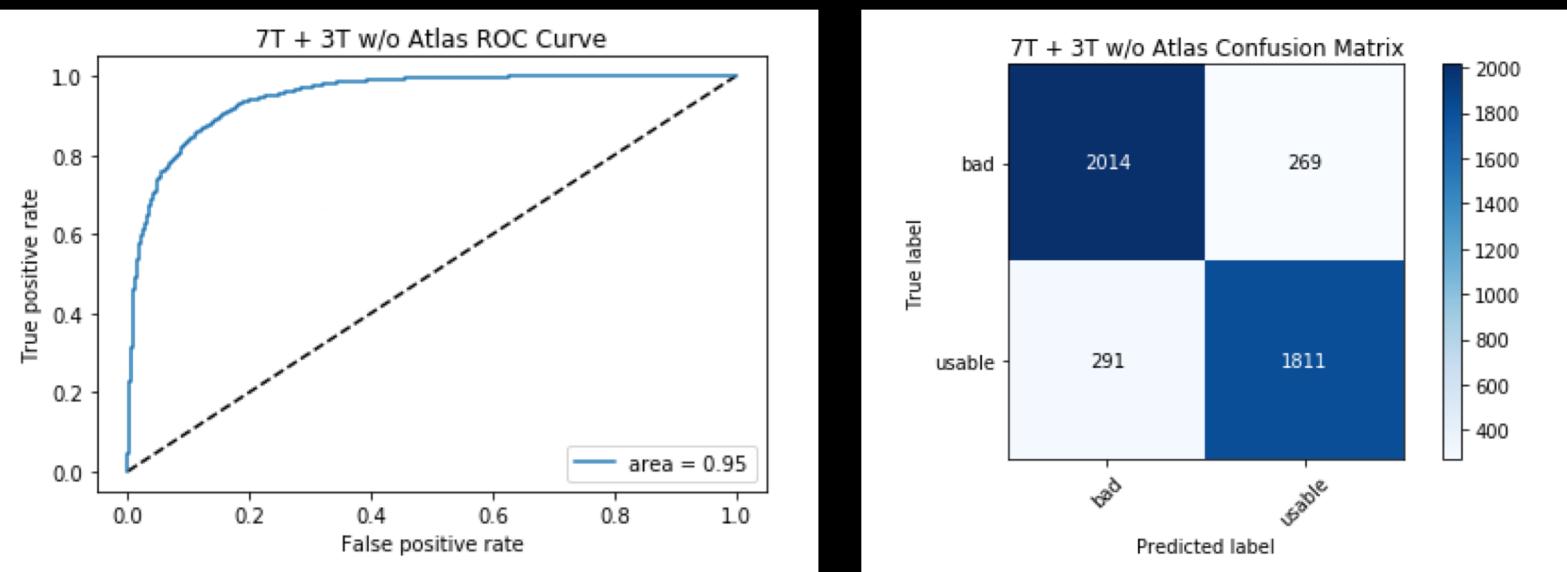
Total voxels: 21738

¹ Within the excited volume, bad voxels were defined by using LcModel CRLB, brain mask, and manual selection by Dr. Yan Li.

Methods



Results



Models	# Training Datasets	# Training voxels	Evaluation Datasets	Accuracy	ROC AUC	Sensitivity	Specificity
7T w/o brain atlas	32	10, 200	8	0.89	0.95	0.90	0.88
7T w/ brain atlas	32	10, 200	8	0.88	0.95	0.90	0.86
7T + 3T w/o brain atlas	40	13, 014	10	0.87	0.95	0.86	0.88
Published paper ²					0.95		

Discussion

The model trained with brain atlas information (7T w/ brain atlas), when compared to the model without brain atlas (7T w/o brain atlas), **barely saw a change in overall metrics.**

We have developed a model (7T + 3T w/o brain atlas), with **metrics comparable to published papers**, that is **able to filter artifacts in MRSI** for a **variety of patients** with **different scan types**.