



# Major Technical Project - Term 1 Presentation

Semester 1: August-December 2021

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Project Title

**MR simulator to create tailored  
datasets of MR Images for AI  
Algorithm Development.**



# Motivation/Background

- 01 Increasing use of MRI for detection of various anomalies like tumors/cancers , kidney stones, etc.
- 02 Human error may cause doctors to miss upon detecting these anomalies.
- 03 Training AI models to detect these anomalies is a challenge due to unavailability of sufficient labelled data.
- 04 Preparation of this data using real MRI machines is a costly and time intensive process.
- 05 Patient health/privacy concerns further add to the difficulties in obtaining large amounts of data for training AI/ML models.



# Literature Review

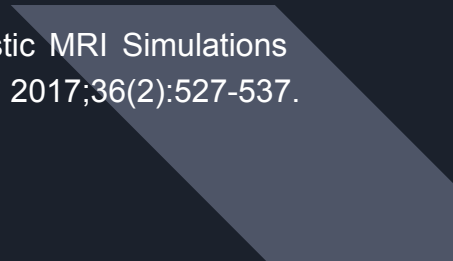
**MRiLab** is a numerical MRI simulator platform. It uses mathematical models called phantoms to imitate various tissues.

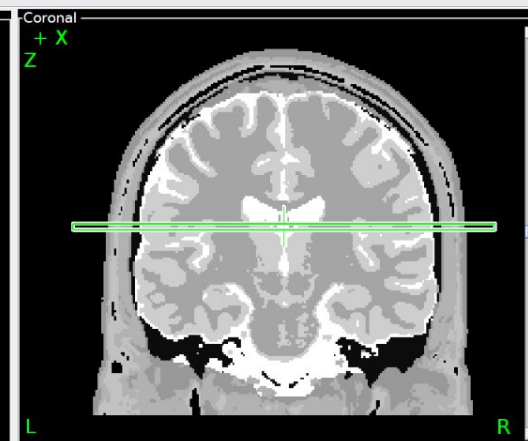
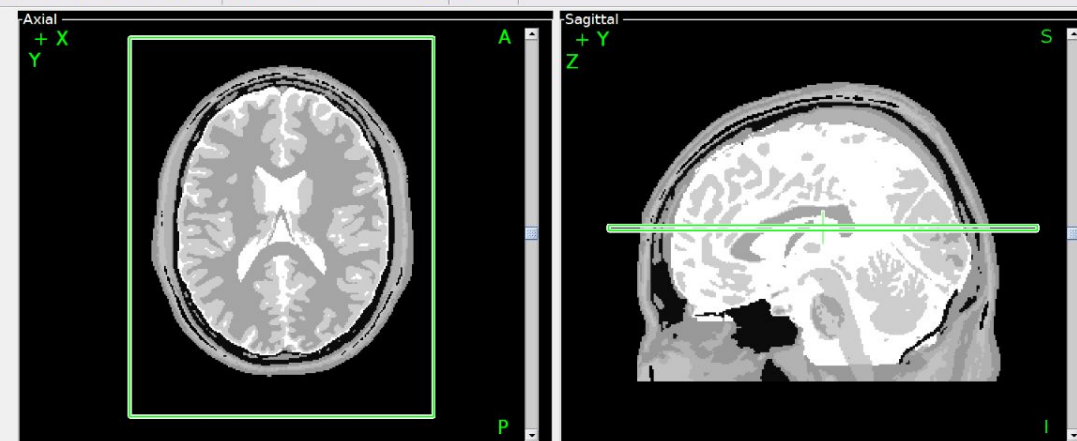
It has a friendly UI to facilitate controlling various parameters like Magnetic field, TE , TR etc. It provides us with multiple tools to design new phantoms and new sequences.

It is written in matlab with computational kernels coded in cpp.

Its uses GPU acceleration and MultiThreading to speed up the simulation process.

Reference: Liu F, Velikina JV, Block WF, Kijowski R, Samsonov AA. Fast Realistic MRI Simulations Based on Generalized Multi-Pool Exchange Tissue Model. *IEEE Trans Med Imaging*. 2017;36(2):527-537. doi:10.1109/TMI.2016.2620961

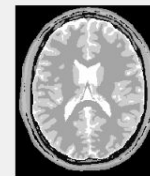




VObj Spin Map

Rho

Display



Max 1.00

Min 0.00

Size 216 180 180

Spin Type

&lt;Localizer

1

Virtual Object Property

Gyro

ChemShift

XDim

YDim

ZDim

XDimRes

YDimRes

ZDimRes

Scan Series



Simulation Settings

Imaging Advanced Hardware Recon

ResFreq 100

FreqDir A/P

FlipAng 30

FOVPhase 20e-2

FOVFreq 25e-2

BandWidth 10e3

TEPerTR 1

TE 25e-3

SliceThick 3e-3

SliceNum 1

ScanPlane Axial

ResPhase 80

TR 100e-3

Field of View

A/P -16

234

S/I 90

93

L/R -9

191

Motion

Gradient

Magnet

Coil

Sequence

Save

Update

Scan

\\_0\\_

PSD Memo [Save current settings into configuration files](#)

PSD Memo :

Simulation Info Log

11:16:49&gt;Viewing virtual object spin map details.

11:16:53&gt;Updating simulation setting parameters.

11:16:53&gt;Updating simulation setting parameters is complete.

PSD : ~

Est. Time Left : ~ : ~ : ~

SNR : 100%

SAR : ~

Channel # : SumofMagn

Echo # : 1

# Visible Problems/Limitations of MRiLab

01

Artifacts and mass distortions for certain ranges of values of various parameters .

02

Change in parameters which should ideally affect the SNR(signal to noise ratio) does not seem to have any effect.

03

No functionality to manually add tumors and or anomalies to the MR images.

04

T1 and T2\* times of materials depend on static magnetic field strength. If this feature is included it needs to be analysed.

Preview : Series1



**PSD : PSD GRE3D SNR : 100%**  
**Est. Time Left : ~ : SAR : ~**



## Semester 1:

Benchmarking MRiLab .

Run Experiments on MRiLab and compare the results obtained with expected results computed using standard equations. ( For Spoiled Gradient Echo Sequence)

Derive conclusions from the comparisons made in the above step,

## Semester 2:

Try to fix problems and irregularities found in semester 1.

Add functionality to simulate tumors in the phantoms.



# Experiments Conducted

- All Experiments were conducted using the Spoiled Gradient Echo Sequence.
  - All experiments were conducted for two Phantoms , Brain High Resolution Phantom and the Water-Fat Phantom.
  - All Experiments were conducted for three noise levels. Noise Level 0, 5, 10.
- 
1. Vary TE while keeping all other parameters constant (TE = 5 - 100 ms; steps: 10 ms (5,10,20,30,...) , TR = 500 ms, FA = 30°).
  2. Vary TR while keeping all other parameters constant (TR = 25 - 500 ms; steps: 25 ms) , TE = 5 ms, FA = 30°).
  3. Vary FA while keeping all other parameters constant (FA 5° - 90°, steps: 5°; TE = 5 ms, TR = 500ms).
  4. Vary FA while keeping all other parameters constant (FA 5° - 90°, steps: 5°; TE = 5 ms, TR = 40ms).





# Theoretical Signal

- The theoretical signal depends on the following properties of material:
  - $\rho$  which is the Proton Density
  - $T_1$  for the given tissue material
  - $T_2^*$  for the given tissue material
- The theoretical signal also depends on the following parameters:
  - $\alpha$  which is the flip angle
  - $TR$  which is the repetition time
  - $TE$  which is the echo time

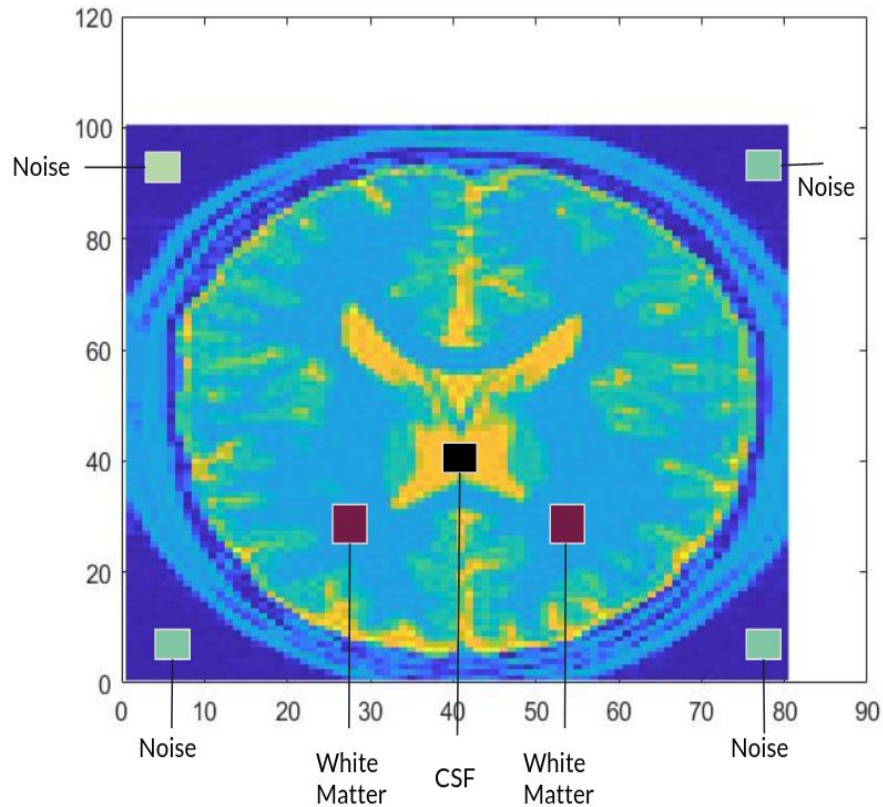
$$\text{Signal} = \rho \frac{\sin \alpha \cdot [1 - \exp(-TR/T_1)] \cdot \exp(-TE/T_2^*)}{1 - \cos \alpha \exp(-TR/T_1)}$$

# Computation of Signal/Noise

64 pixels from 4 patches for noise

16 pixels for CSF signal

32 pixels from 2 patches for White Matter  
Signal





# Signal to Noise Ratio (SNR) in MRiLab

- The SNR Computation of MRiLab is flawed.
- It uses the following equation, which does not changes on varying TE, TR, Flip angle.
- The problem was verified by manually calculating the SNR using the patches.
- Computation of Signal itself has multiple flaws (shown in future slides).

$$SNR = B0 \times RFreq \times RPhase \times RSlice \times \frac{\sqrt{ResFreq \times ResPhase \times SliceNum \times NEX}}{NoiseLevel \times \sqrt{BandWidth}}$$

where,

B0 = Magnetic Field; Noise Level = User-input std deviation of noise;

RFreq = FOV Freq/ResFreq; RPhase = FOV Phase/ResPhase; RSlice = SliceThick.

# Noise computation model of MRiLab

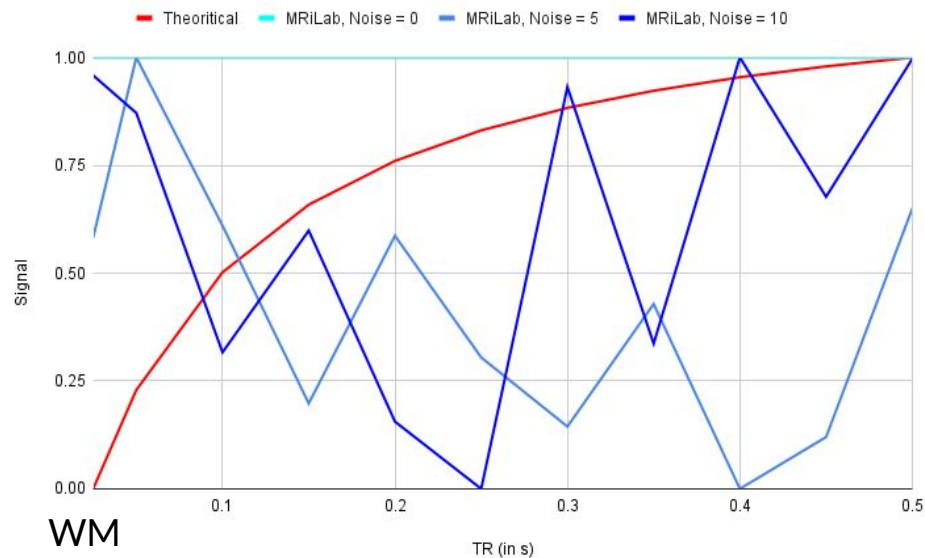
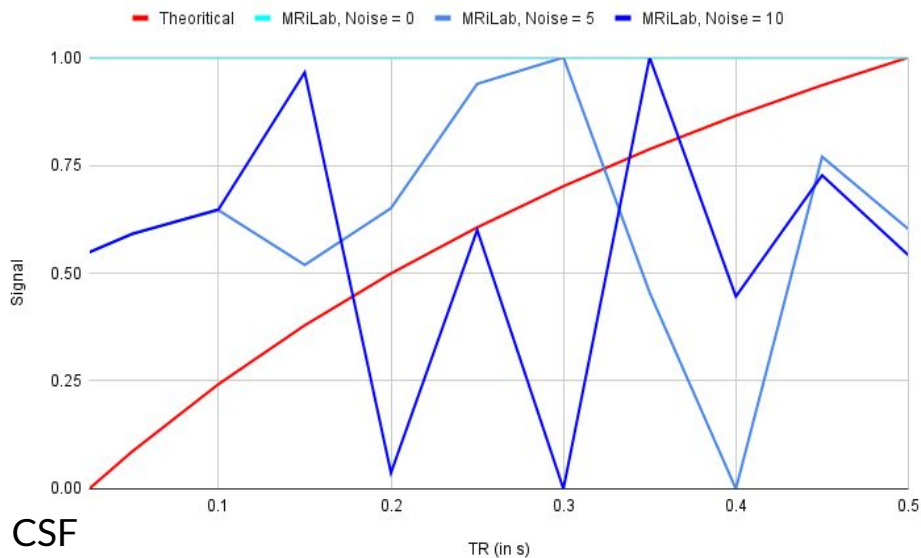
- Added noise is a Gaussian Noise with zero mean and user-defined standard deviation
- This standard deviation is referred to as Noise Level
- Noise is added to acquired k-space data

$$Noise = \frac{\frac{NoiseLevel}{NoiseRef} \times \sqrt{\frac{BandWidth}{BWRef}}}{\frac{B0}{B0Ref} \times \frac{RFreq \times RPhase \times RSlice}{VolRef} \times \sqrt{\frac{NEX}{NEXRef} \times \frac{ResFreq \times ResPhase \times SliceNum}{ADCRef}}}$$

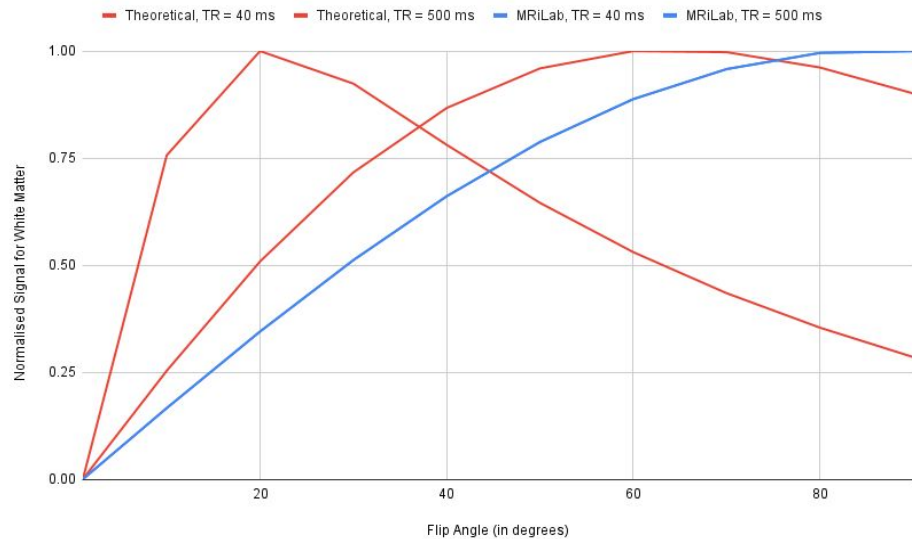
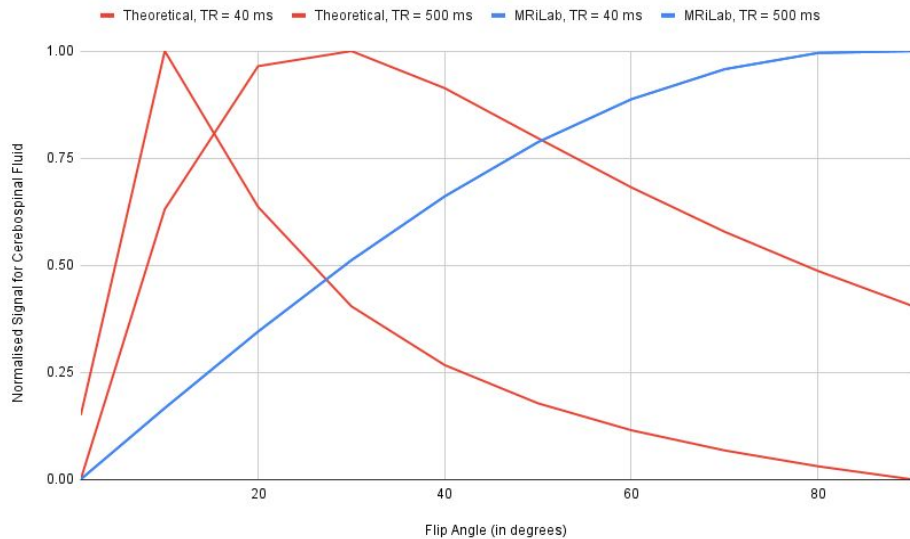
where,

BWRef = 1e3; NoiseRef = 1; B0Ref = 1.5; VolRef = 1e - 9; NEXRef = 1;  
ADCRef = 1e4; RFreq = FOV Freq/ResFreq; RPhase = FOV Phase/ResPhase; RSlice  
= SliceThick.

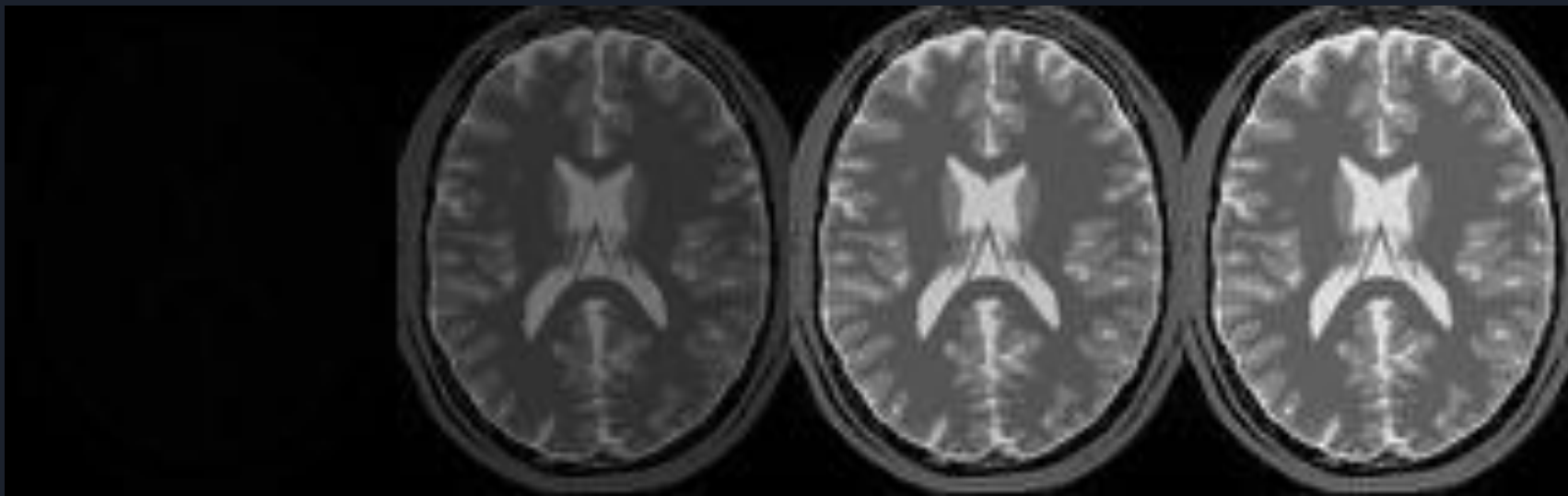
# Theoretical Vs Actual Normalised Signal on varying TR



# Theoretical Vs Actual Normalised Signal on varying Flip Angle



MRiLab Simulated images on increasing flip  
angle (left to right)





# Shortcomings

MRiLab (Spoiled Gradient Echo Sequence) was found to be unable to:

- Perform simulations for certain TE values ( $TE \geq 20\text{ms}$  at  $TR=500\text{ms}$  &  $\alpha = 30^\circ$ )
- Vary signal (constant signal value) with different TR values at Noise level = 0
- Resemble theoretical signal while varying TR values at any Noise level
- Vary the signal vs flip angle curve for different TR values
- Resemble theoretical signal while varying flip angle values
- Resemble basic MRI concept of maximum signal at Ernst angle while varying flip angle

**For the Spoiled Gradient Echo Sequence, we conclude that the MRiLab simulations couldn't model the signal accurately.**

**Since, the MRiLab simulations had multiple computation flaws in calculation of Signal, in the next semester we'll try to fix the computation of Signal.**



	Sem1				Sem2			
	August	September	October	November	February	March	April	May
Learn about MRI and working of MRIlab								
Experiments to determine Shortcomings of MRIlab								
Derice conclusions from Experiment Results								
Documentation								
Semester 1 Ends								
Fix problems found in benchmarking								
Add Functionality to add Tumors in phantoms								



## References

1. F. Liu, J. V. Velikina, W. F. Block, R. Kijowski and A. A. Samsonov, "Fast Realistic MRI Simulations Based on Generalized Multi-Pool Exchange Tissue Model," in *IEEE Transactions on Medical Imaging*, vol. 36, no. 2, pp. 527-537, Feb. 2017, doi: 10.1109/TMI.2016.2620961.
2. Willemink MJ, Koszek WA, Hardell C, et al. Preparing Medical Imaging Data for Machine Learning. *Radiology*. 2020;295(1):4-15.  
doi:10.1148/radiol.2020192224
3. Kohli, M.D., Summers, R.M. & Geis, J. Medical Image Data and Datasets in the Era of Machine Learning—Whitepaper from the 2016 C-MIMI Meeting Dataset Session. *J Digit Imaging* **30**, 392–399 (2017).  
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