

EEE 473 Medical Imaging Homework #4 Boran Kılıç 22103444

a)
$$\lambda = \lambda \int \beta_{1}^{e}(t) e^{-j2\pi(u-v_{e})t} e^{j2\pi ut}$$

$$= \lambda \int \beta_{1}^{e}(t) e^{-j2\pi(u-v_{e})t} e^{j2\pi ut}$$

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$$= \lambda \int \beta_{1}^{e}(t) e^{-j2\pi u_{e}} e^{-j2\pi u_{e$$

$$= \sum_{k=0}^{\infty} \frac{1}{\pi^{2}} e^{2k} = \sum_{k=0}^{\infty} \frac{1}{\pi^{2}} e^{2k$$

b)
$$\Delta V =
delta G_{\pm} \Delta 2 \Rightarrow G_{\pm} = \frac{\Delta V}{\mathcal{X} \Delta 2}$$

$$G_2 = \frac{1.7) LH_2}{6} = 0.1 \frac{mT}{cm} = 10 \frac{G}{cm}$$

$$42.58 \frac{MHB}{T} \cdot 4m$$

$$\mathcal{L}(2=0) = \mathcal{V}A \sqrt{\pi} = \frac{\pi}{3} = A = \frac{\pi/3}{2\pi \mathcal{V}\sqrt{\pi}} = \frac{\pi/3}{2\pi (42.58 \text{ MHz/}\tau) (\sqrt{\pi} 0.3 \sim s)}$$

$$= A = \frac{1}{6\sqrt{\pi} \cdot 42.58 \times 10^{6} \cdot 0.3 \times 10^{5}} = 0.0736 \text{ G}$$

Q2) Let us food the pathern:

$$M_{2}^{\circ}(t) = M_{\circ}(1-e^{-\frac{t}{T_{\circ}}})_{1} M_{\circ} \cos(\alpha) e^{-\frac{t}{T_{\circ}}}$$

for $t \in [0^{t}, T_{k}]$

$$M_{2}^{\prime}(f) = M_{1}\left(1 - e^{-\frac{f-T_{R}}{T_{1}}}\right) + M_{2}^{\circ}\left(T_{R}^{-}\right)\cos\left(\lambda\right)e^{-\frac{\left(f-T_{R}\right)}{T_{1}}}$$
 for $f \in \left[T_{R}^{-}, 2T_{R}^{-}\right]$

$$\mathcal{M}_{2}^{n}(t) = \mathcal{M}_{0}\left(1 - e^{\frac{-t - nT_{k}}{T_{1}}}\right) + \mathcal{M}_{2}^{n+1}\left(nT_{k}^{-}\right)\cos\left(\alpha\right) e^{\frac{-t - nT_{k}}{T_{1}}} \qquad \text{for } t \in \left[nT_{k}^{+}, \left(n\pi\right)T_{k}^{-}\right]$$

In order to reach steady - state:

$$M_2^{55} = M_2^{0-1} (n T_E^-) = M_2^0 ((a+1) T_E^-)$$

$$M_2^{ss} = M_0 \left(1 - e^{-\frac{TR}{T_1}} \right)_T M_2^{ss} \cos(\alpha) e^{-\frac{TR}{T_1}}$$

=)
$$M_2^{55}(1-\cos(x)e^{-\frac{\tau_e}{\tau_i}})=M_0(1-e^{-\frac{\tau_e}{\tau_i}})$$

=>
$$M_{E}^{55} = M_{0} \frac{1 - e^{-\frac{T_{k}}{T_{1}}}}{1 - \cos(\kappa) e^{-\frac{T_{k}}{T_{1}}}}$$

$$\beta_{1}^{e}(t)$$

$$\frac{-3}{\Delta V} \qquad \frac{-2}{\Delta V} \qquad \frac{1}{\Delta V} \qquad \frac{3}{\Delta V} \qquad t$$

$$\frac{6}{\Delta V} = T_p = 4 ms$$

$$\Delta V = \chi G_2 \Delta^2 = G_2 = \frac{\Delta V}{\chi \Delta^2} = \frac{1.5 L H_2}{42.58 \frac{MH_2}{T} \cdot 3mn} = G_2 = 11.74 \frac{mT}{m}$$

$$\alpha = X \int_{1}^{2} \beta_{1}^{e}(t) dt = X \int_{1}^{2} A_{siac}(1.5t) dt = XA 0.71 = \frac{\pi}{2}$$

$$A = \frac{\pi/2}{2\pi \cdot 42.58 \frac{MH_2}{T} \cdot 0.31rs} = \frac{1}{4 \cdot 42.58 \times 10^6 \cdot 0.21 \times 10^{-3}} = \frac{8.22 \,\text{pT}}{4 \cdot 42.58 \times 10^6 \cdot 0.21 \times 10^{-3}}$$

=>
$$G_{x} = \frac{1}{\sqrt{FOV_{x}T}} = \frac{1}{(42.58 \frac{MH^{2}}{T})(16 cm)(845)} = 0.18 \frac{mT}{cm}$$

d)
$$B_{1}(t) = b_{1}^{e}(t) e^{-j2\pi v_{1}t}$$
 where $v_{1} = v_{2} + f = 2$

$$v_{2} = l_{1}$$

$$v_{1} = f = f = 11.74 \frac{r}{r}$$

$$v_{2} = l_{2} = l_{2}$$

$$v_{3} = f = 11.74 \frac{r}{r}$$

$$v_{4} = 127.745 \text{ MH2}$$

$$\frac{|MG_{1}|}{|MG_{2}|} = \frac{AM_{0}(x,y) \sin(\alpha) \exp\left(-\frac{\tau E_{1}}{\tau_{2}(x,y)}\right)}{AM_{0}(x,y) \sin(\alpha) \exp\left(\frac{-\tau E_{2}}{\tau_{2}(x,y)}\right)} = \frac{\exp\left(-\frac{\tau E_{1}}{\tau_{2}(x,y)}\right)}{\exp\left(-\frac{\tau E_{2}}{\tau_{2}(x,y)}\right)} = \exp\left(\frac{\tau E_{2} - \tau E_{1}}{\tau_{2}(x,y)}\right)$$

$$\ln\left(\frac{1MG_1}{1MG_2}\right) = \frac{TE_2 - TE_1}{T_2(x,y)}$$

$$T_2(x,y) = \frac{TE_2 - TE_1}{\ln\left(\frac{1MG_1}{1MG_2}\right)}$$

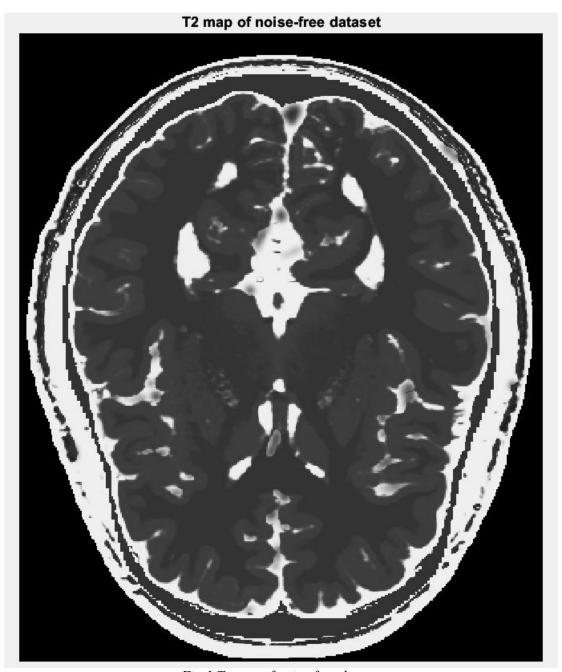


Fig.1 T₂ map of noise-free dataset

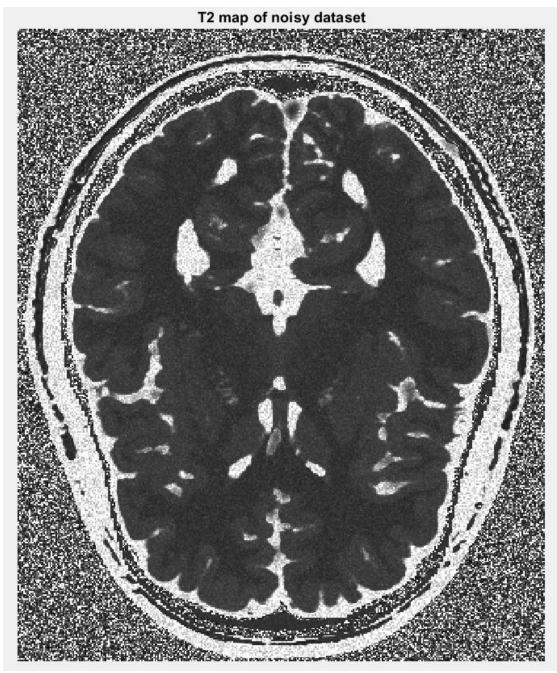


Fig.2 T₂ map of noisy dataset

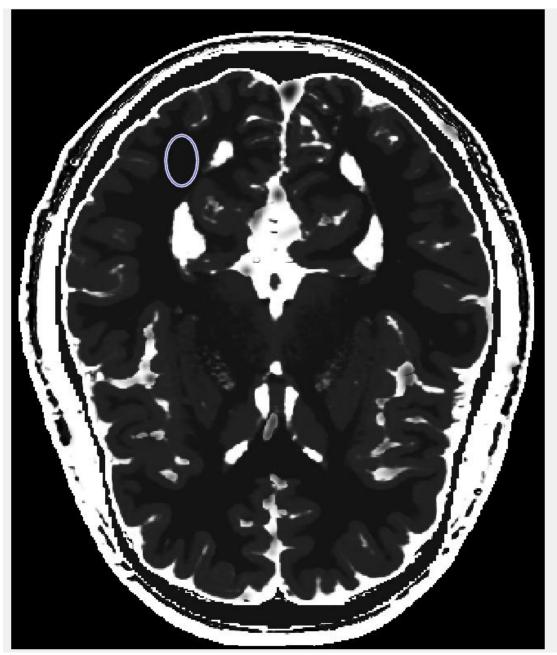


Fig. 3 T₂ map with an ellipse on a region that contains white matter only

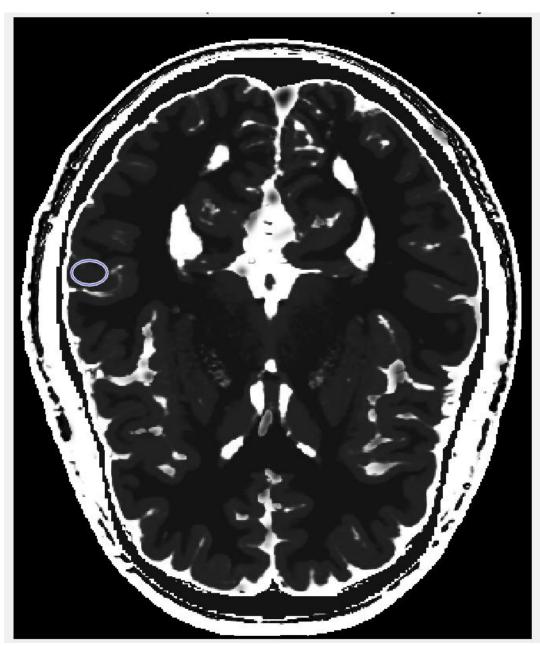


Fig. 4 T_2 map with an ellipse on a region that contains gray matter only

f)

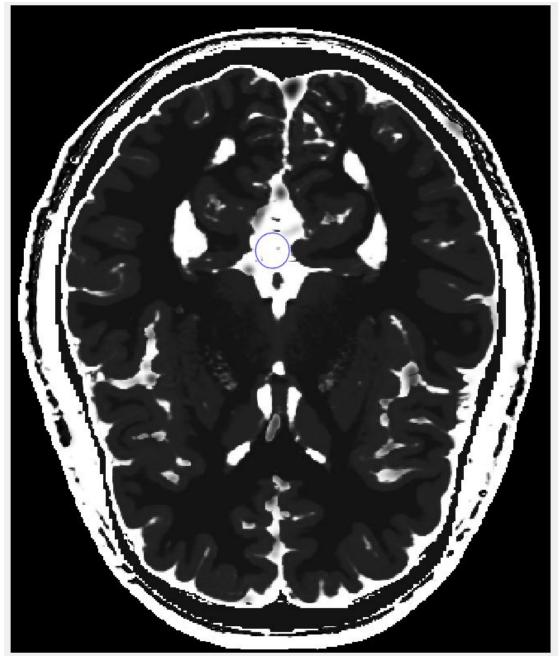


Fig. 5 T₂ map with an ellipse on a region that contains CSF (cerebrospinal fluid) only

g)

Table.1 T₂ estimations and deviation percentages

T ₂ estimations (ms)	Noise-Free Dataset	Noisy Dataset	Error (%)
White matter	70.0911	70.7212	0.8989
Gray matter	84.8766	86.1736	1.5280
CSF	324.5989	333.1363	2.6301

Table.1 depicts that the T₂ estimation of the CSF showed the biggest deviation in the noisy dataset compared to noise-free dataset with around 2.63% deviation. This higher deviation can be attributed to the inherent characteristics of CSF, which has a

longer T2 relaxation time. As a result, the signal is more susceptible to noise interference, particularly in the presence of Gaussian noise. Additionally, the logarithmic calculation in the T2 mapping amplifies small differences in signal intensity, leading to a greater deviation for regions with high T2 values. In contrast, white matter, with its lower T2 value, exhibited the smallest deviation of 0.89%, as its signal is less affected by noise. These findings underscore the need for effective noise reduction techniques, particularly for tissues like CSF.

Appendix - MATLAB code

PART4.m

```
close all; clear; clc
load('brainT2_mri.mat');

TE1=60;
TE2=120;
T2map=(TE2-TE1)./log(image1./image2);
T2map_noisy=(TE2-TE1)./log(image1_noisy./image2_noisy);

figure; imshow(abs(T2map),[0,350]);
title("T2 map of noise-free dataset");

figure; imshow(abs(T2map_noisy),[0,350]);
title("T2 map of noisy dataset");

figure; imshow(T2map,[])
mask=roiellipse;

T2_est=mean(T2map(mask))
T2_est_noisy=mean(T2map_noisy(mask))
error = (abs(T2_est - T2_est_noisy)/T2_est)*100
```