## EE16B Designing Information Devices and Systems II

Lecture 14B Last Lecture

#### Intro

- Last time:
  - Change of basis
  - Frequency analysis through projections onto complex harmonics
  - Discrete Fourier Transform

 Today – Wrap up DFT How MRI works

#### Change of Coordinates (Basis)

 We can compute new coordinates by projections onto orthonormal basis vectors





#### How can we find the frequency of this N=32 length signal?



Project on unit sinusoidal vectors?

#### Frequency Analysis Through Projections

N-length normalized discrete frequency:



## $u_{\omega}[n] = \frac{1}{\sqrt{N}} e^{j\omega n} \qquad 0 \le n < N \qquad 0 \le \omega < 2\pi$

$$= \sum_{n=0}^{N-1} x[n]e^{-j\omega n}$$

#### Also the DTFT of the finite sequence x

#### DFT vs DTFT

$$\vec{u}_k = \frac{1}{\sqrt{N}} \begin{bmatrix} e^{j\frac{2\pi k \cdot 0}{N}} \\ e^{j\frac{2\pi k \cdot 1}{N}} \\ \vdots \\ e^{j\frac{2\pi k \cdot (N-1)}{N}} \end{bmatrix}$$

$$X[k] = \vec{u}_k^* \vec{x}$$
$$X[k] = \frac{1}{N-1} \sum_{n=0}^{N-1} x[n] e^{-j\frac{2\pi kn}{N}}$$

 $\vec{u}_{\omega} = \frac{1}{\sqrt{N}} \begin{bmatrix} e^{j\omega 0} \\ e^{j\omega 1} \\ \vdots \\ e^{j\omega(N-1)} \end{bmatrix}$ 

 $X(\omega) = \vec{u}_{u}, \vec{x}$ 

N-1 $X(\omega) = \sum_{w} x[n]e^{-j\omega n}$ n=0





DFT

 $\vec{u}_{k} = \frac{1}{\sqrt{N}} \begin{bmatrix} e^{j\frac{2\pi k \cdot 0}{N}} \\ e^{j\frac{2\pi k \cdot 1}{N}} \\ \vdots \\ e^{j\frac{2\pi k \cdot (N-1)}{N}} \end{bmatrix} = \frac{1}{\sqrt{N}} \begin{bmatrix} W_{N}^{k \cdot 0} \\ W_{N}^{k \cdot 1} \\ \vdots \\ W_{N}^{k \cdot (N-1)} \end{bmatrix}$  $\Rightarrow X[k] = \vec{u}_k^* \vec{x}$ 

 $\vec{u}_0$ 



#### DH

#### • DFT Analysis

X[k]

## X[0]

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 $\vec{X} = F^* \vec{x}$ 

N-1 $\frac{1}{\sqrt{N}} \sum_{n=0}^{\infty} x[n] W_N^{-nk}$ 

#### 







 $\vec{u}_{0}^{*}\vec{x} =$ 



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 $\vec{u}_1 =$ 

 $\vec{u}_1^* \vec{x} =$ 



#### Example cont



#### Example



#### Example





#### Example





#### Complexity computing thf DFT

• What's the complexity to compute the DFT?



#### Exploit structure in W<sub>N</sub><sup>nk</sup> to speed up! —The Fast Fourier transform (FFT)

A:  $O(N \log(N))$ 

# $\vec{X} = \begin{bmatrix} | & | & | & | \\ \vec{u}_0 & \vec{u}_1 & \cdots & \vec{u}_{N-1} \\ | & | & | & | \end{bmatrix}^* \vec{x}$ $\stackrel{\Delta}{=} F^*$

A: Generally O(N<sup>2</sup>)

## MRI vs CT MRI is VERY VERY different from CT

#### CT

Based on Magnetism No moving parts No ionizing radiation Sensitive to soft tissue Complicated to operate

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#### MRI

#### Based on X-ray Rapidly moving parts Uses ionizing radiation Less sensitive to soft tissue Easy to operate



## How Does MRI Work?

- Magnetic Polarization -- Very strong uniform magnet
- Excitation -- Very powerful RF transmitter
- Acquisition -- Very powerful audio amps

-- Location is encoded by gradient magnetic fields

## Polarization

 Protons have a magnetic moment Protons have spins Like rotating magnets

## Polarization

- Body has a lot of protons
- magnetization



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#### In a strong magnetic field B0, spins align with B0 giving a net





## Polarizing Magnet

- 0.1 to 12 Tesla
- 0.5 to 3 T common
- 1 T is 10,000 Gauss
- Earth's field is 0.5G
- Typically a superconducting magnet









## Typical MRI Scanner











## Polarizaion







### Free Precession

- Much like a spinning top
- Frequency proportional to the field
- f = 127MhZ @ 3T



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#### MIT physics demos



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# $\begin{array}{c|ccc} \gamma B_0 & 0 \\ 0 & 0 \\ 0 & 0 \end{array} \left[ \begin{array}{c} M_x(\vec{r},t) \\ M_y(\vec{r},t) \\ M_z(\vec{r},t) \end{array} \right]$



#### Free Precession

## Precession induces magnetic flux Flux induces voltage in a coil



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 $\vec{y}(t) = \int_{\vec{R}} CM(\vec{r}, t) d\vec{r}$  $C = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$ 

#### Signal





isk space re	emains fo	r recording 761 hours and 21 minutes	
oject rate:	8192	Cursor: 0:00.000000 min:sec [Snap-To Off]	





courtesy Boris Keil, Larry Wald, MGH

## Intro to MRI - The NMR signal

- Signal from <sup>1</sup>H (mostly water)
- Magnetic field  $\Rightarrow$  Magnetization
- Radio frequency  $\Rightarrow$  Excitation
- Frequency  $\propto$  Magnetic field

![](_page_29_Figure_5.jpeg)

![](_page_29_Figure_7.jpeg)

## Intro to MRI - The NMR signal

- Signal from <sup>1</sup>H (mostly water)
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![](_page_30_Figure_5.jpeg)

![](_page_30_Figure_7.jpeg)

## Intro to MRI - Imaging

•  $B_0$  Missing spatial information

![](_page_31_Figure_2.jpeg)

## Intro to MRI - Imaging

- B<sub>0</sub> Missing spatial information
- Add gradient field, G

![](_page_32_Figure_3.jpeg)

## Intro to MRI - Imaging

- B<sub>0</sub> Missing spatial information
- Add gradient field, G
- Mapping: spatial position  $\Rightarrow$  frequency

![](_page_33_Picture_4.jpeg)

#### MRI Pulse Sequence

![](_page_34_Figure_1.jpeg)

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#### Repeat n times rate = TR seconds

## MR Imaging

magnitude k-space (Raw Data)

**Discrete Fourier transform** 

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#### Fourier

![](_page_35_Picture_5.jpeg)

#### Image

Video courtesy Brian Hargreaves

![](_page_35_Picture_9.jpeg)

#### MRI is all about contrast.....

![](_page_36_Picture_2.jpeg)

Relaxation

![](_page_37_Picture_2.jpeg)

 $\begin{bmatrix} \dot{M}_x(\vec{r},t) \\ \dot{M}_y(\vec{r},t) \\ \dot{M}_z(\vec{r},t) \end{bmatrix} = \begin{bmatrix} -\frac{1}{T_2} & 0 & 0 \\ 0 & -\frac{1}{T_2} & 0 \\ 0 & 0 & -\frac{1}{T_1} \end{bmatrix} \begin{bmatrix} M_x(\vec{r},t) \\ M_y(\vec{r},t) \\ M_z(\vec{r},t) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ \frac{1}{T_1} \end{bmatrix} M_0(\vec{r})$ 

## The Toilette Analogy (©2009 Al Macovski)

• Excitation = Flush

- Dynamics:
  - Water drains = signal decays, equivalent to T2
  - Tank refills = Magn. recovers, equivalent to T1
- Observed signal = water in the bowl
- Different toilettes = different tissues

to T2 o T1

he bowl tissues

![](_page_38_Picture_10.jpeg)

## The Toilette Analogy (©2009 Al Macovski)

## T2 Weighting

Graphics courtesy of Brian Hargreaves

![](_page_39_Picture_4.jpeg)

## The Toilette Analogy (©2009 Al Macovski)

#### T1 Weighting

#### Graphics courtesy of Brian Hargreaves

![](_page_40_Picture_4.jpeg)

### .... House Prefers T2

You -- Get cervical, thoracic and lumbar T2 weighted Fast Spin-Echo MRIs

![](_page_41_Picture_3.jpeg)

#### Summary

- MRI is about the interactions of magnetic fields with Nuclear spins
- Governed by a linear dynamical system! Dynamics result in rotations, which frequency depend
- on position
  - Decode via DFT!
- Damping causes exponential relaxation which we use to set the image contrast!
- To maximize signal in antenna we use LC resonance!

![](_page_43_Picture_0.jpeg)

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![](_page_43_Picture_2.jpeg)

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![](_page_43_Picture_32.jpeg)

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![](_page_43_Picture_35.jpeg)

![](_page_43_Picture_36.jpeg)

![](_page_43_Picture_37.jpeg)