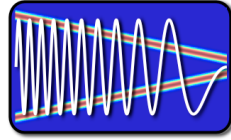


EE123



Digital Signal Processing

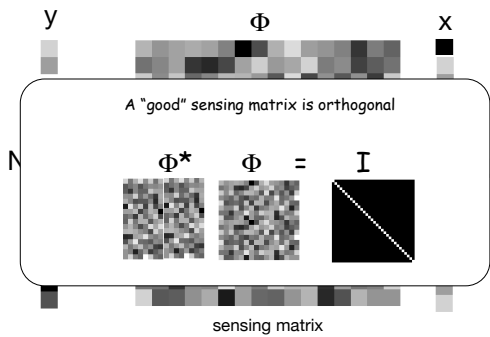
Lecture 24 Compressed Sensing III

RADIOS

- <https://inst.eecs.berkeley.edu/~ee123/sp15/radio.html>
- Interfaces and radios on Wednesday -- please come to pick up
- Midterm II this Friday -- same deal - open everything covers everything including 2D

Traditional Sensing

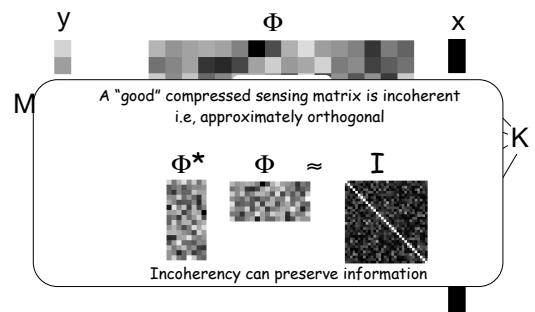
- $x \in \mathbb{R}^N$ is a signal
- Make N linear measurements



Compressed Sensing

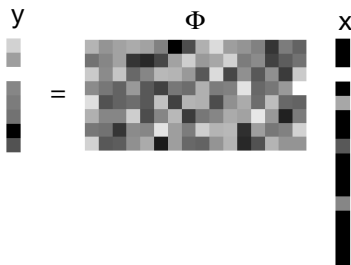
(Candes, Romber, Tao 2006; Donoho 2006)

- $x \in \mathbb{R}^N$ is a K -sparse signal ($K \ll N$)
- Make M ($K < M \ll N$) incoherent linear projections



CS recovery

- Given $y = \Phi x$
find x } Under-determined
- But there's hope, x is sparse!



CS recovery

- Given $y = \Phi x$
find x } Under-determined
- But there's hope, x is sparse!

CS recovery

- Given $y = \Phi x$
find x } Under-determined
- But there's hope, x is sparse!

$$\begin{aligned} &\text{minimize } \|x\|_2 \\ &\text{s.t. } y = \Phi x \end{aligned}$$

WRONG!

CS recovery

- Given $y = \Phi x$
find x } Under-determined
- But there's hope, x is sparse!

$$\begin{aligned} &\text{minimize } \|x\|_0 \\ &\text{s.t. } y = \Phi x \end{aligned}$$

HARD!

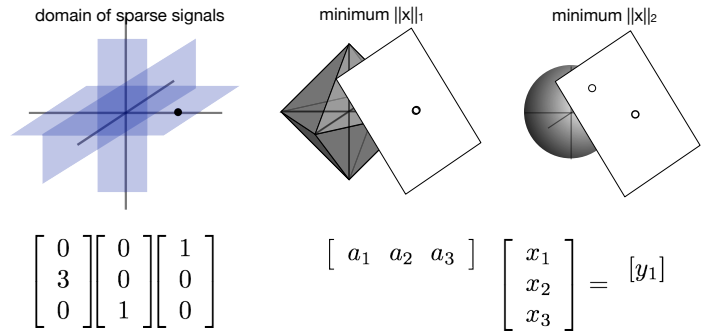
CS recovery

- Given $y = \Phi x$
find x } Under-determined
- But there's hope, x is sparse!

$$\begin{aligned} &\text{minimize } \|x\|_1 \\ &\text{s.t. } y = \Phi x \end{aligned}$$

need $M \approx K \log(N) \ll N$
Solved by linear-programming

Geometric Interpretation

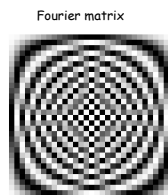


A non-linear sampling theorem

- $f \in C^N$ supported on a set Ω in Fourier
- Shannon:
 - Ω is known connected set, size B
 - Exact recovery from B equispaced time samples
 - Linear reconstruction by sinc interpolation
- Non-linear sampling theorem
 - Ω is an arbitrary, unknown set of size B
 - Exact recovery from $\sim B \log N$ (almost) arbitrary placed samples
 - Nonlinear reconstruction by convex programming

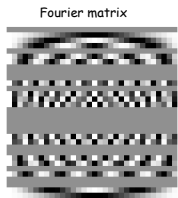
Practicality of CS

- Can such sensing system exist in practice?



Practicality of CS

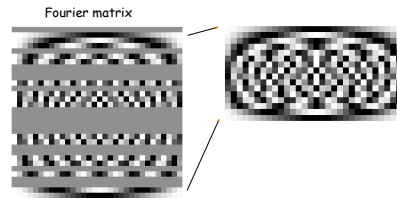
- Can such sensing system exist in practice?



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Practicality of CS

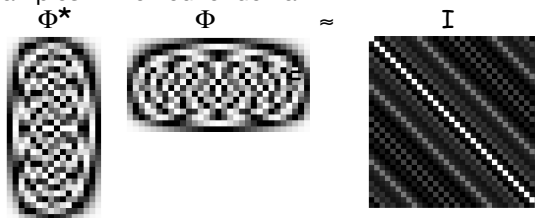
- Can such sensing system exist in practice?



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Practicality of CS

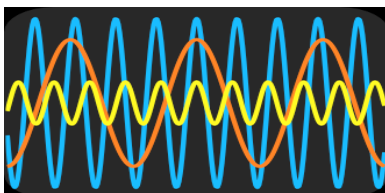
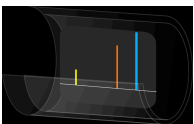
- Can such sensing system exist in practice?
- Randomly undersampled Fourier is incoherent
- MRI samples in the Fourier domain!



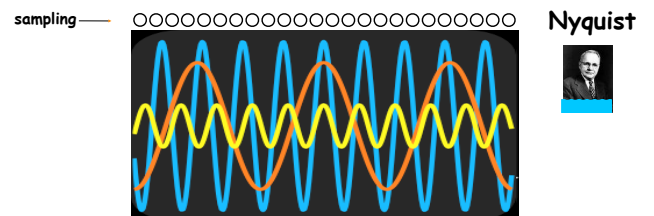
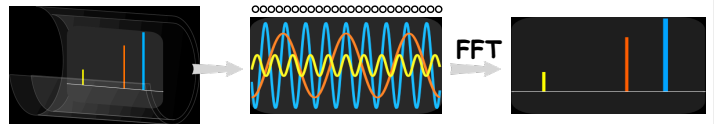
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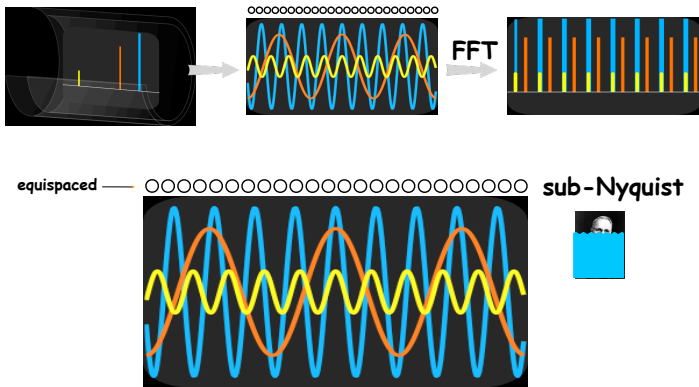
Intuitive example of CS



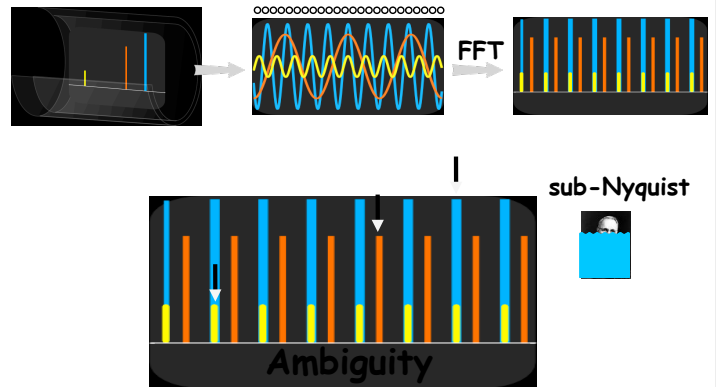
Intuitive example of CS



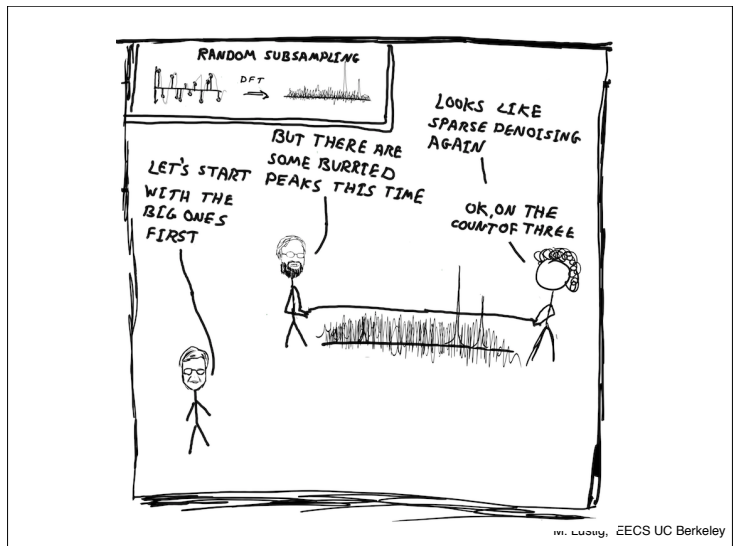
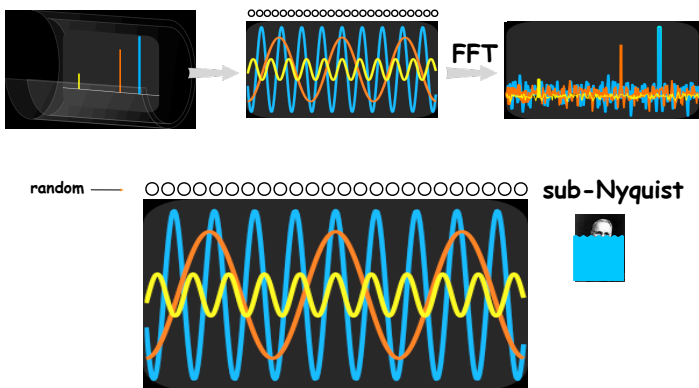
Intuitive example of CS



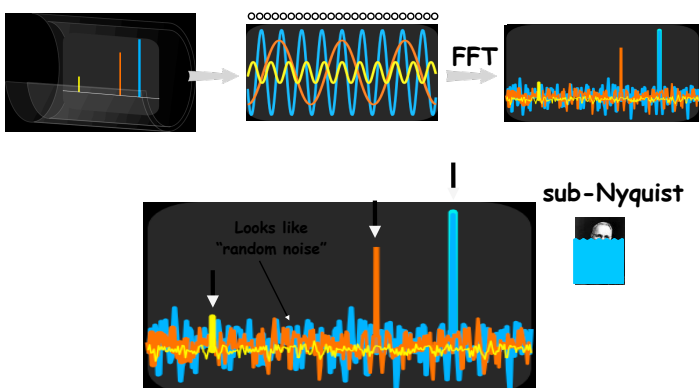
Intuitive example of CS



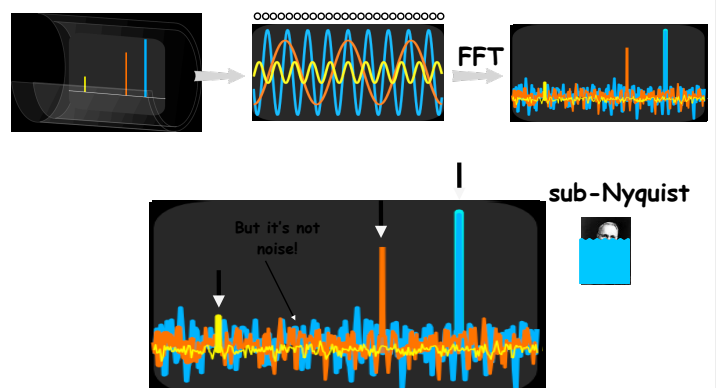
Intuitive example of CS

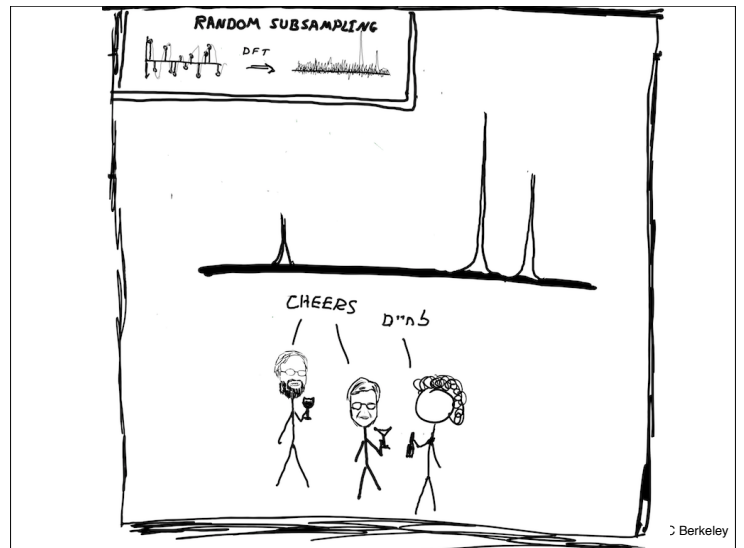
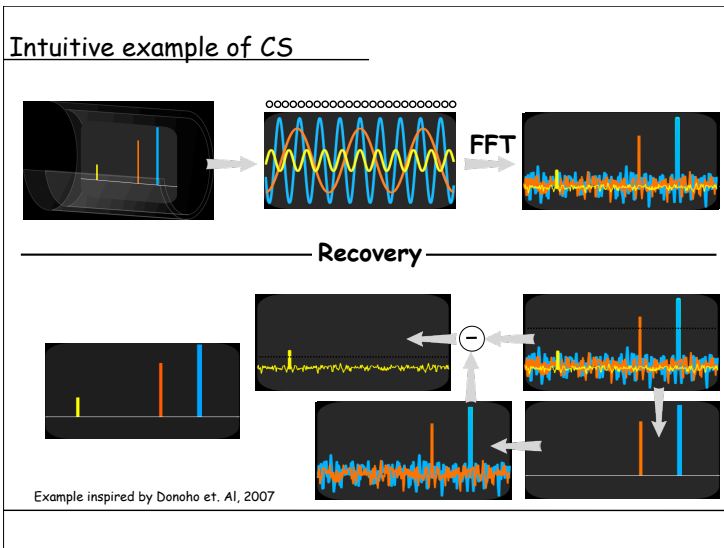
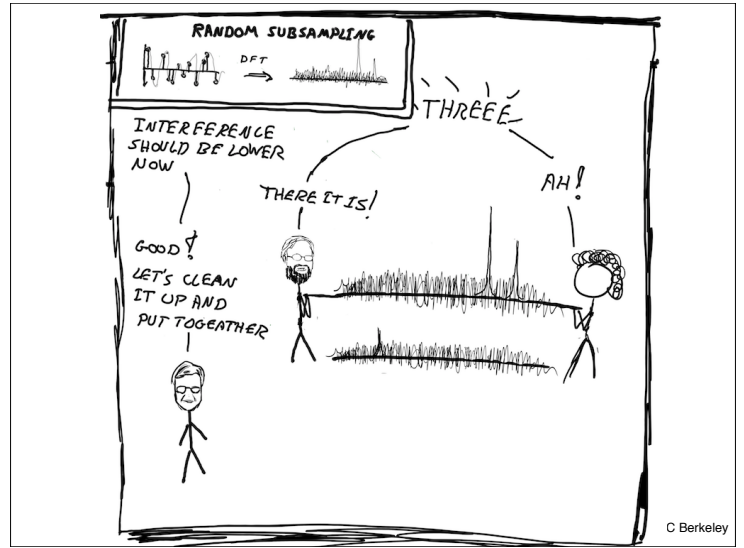
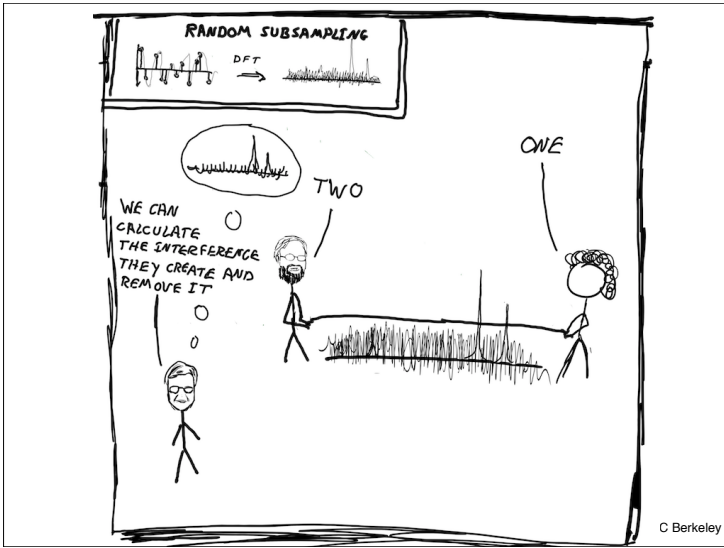


Intuitive example of CS



Intuitive example of CS



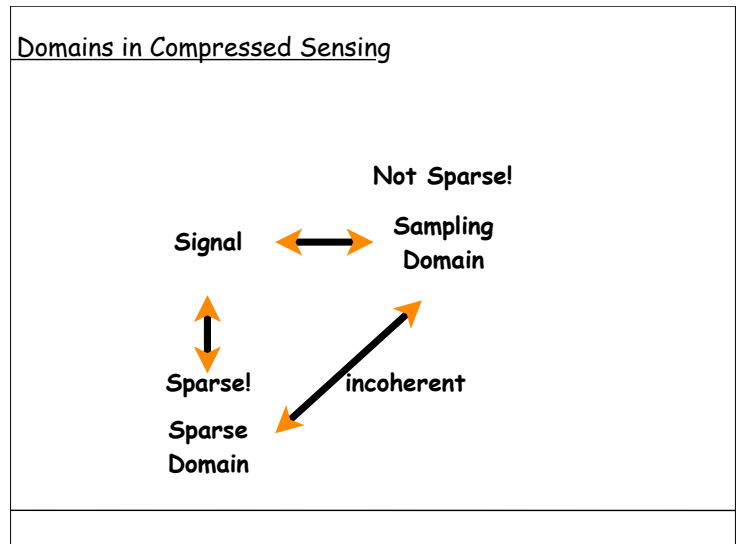


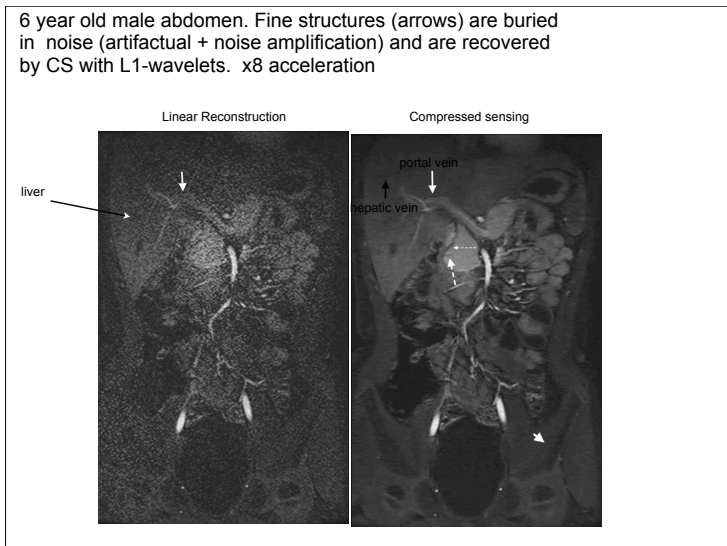
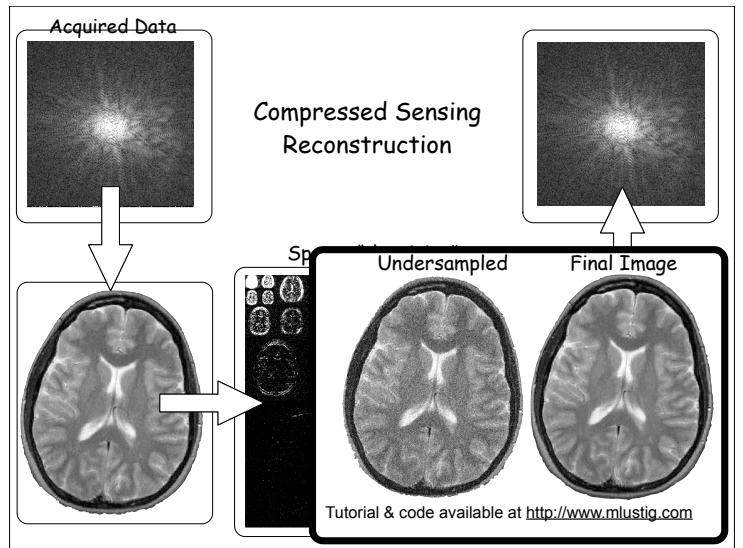
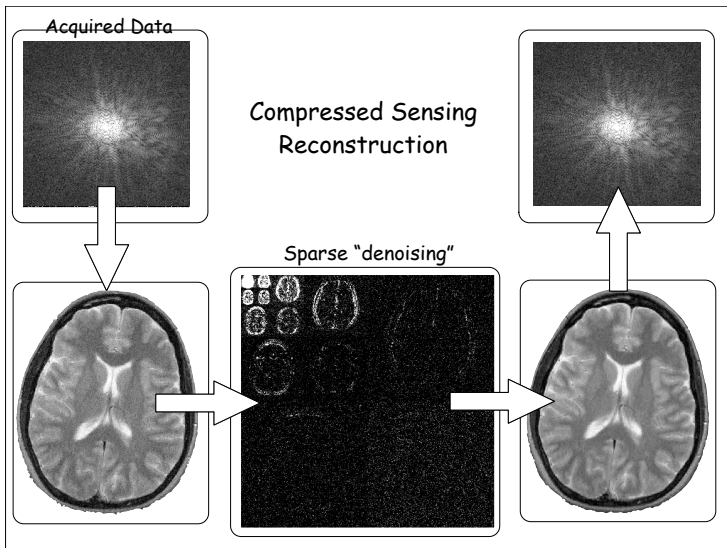
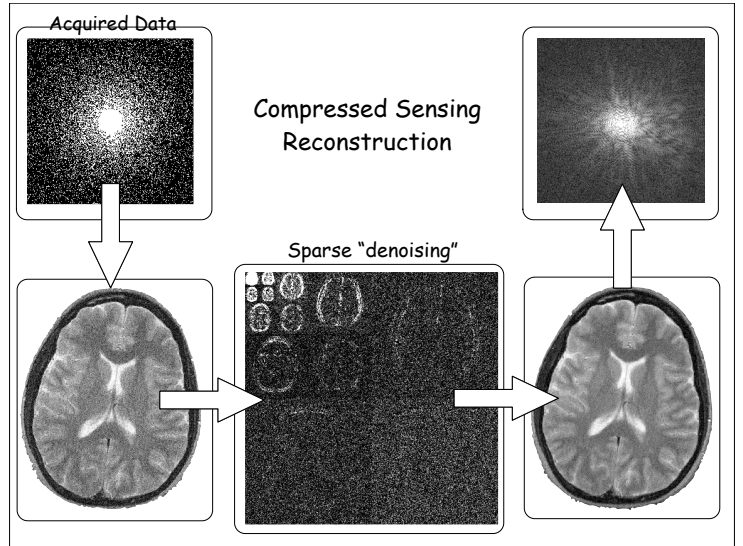
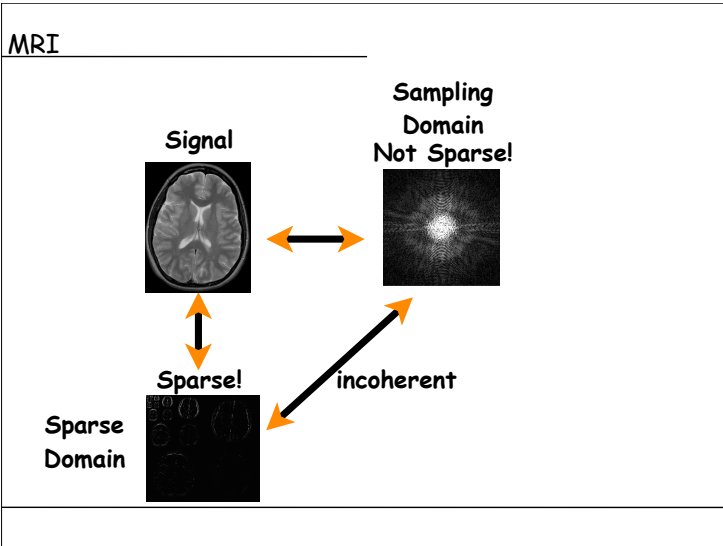
Question!

- What if this was the signal?
- Would CS still work?

random

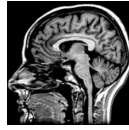
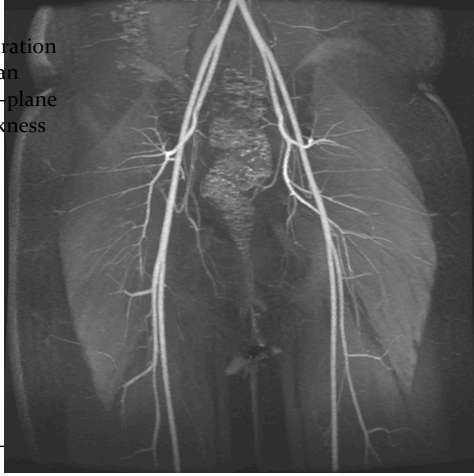
sub-Nyquist



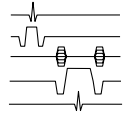


Back to Results

6 year old
8-fold acceleration
16 second scan
0.875 mm in-plane
1.6 slice thickness

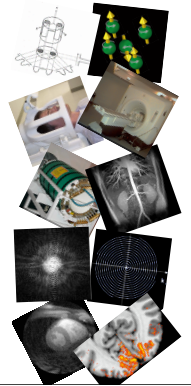


Principles of
Magnetic Resonance Imaging
EE c225E / BIOE c265



Spring 2016

Shameless Promotion



Other Applications

- Compressive Imaging
- Medical Imaging
- Analog to information conversion
- Biosensing
- Geophysical Data Analysis
- Compressive Radar
- Astronomy
- Communications
- More

Resources

- CS + parallel imaging matlab code, examples
<http://www.eecs.berkeley.edu/~mlustig/software/>
- Rice University CS page: papers, tutorials, codes,
<http://www.dsp.ece.rice.edu/cs/>
- IEEE Signal Processing Magazine, special issue on compressive sampling 2008;25(2)
- March 2010 Issue Wired Magazine: "Filling the Blanks"
- Igor Caron Blog: <http://nuit-blanche.blogspot.com/>

Thank you!
תודה רבה