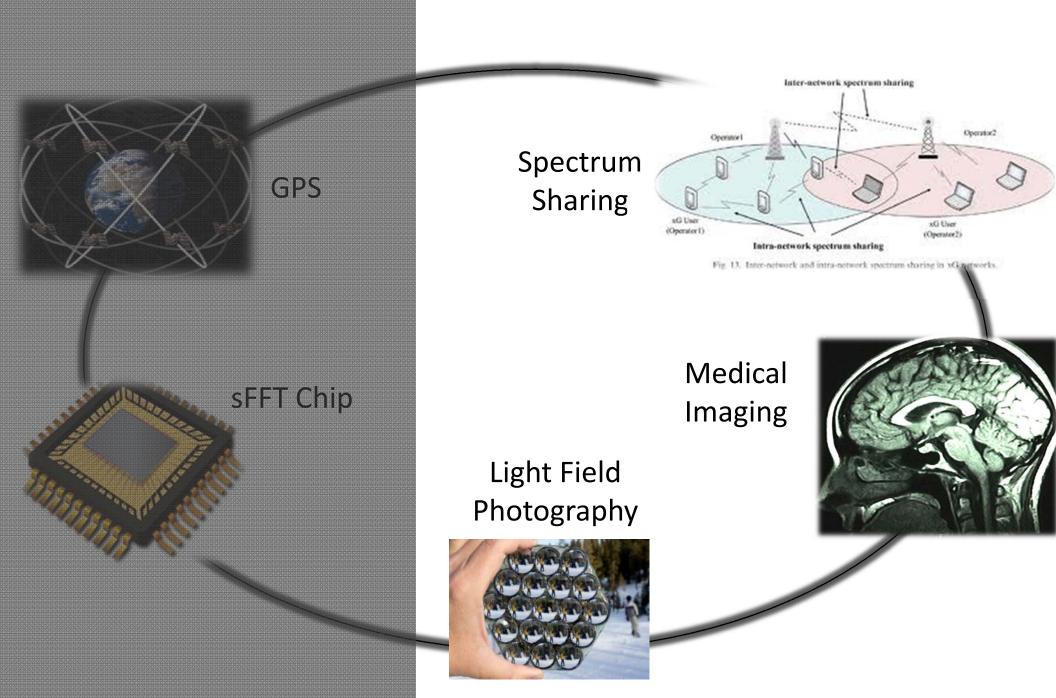
## The Sparse FFT: From Theory to Practice

### Dina Katabi

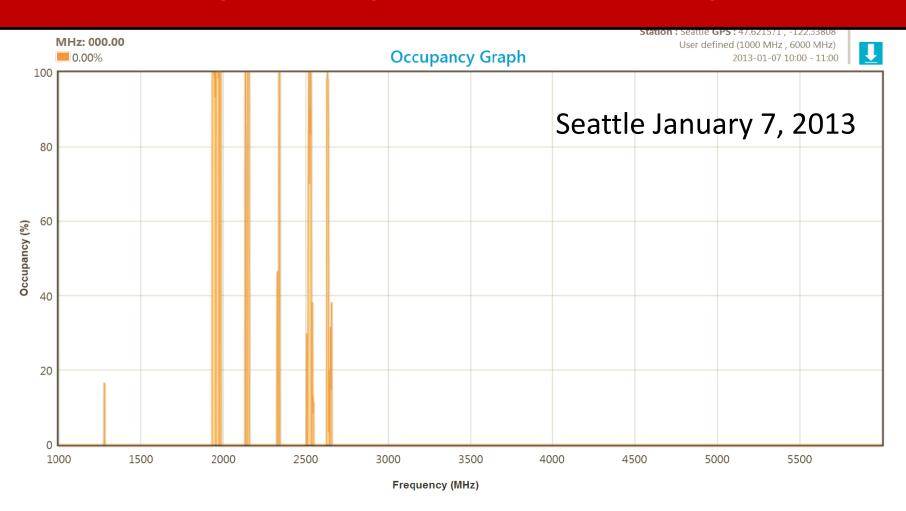
O. Abari, E. Adalsteinsson, A. Adam, F. adib, A. Agarwal, O. C. Andronesi, Arvind, A. Chandrakasan, F. Durand, E. Hamed, H. Hassanieh, P. Indyk, B. Ghazi, E. Price, L. Shi, V. Stojanovik



### Ongoing sFFT Projects (Beyond Theory)



# Spectrum Sharing Sense to find unused bands; Use them! How do you capture GHz of spectrum?



### Challenges in Sparse GHz Acquisition

GHz sampling is expensive and high-power



Tens of MHz ADC

< a dollar Low-power



A Few GHz ADC

Hundreds of dollars 10x more power

 Compressive sensing using GHz analog mixing is expensive, and requires heavy computation

### Recap of sFFT

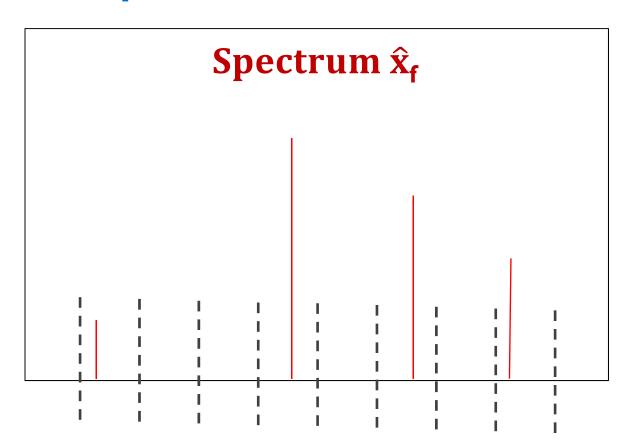
### 1- Bucketize

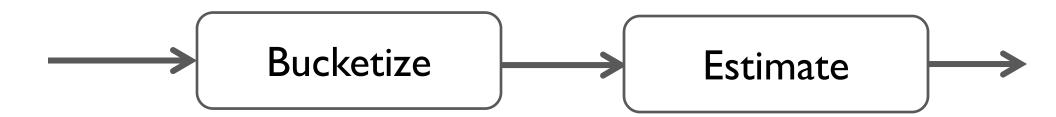
Hash the spectrum into a few buckets

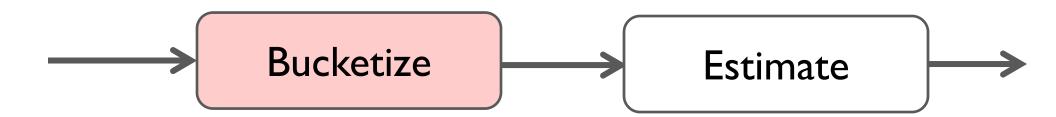
Can ignore empty bucket

### 2- Estimate

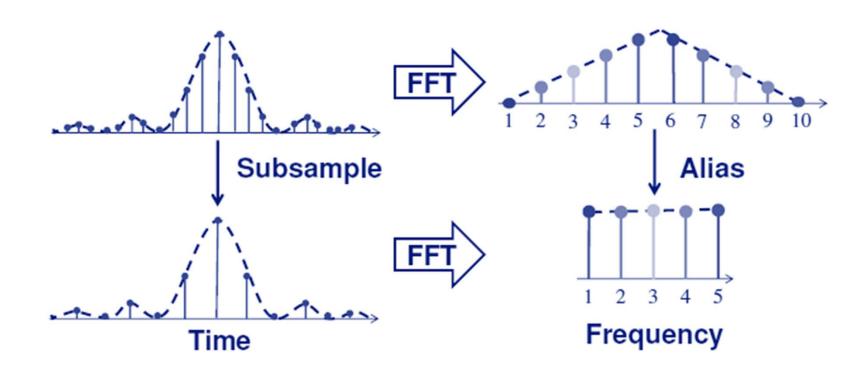
Estimate the large coefficient in each non-empty bucket

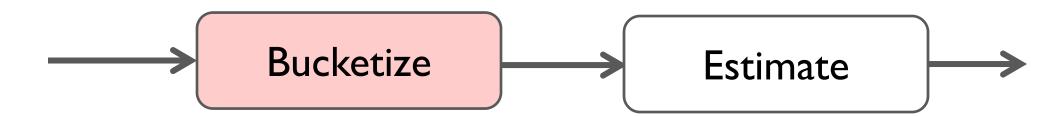






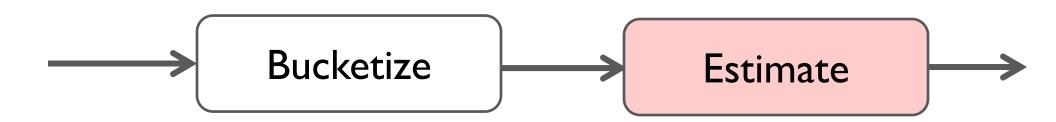
Sub-sampling time - Aliasing the frequencies





- Hash freqs. using multiple co-prime aliasing filters
  - Same frequencies don't collide in two filters
- Identify isolated freq. in one filter and subtract them from the other; and iterate ...

Low-speed ADCs, which are cheap and low-power



Estimate frequency by repeating the bucketization with a time shift  $\Delta T$ 

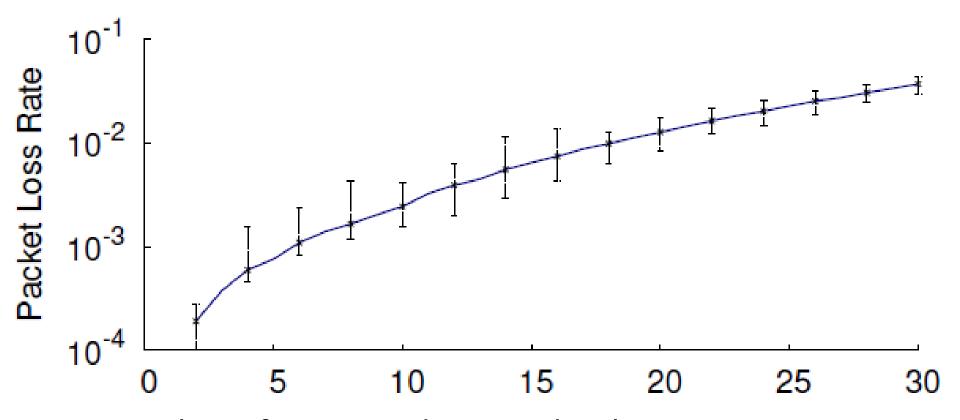
$$\Delta Phase = 2\pi f \Delta T$$

### Low-Power GHz Receiver

 Built a 0.9 GHz receiver using three 50 MHz software radios

 First off-the-shelf receiver that captures a sparse signal larger than its own digital bandwidth

### Concurrent Senders Hopping in 0.9 GHz

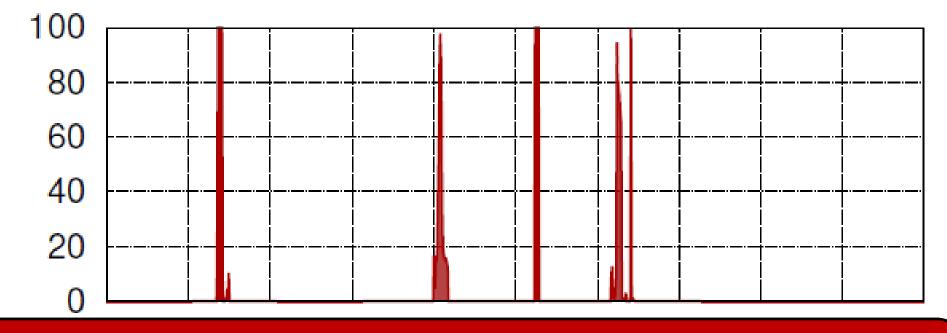


Number of MHz Senders Randomly Hopping gin in 0.9 GHz

### Realtime GHz Spectrum Sensing

Cambridge, MA January 2013

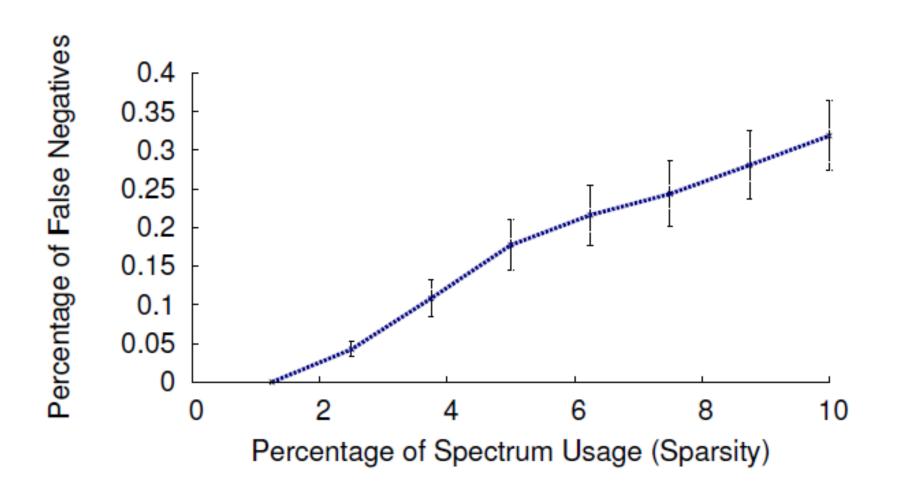
Occupancy from 2GHz to 3GHz (10 ms FFT window)



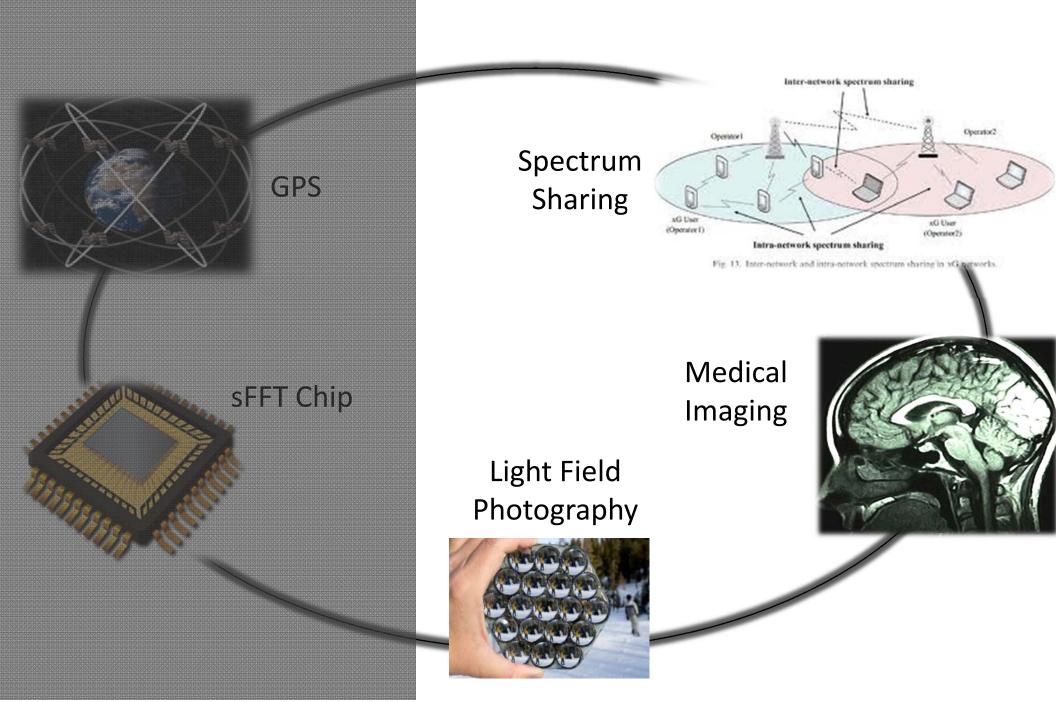
sFFT enables a GHz low-power receiver using only a few MHz ADCs

Occupancy %

### Probability of Declaring a Used Frequency as Unused



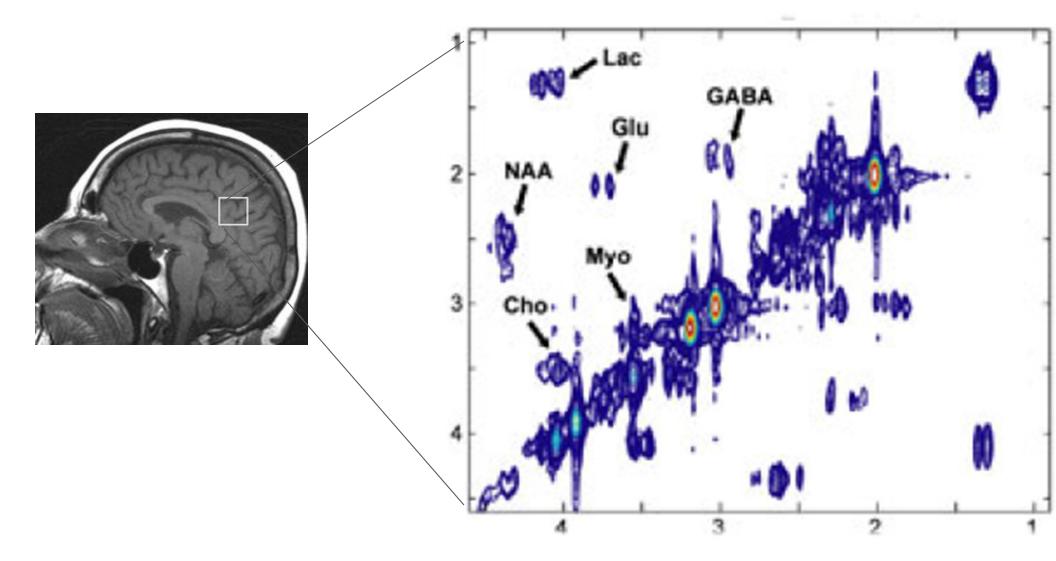
### Ongoing sFFT Projects (Beyond Theory)



### Magnetic Resonance Spectroscopy

Analyses the chemical making of a brain voxel

→ Disease Bio-markers



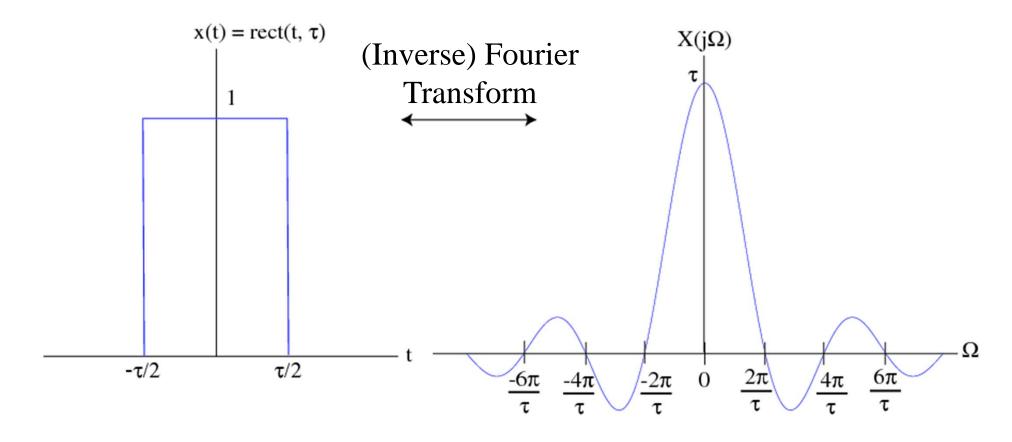
### Challenges

- Long acquisition time
  - -patient is in the machine for 40min to hours

Artifacts due to acquisition window

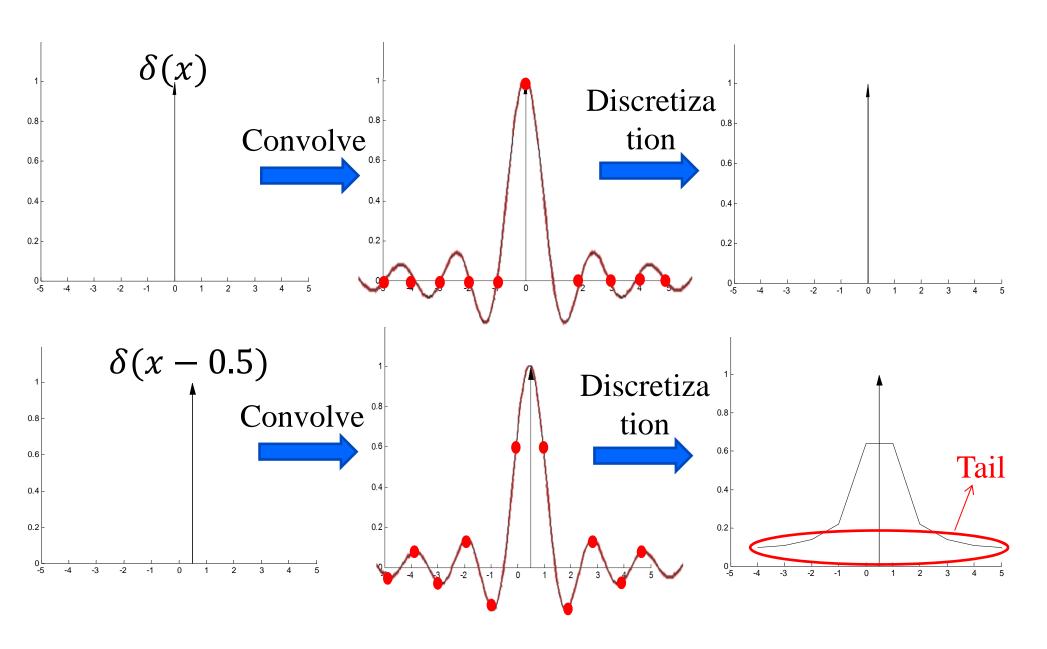
### Windowing Artifacts

Fourier transform of a window is a sinc



Acquisition Window - Convolution with a since

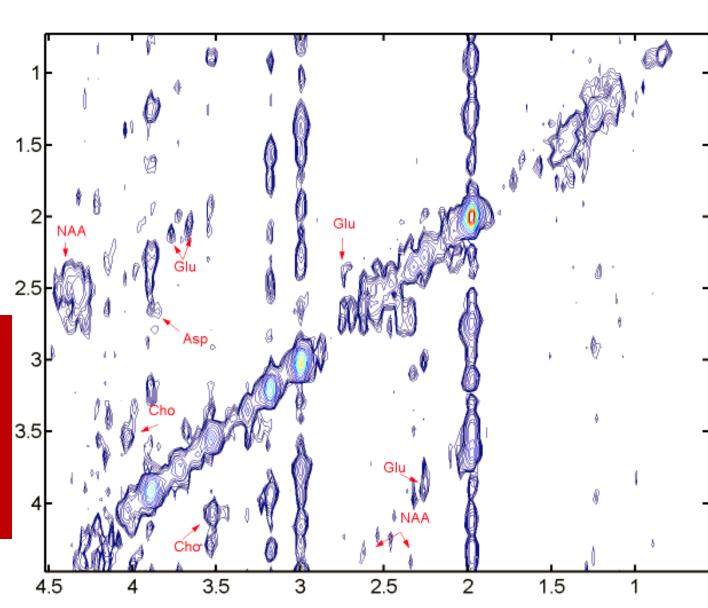
### Windowing Artifacts



### Challenges with In-Vivo Brain MRS

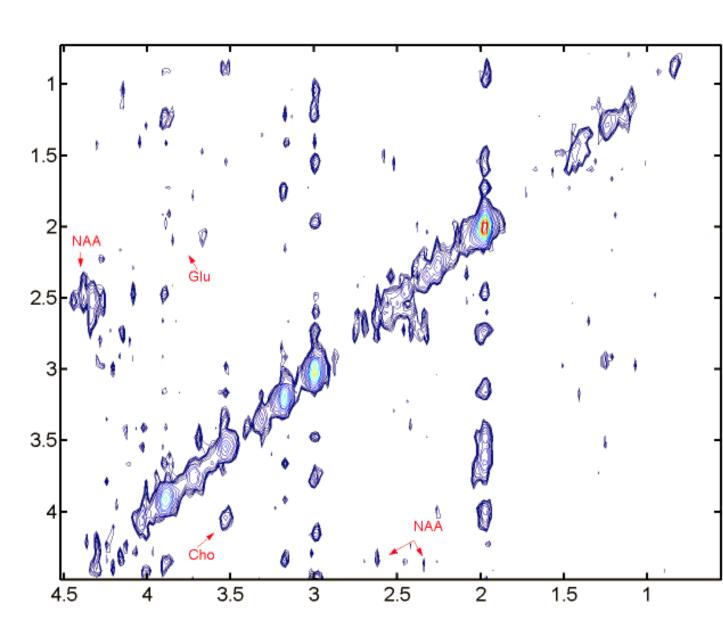
- I) clutter due to sinc tail
- 2) hours in machine

Can sparse recovery help?



### Compressive Sensing + 30% data

Lost some Biomarkers



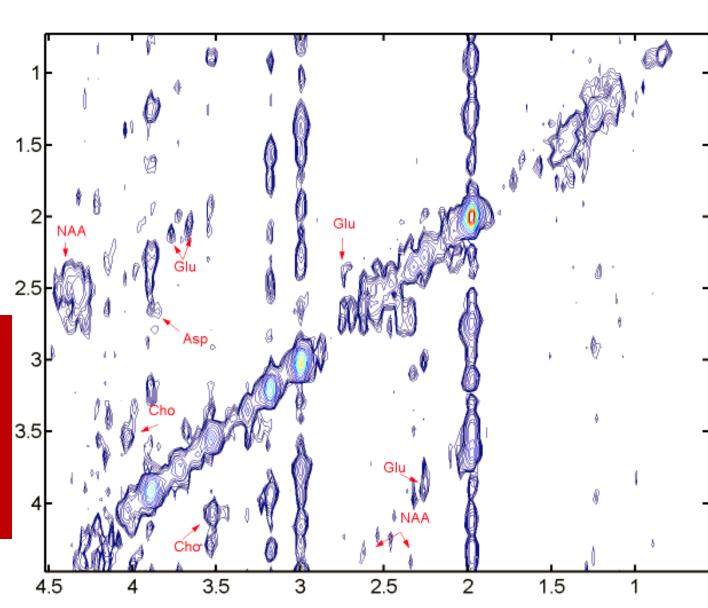
### Non-Integer Sparse FFT

- Problem and Model
  - Sparse in the continuous case
  - The railings are because of non-integer frequencies
- Algorithm
  - Use original sparse FFT to estimate integer frequencies
  - Use gradient descent algorithm to find the noninteger frequencies to minimize the residue of our estimation over the samples

### Challenges with In-Vivo Brain MRS

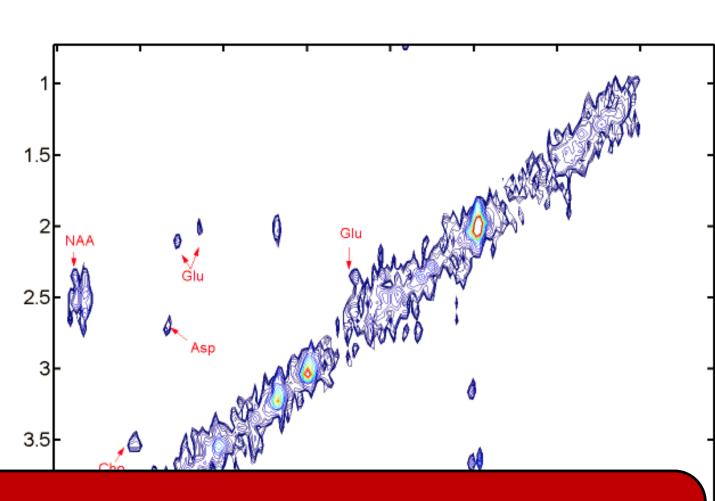
- I) clutter due to sinc tail
- 2) hours in machine

Can sparse recovery help?



### Sparse FFT + 30% of data

Removed
Clutter
without losing
Biomarkers

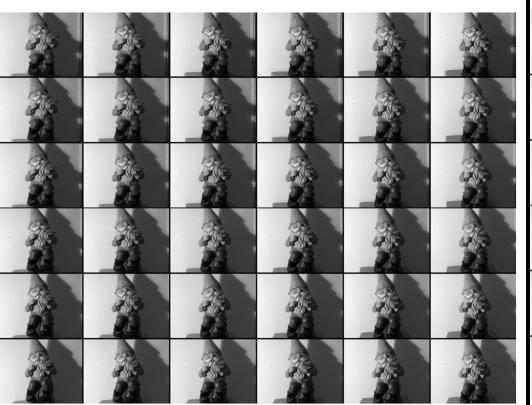


sFFT provides clearer images while reducing the acquisition time by 3x

### Light-Field Photography

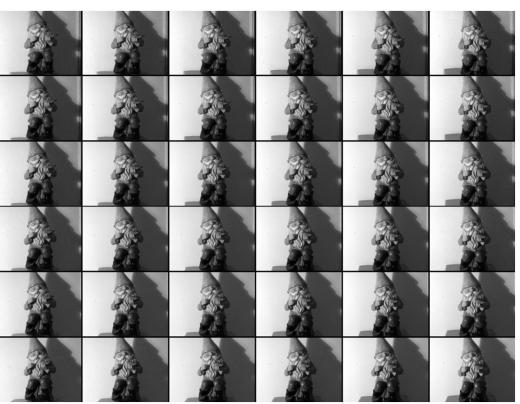
- Generate depth and perspective using images from a 2D camera array
- Images are correlated  $\rightarrow$ 4D frequencies are sparse
- Goal: Same performance but with fewer images





Original





Reconstructed with 11% of data



### Conclusion

- Many applications are sparse in the frequency domain and hence can benefit from sFFT
- We showed that sFFT enables GHz low-power spectrum sensing and decoding, and improves MRS medical imaging and 4D light-filed capture
- We just scratched the surface and expect more applications soon