



**Alberta Health
Services**

Potential for Compressed Sensing in Clinical MR Imaging



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CALGARY

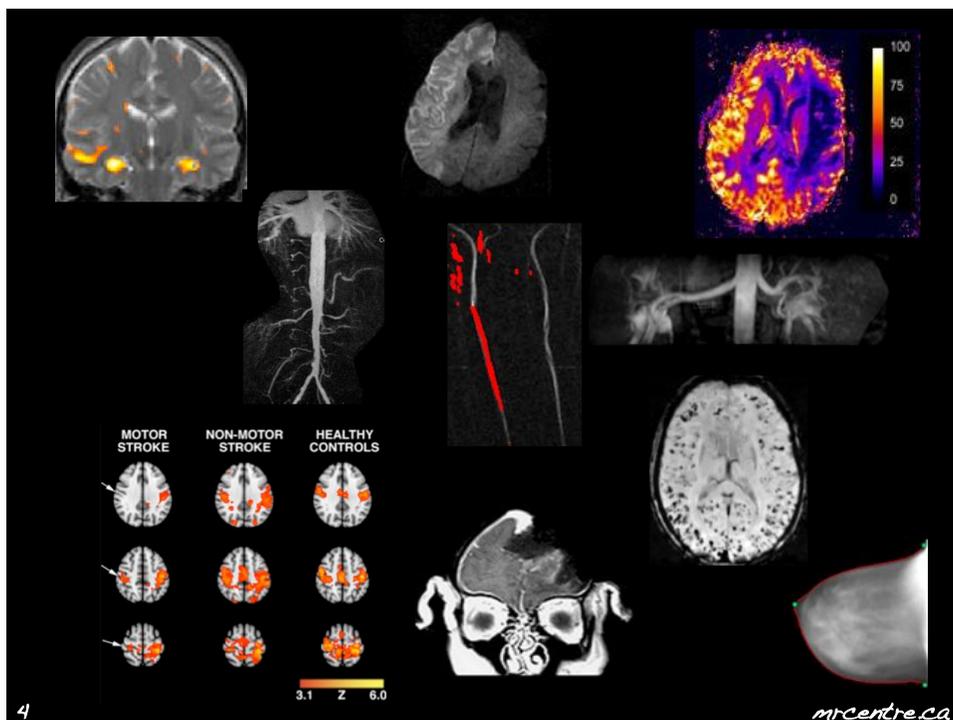
CFTCKI
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Outline

- MR Imaging – Acquisition and Reconstruction
 - resolution
 - signal-to-noise
 - total acquisition time
- Image Compression
 - compressed sensing
 - maximising trade-off the between resolution, SNR and total acquisition time
- Clinical applications
 - scanning faster with higher resolution and better SNR

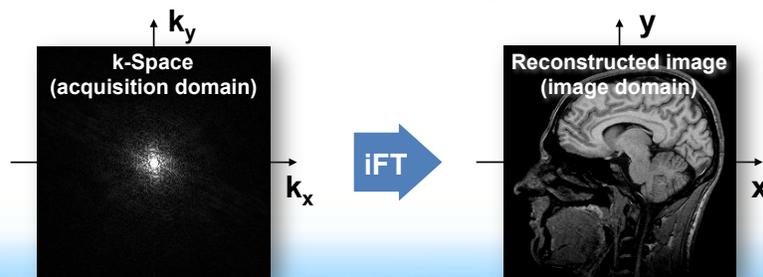
Magnetic Resonance (MR) Imaging

- Non-invasive clinical imaging modality
 - visualization of anatomic structure
 - investigation of body functions
- Detection or staging of numerous diseases
 - cancer, cardiac disease, stroke, etc.
- Sensitivity to detect pathologies (disease) relies heavily on quality of acquired images, e.g.,
 - spatial and temporal resolution, signal-to-noise ratio
 - total acquisition time
 - image contrast and artifacts, ...



Conventional MR Imaging

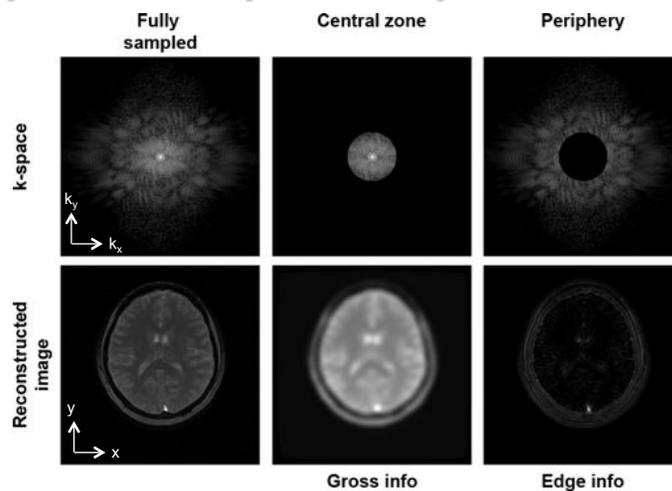
- Data acquisition requires complete and uniform sampling of k-space
- Reconstruction by inverse Fourier transform (iFT)
- Acquisition of complete k-space is slow
 - can limit spatial and/or temporal resolution



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Important k-Space Properties

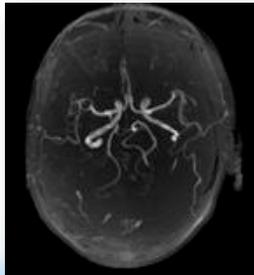


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The Tradeoff in MR Angiography

- Critical tradeoff between spatial resolution, acquisition time and SNR
- Typical coverage: 20 cm × 20 cm × 10 cm



Acq matrix = ~~256 × 256 × 150~~ 100
 Resolution = ~~784 μm × 784 μm × 2 mm~~
 Scan time = ~~4.694 min~~
 SNR = 100%
 SNR_{eff} = 100%

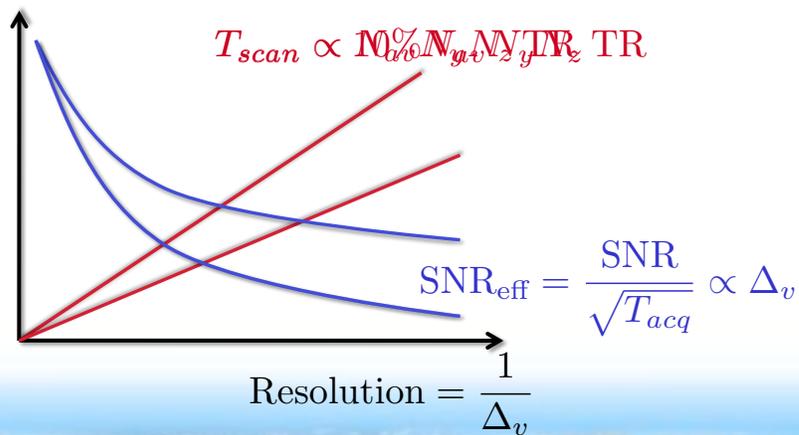
$$\text{SNR} \propto \Delta_v \sqrt{T_{acq}}$$

$$\text{SNR}_{\text{eff}} = \frac{\text{SNR}}{\sqrt{T_{acq}}} \propto \Delta_v$$

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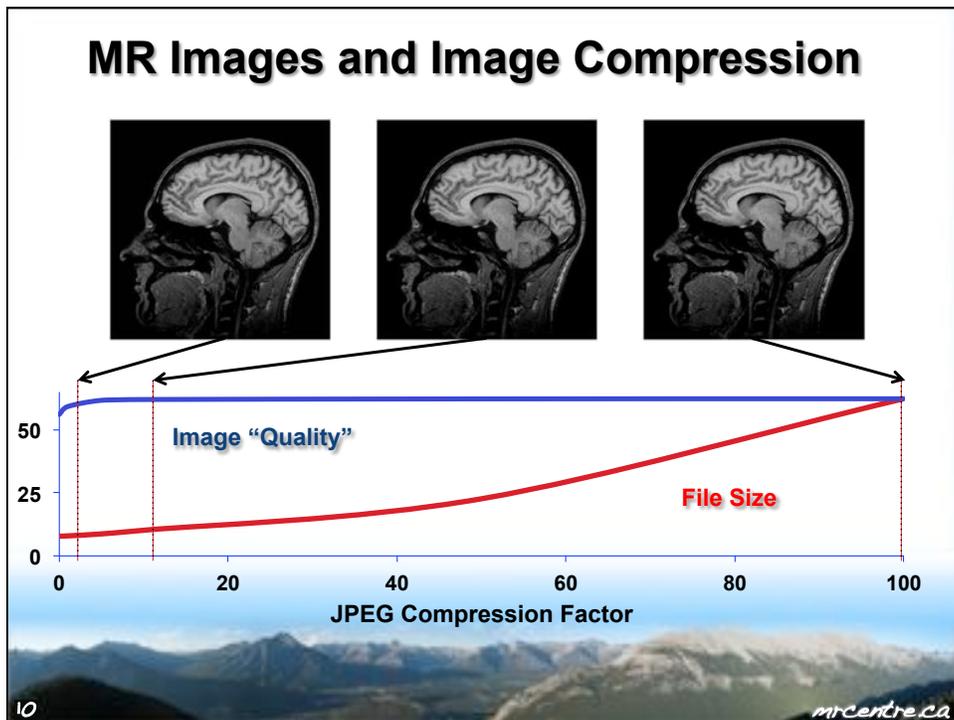
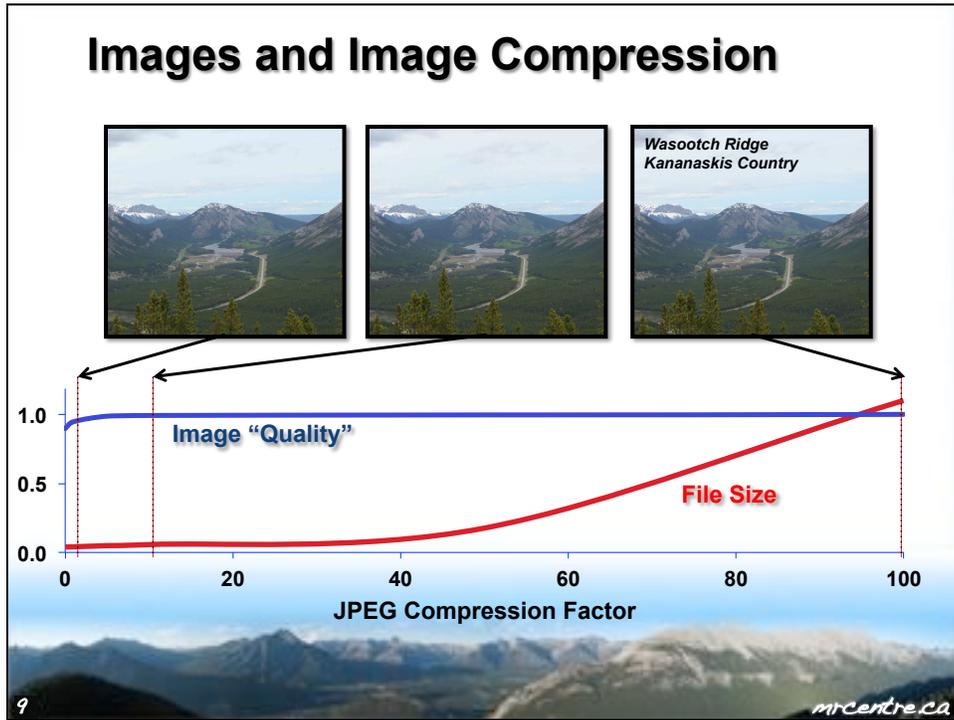
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Scan Time and SNR Efficiency



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Reprinted with corrections from *The Bell System Technical Journal*, Vol. 27, pp. 379-423, 623-656, July, October, 1948.

A Mathematical Theory of Communication

By C. E. SHANNON

What is the minimum amount of data that need to sent over the communication channel that allows for accurate reconstruction of the original signal?

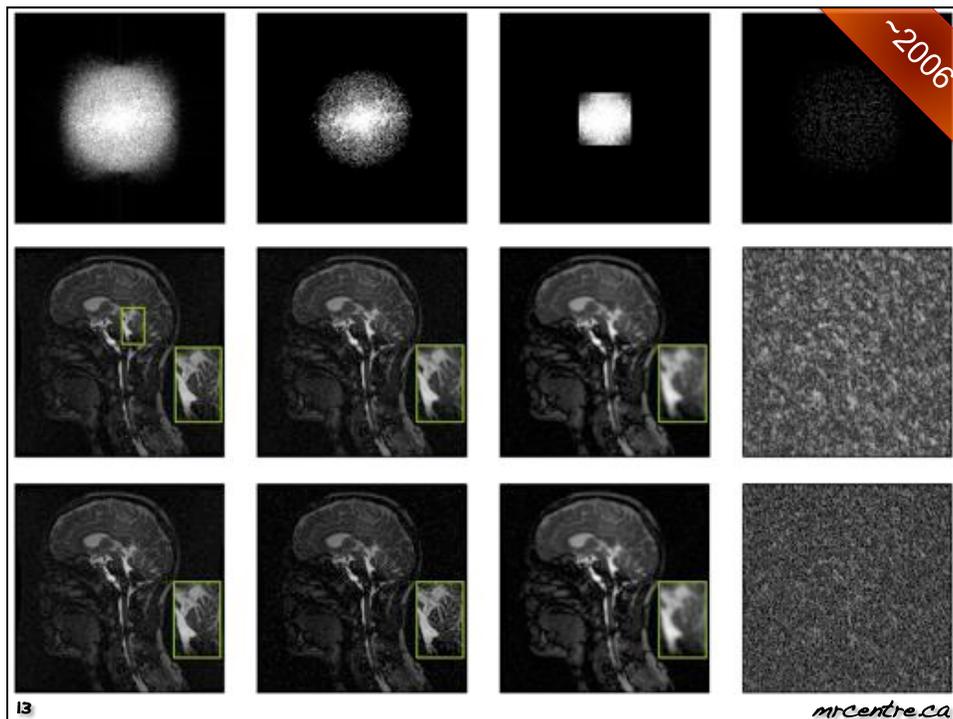
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Reprinted with corrections from *The Bell System Technical Journal*, Vol. 27, pp. 379-423, 623-656, July, October, 1948.

Data Acquisition **Image Reconstruction**

What is the minimum amount of data that need to sent over the communication channel that allows for accurate reconstruction of the original object?

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Compressed Sampling (CS)

- Mathematical framework for understanding reconstruction from partial samples
 - Candes and Tao, 2006
 - Donoho, 2006
- Nyquist-Shannon sampling theory states that
 - number of Fourier samples = number of pixels
- Compressed sampling theory suggests that it is possible to reconstruct accurate images from fewer samples

CS Requirements

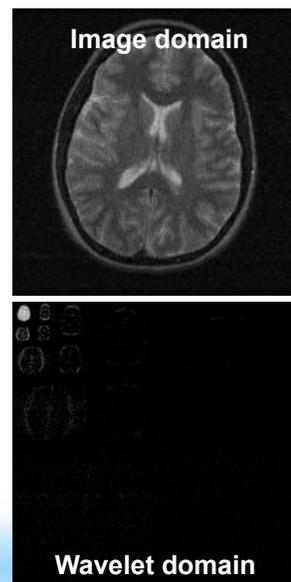
1. Transform sparsity
2. Incoherence of aliasing interference due to undersampling
3. Non-linear iterative reconstruction algorithm

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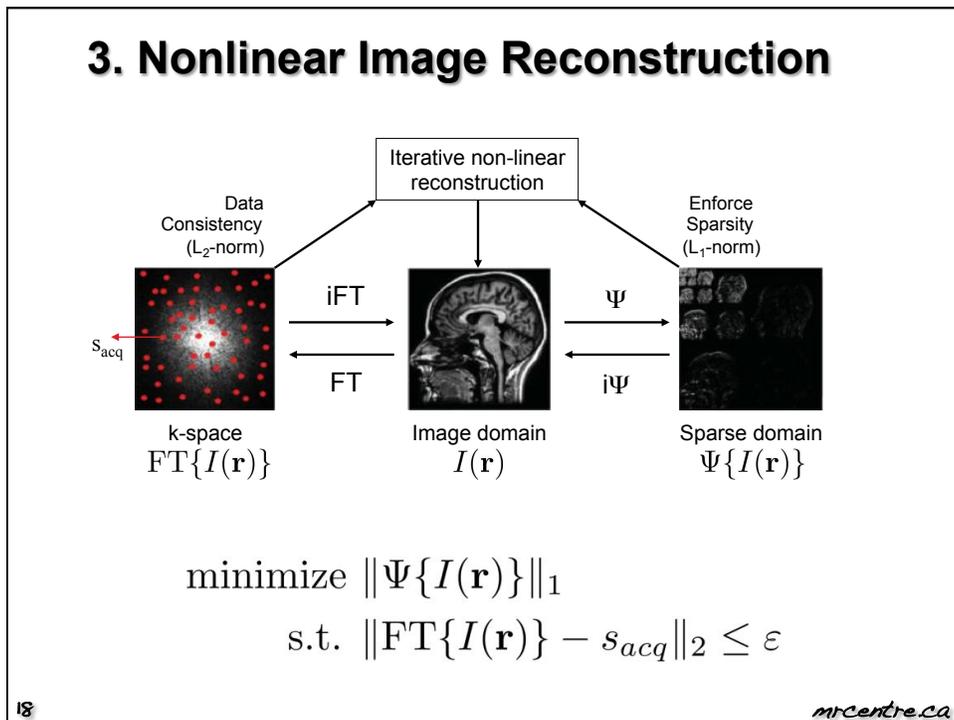
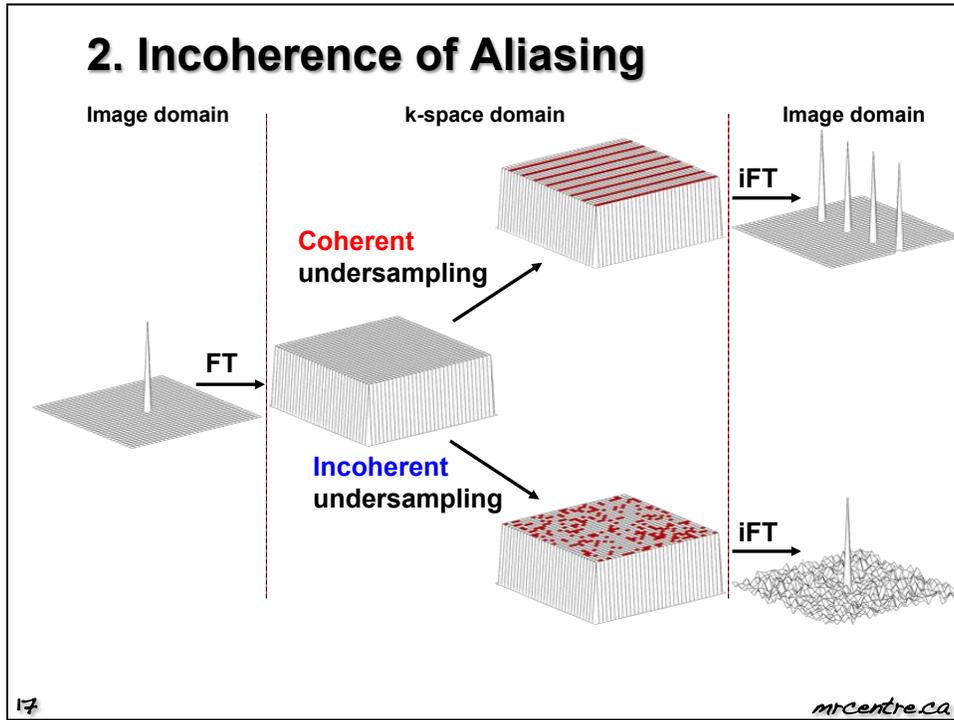
1. Transform Sparsity

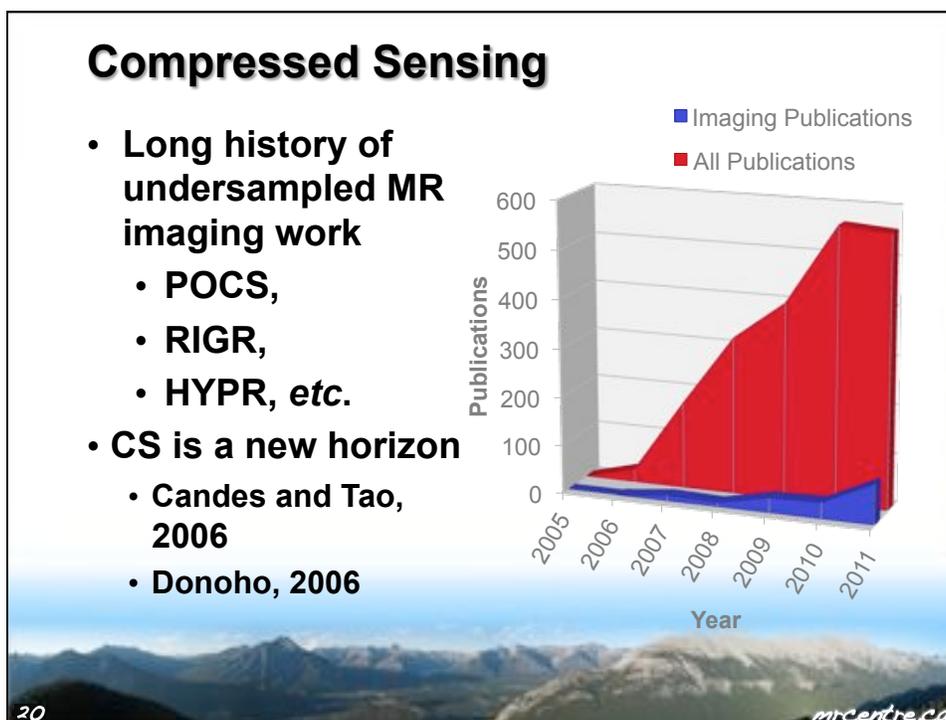
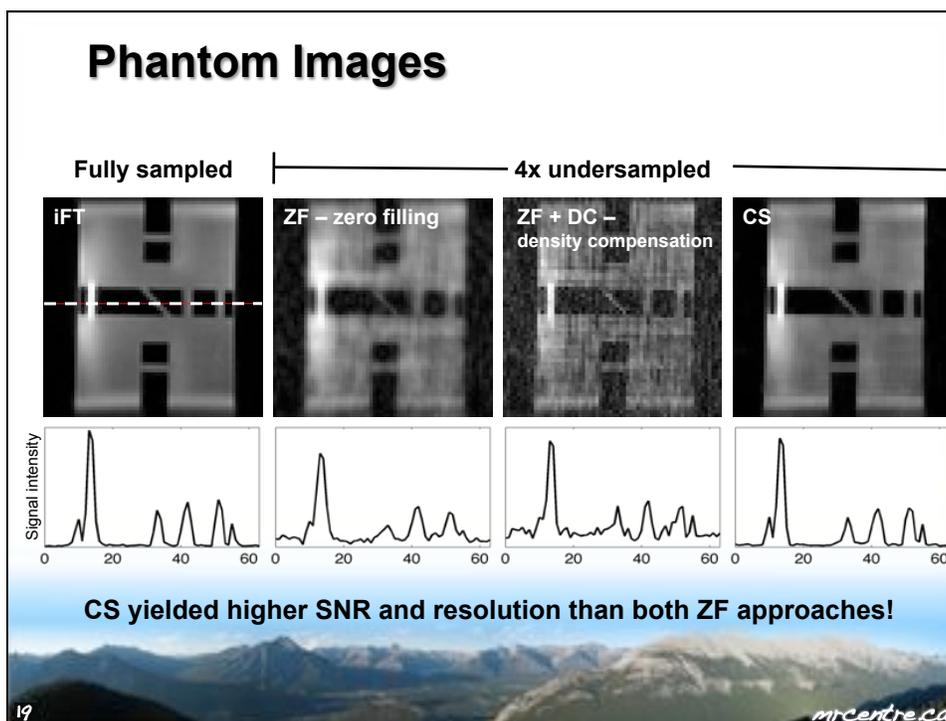
- Sparse representation
 - few large coefficients, many small coefficients
- Sparsity via a transform domain
 - discrete cosine transform (DCT),
 - wavelet transform,
 - finite differences, etc.



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“Fast” MR Imaging

- Imaging dynamic phenomena
 - cardiac imaging
 - functional imaging
- Shortening length of imaging protocol
- Increasing image resolution
- Patient comfort
- Minimizing patient motion artifacts
- Surgical intervention
 - MR-guided catheter-based therapy

Compressed Sensing may open up new opportunities for “fast” MR

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Design Goals for Undersampled MR

Accurately reconstructed the image

- Any errors must be obvious
- Image contrast is preserved
- Artifacts (e.g., due to motion) must be well understood

Secondary

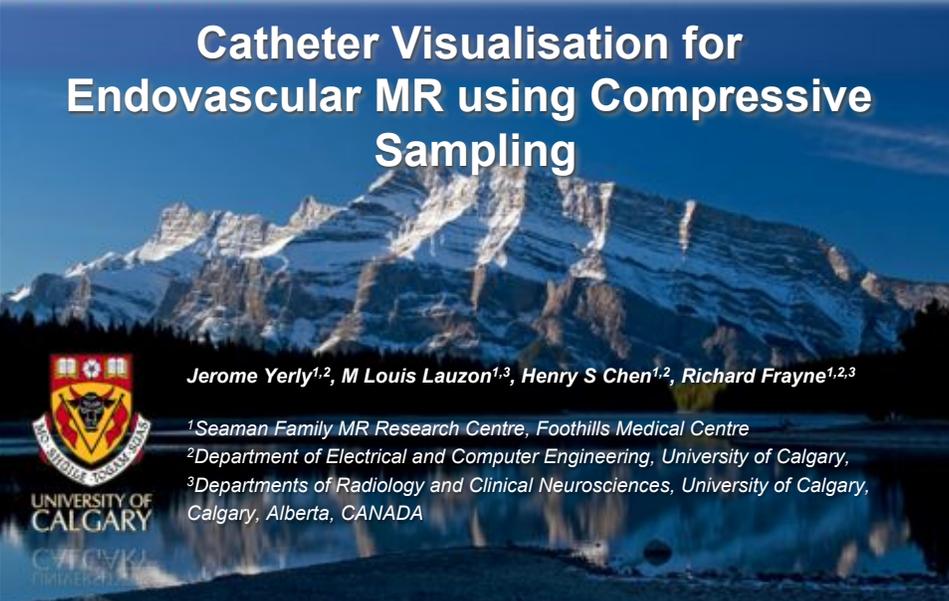
- Reconstruction time must be short

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Catheter Visualisation for Endovascular MR using Compressive Sampling



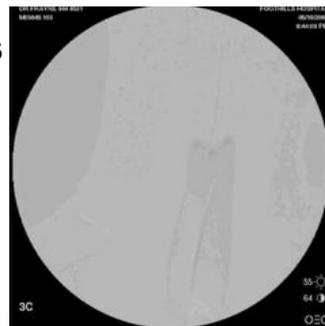
Jerome Yerly^{1,2}, M Louis Lauzon^{1,3}, Henry S Chen^{1,2}, Richard Frayne^{1,2,3}

*¹Seaman Family MR Research Centre, Foothills Medical Centre
²Department of Electrical and Computer Engineering, University of Calgary,
³Departments of Radiology and Clinical Neurosciences, University of Calgary,
 Calgary, Alberta, CANADA*

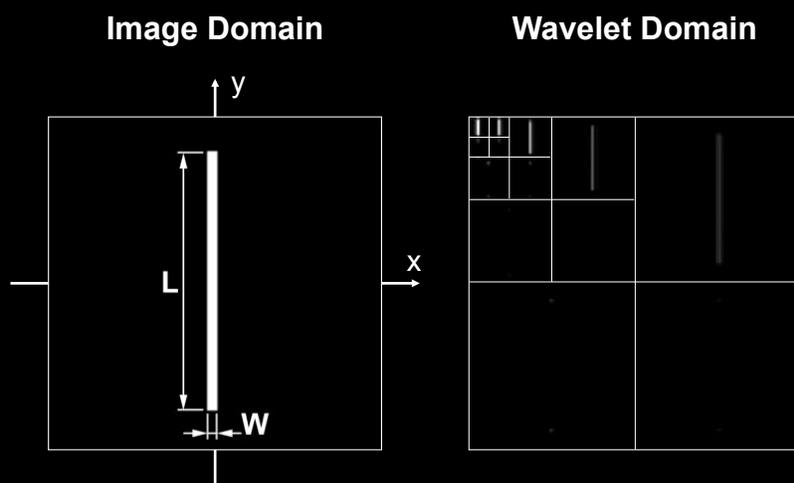


Endovascular Intervention

- Treatment of vascular diseases
- Minimally invasive
- Catheter-based procedures
- Device / drug deployment
- Catheter guiding
- Real-time imaging
 - X-ray (gold standard)
 - MR imaging
 - 10 Hz image rates



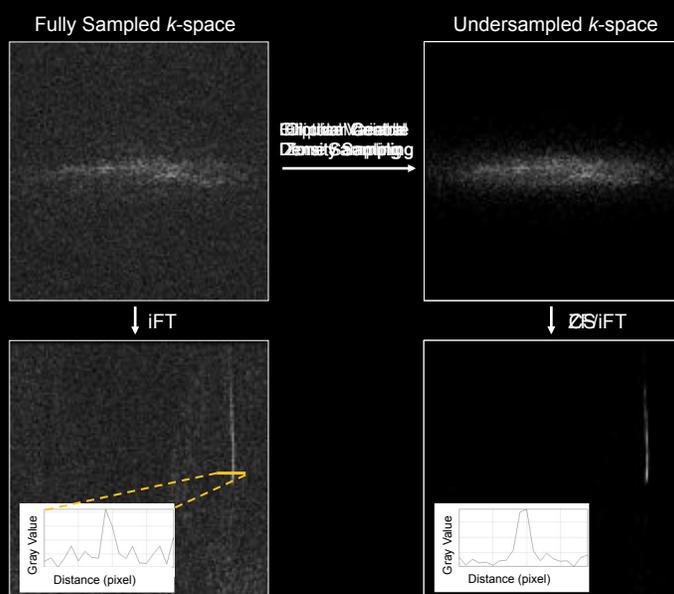
Sparsity in Transform Domain



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Zero Filling (ZF) versus CS

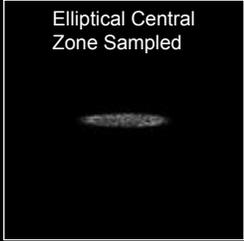
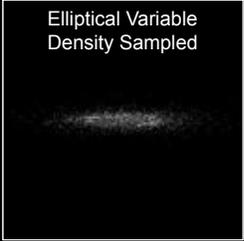
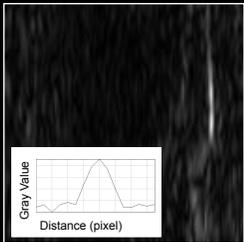
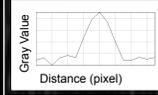
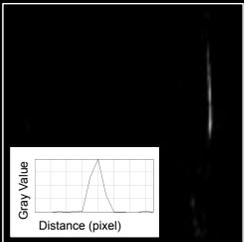


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Undersampling Factor

~~30x Undersampled~~

<p>Elliptical Central Zone Sampled</p> 	<p>Elliptical Variable Density Sampled</p> 
↓ ZF/IFT	↓ CS
 <p>Gray Value Distance (pixel)</p> 	 <p>Gray Value Distance (pixel)</p> 

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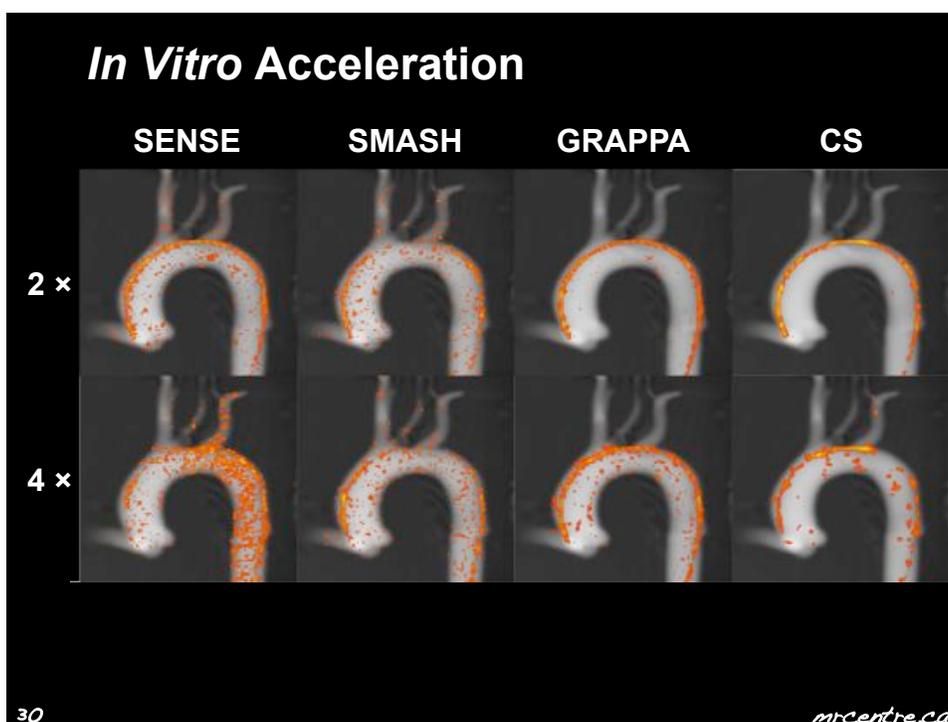
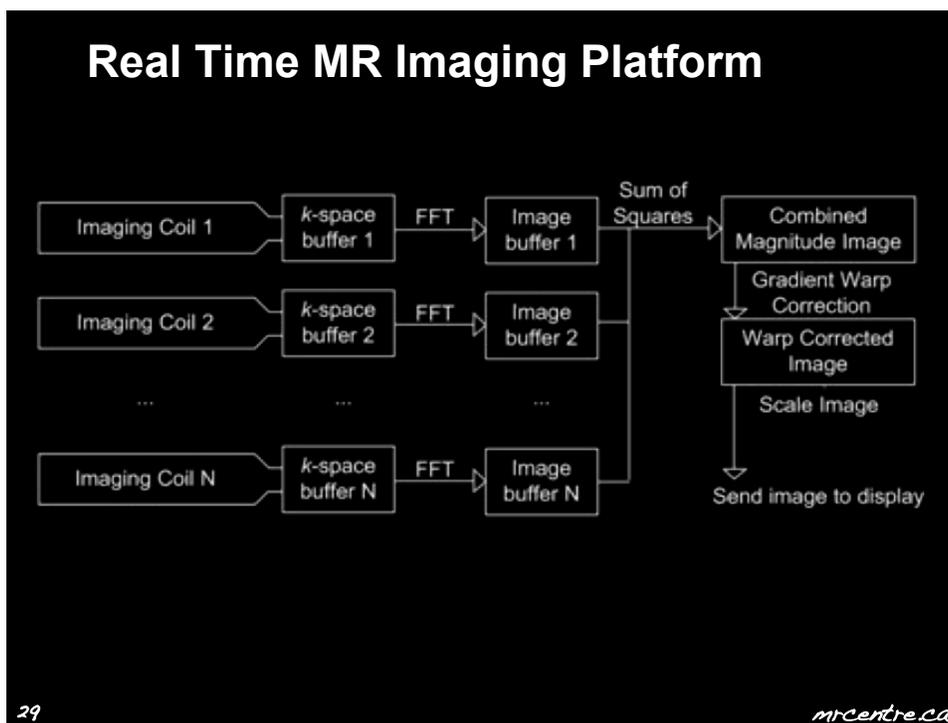
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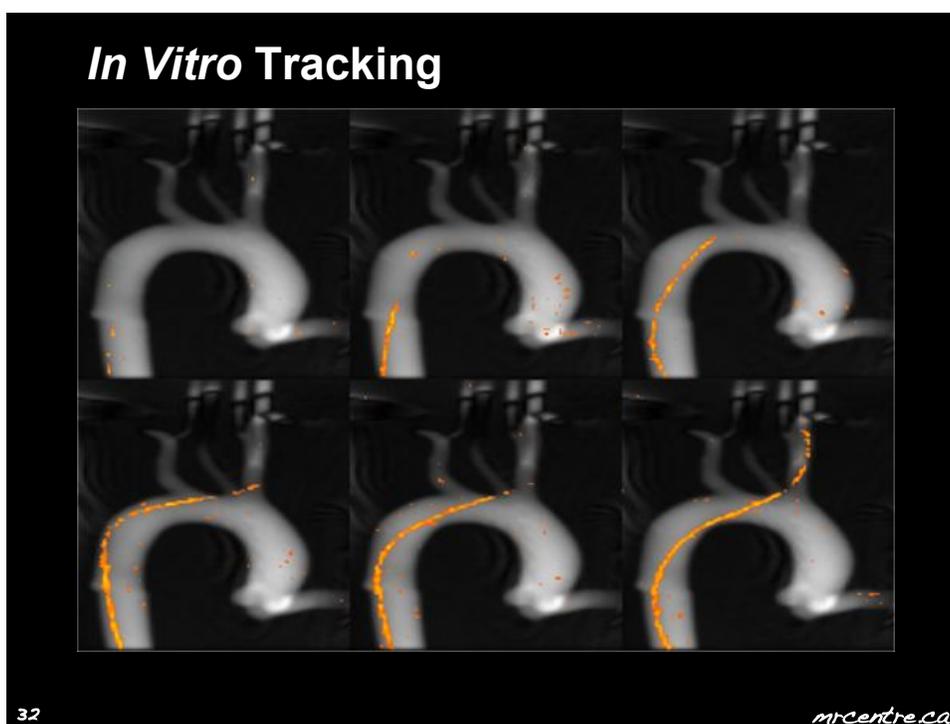
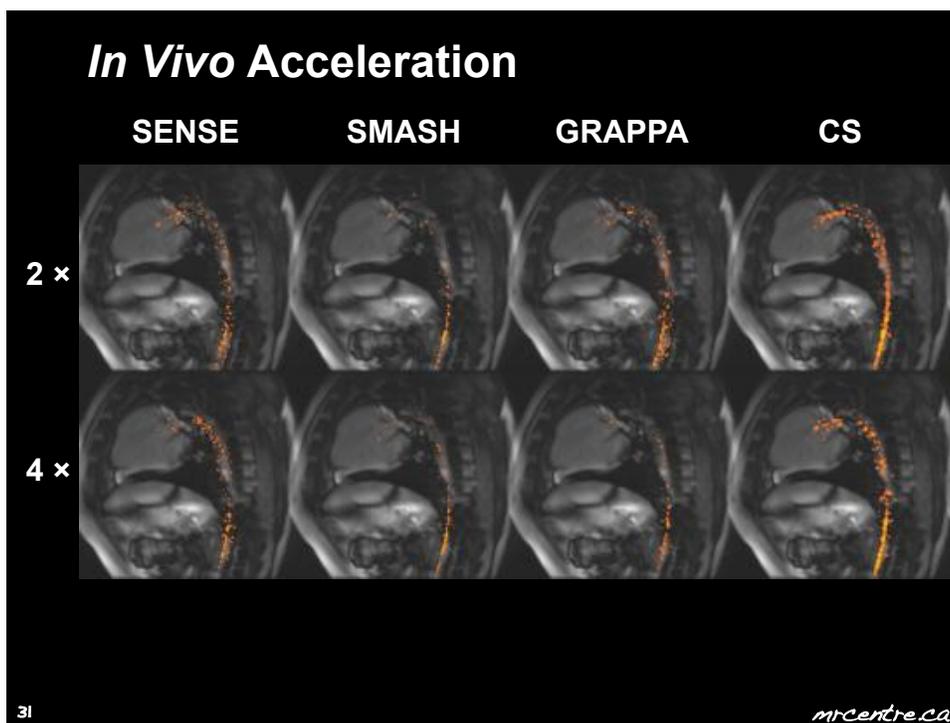
Passive Catheter Tracking into the Carotid Artery using Accelerated Magnetic Resonance Imaging



Matthew Ethan MacDonald
*Biomedical Engineering Graduate Program,
 Department of Electrical and Computer Engineering,
 University of Calgary,
 Seaman Family MR Research Centre, Foothills Medical Centre*



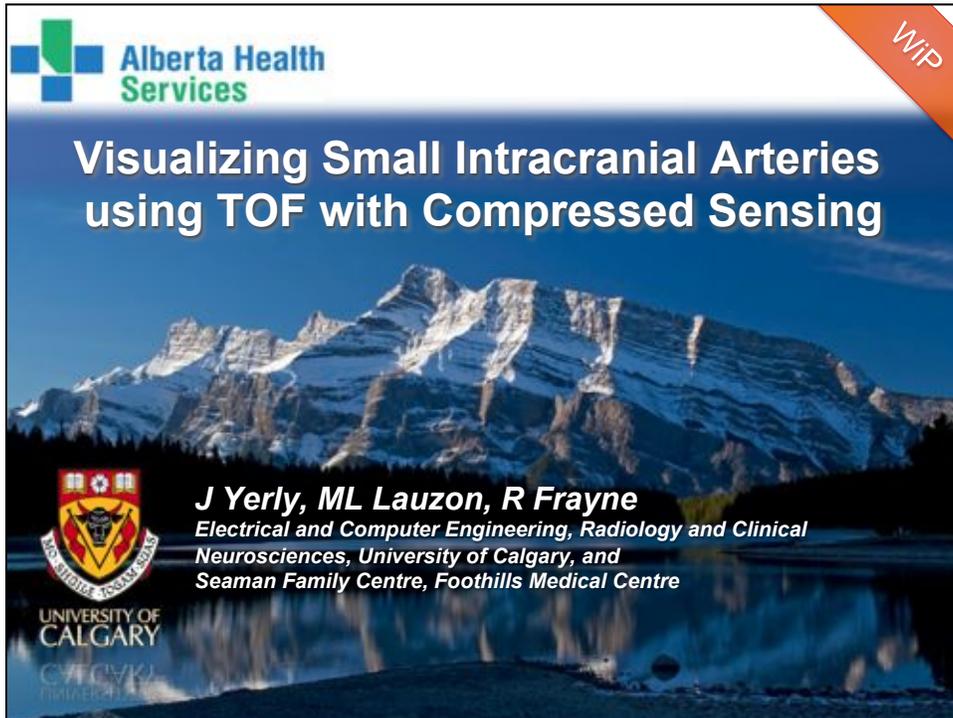




WIP

Alberta Health Services

Visualizing Small Intracranial Arteries using TOF with Compressed Sensing



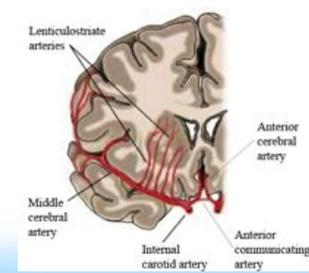


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Small Intracranial Arteries

- Lenticulostriate arteries
- Diameter: 100 μm to 900 μm
- Lacunar strokes
- Challenging to visualize with ≤ 3 T scanner



Purves *et al.* Neuroscience 2001; Figure 1.20

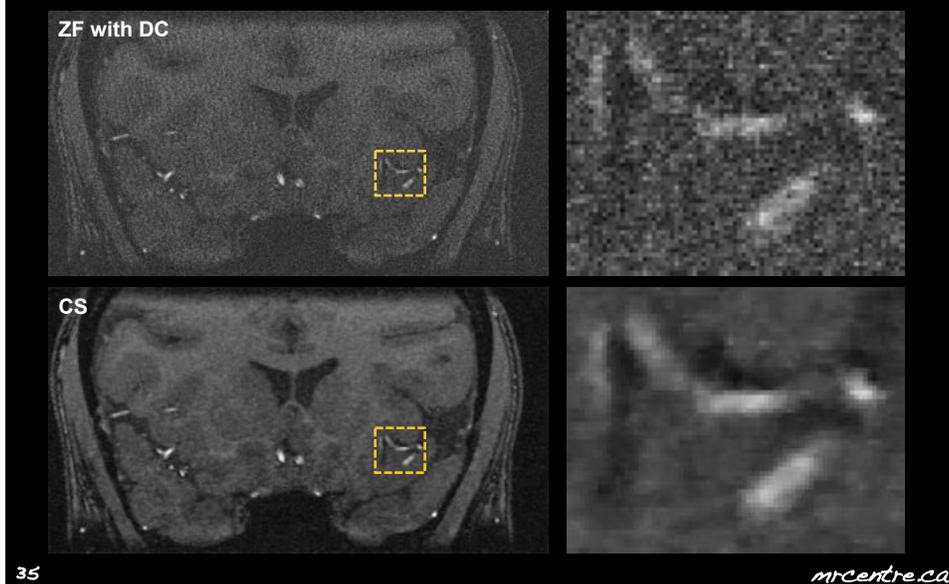


Kang *et al.* Int J Stroke 2010;5:374

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MR Angiography - Source Images

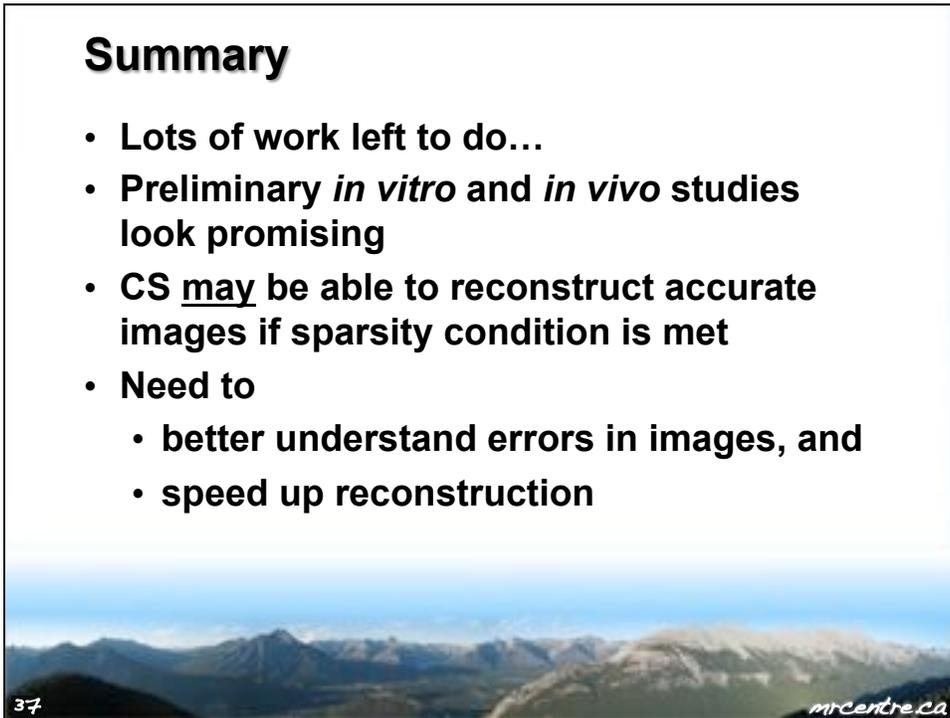


Maximum Intensity Projection Images



Summary

- Lots of work left to do...
- Preliminary *in vitro* and *in vivo* studies look promising
- CS may be able to reconstruct accurate images if sparsity condition is met
- Need to
 - better understand errors in images, and
 - speed up reconstruction



Acknowledgments



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Canada Research Chairs / **Chaires de recherche du Canada**

Hopewell

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