#### Accelerating Advanced MR Image Reconstruction using GPUs

#### Justin Haldar August 10, 2009



## Faculty

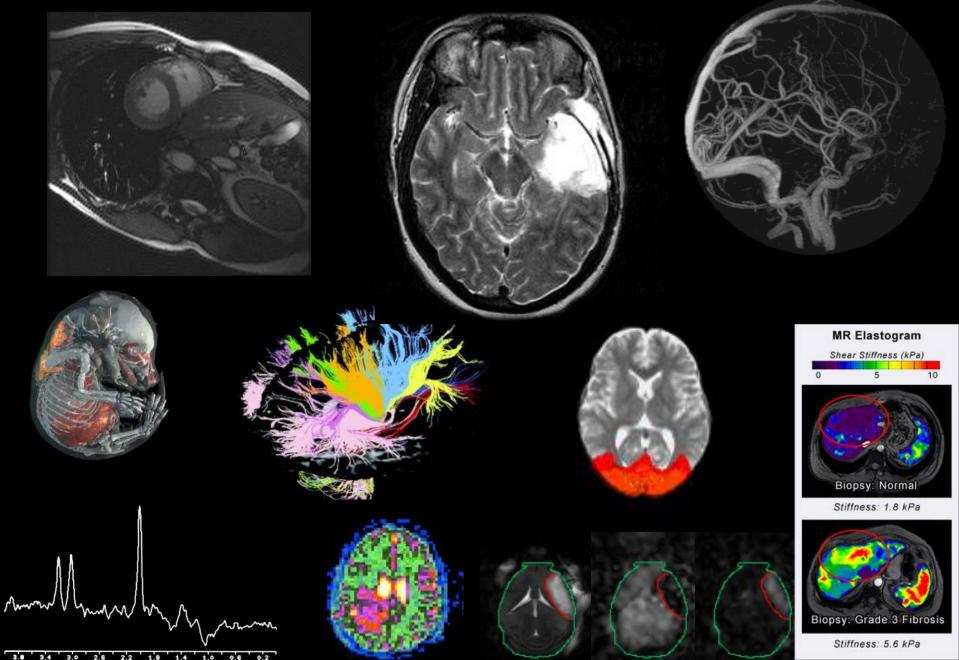
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People

#### Introduction to MRI

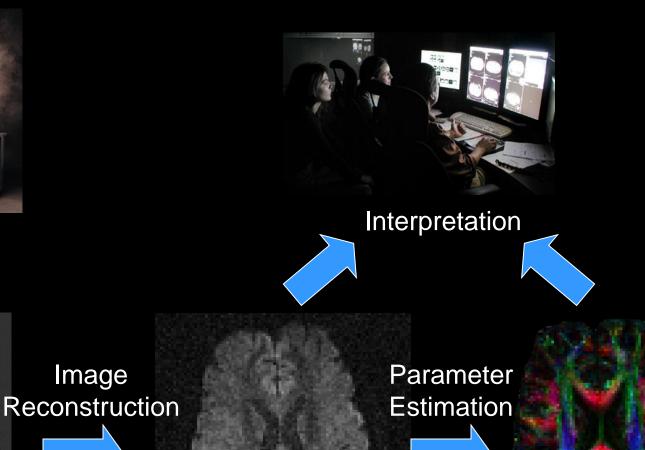


### MRI Pipeline

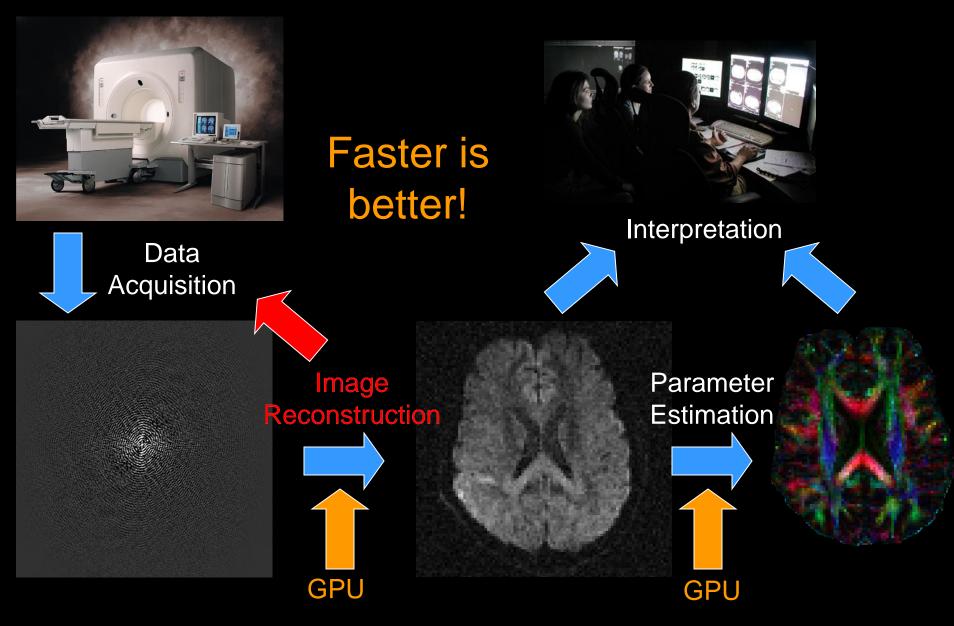
Image



Data Acquisition

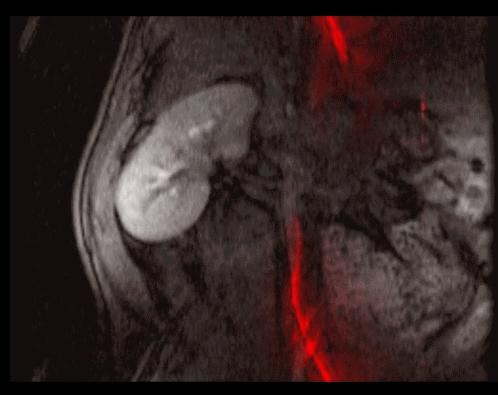


### MRI Pipeline



#### Example: Interventional MRI





Real-time reconstruction is necessary to provide feedback to surgeon

### MRI data in Fourier Space

 Ignoring several effects, MRI image and signal are a Fourier transform pair

true image

#### Fourier kernel



noise

#### data samples

#### **Discretization**

 Infinite dimensional variables are inconvenient for computation

Finite dimensional image representation

voxel basis function

**Integral equation** 

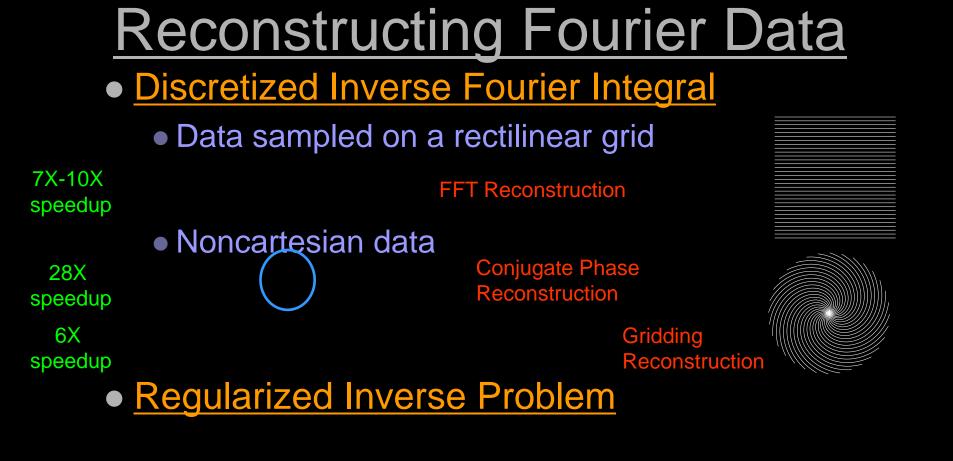
Matrix equation

## Very Large Matrix Equations

Typical 2D images: N of ρ =256x256
Typical 3D images: N of ρ =256x256x256

If thinking in megapixels – this is a low res camera

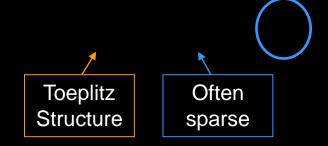
F Matrix entries are complex floats, so storage of matrix (single precision):
2D: dimension of F is (256X256)<sup>2</sup> ~ 34 GB
3D: (256X256X256)<sup>2</sup> ~ 2 PB

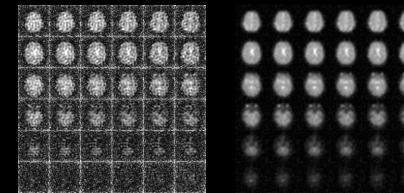


#### Solutions often derived by solving one or more

matrix inversions, e.g.,

11X speedup





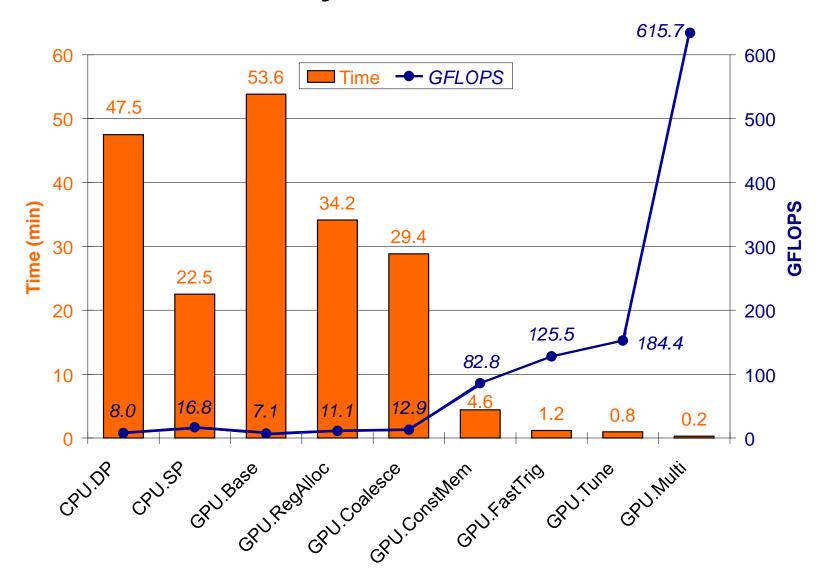
•  $F - M \times N$  matrix

• Typical *M* and *N* range from 2<sup>14</sup>-2<sup>24</sup> (higher in certain applications)

Ordinary matrix-vector multiplication
Complexity O(MN)
Easily parallelized
Matrix entries can be calculated on the fly
Lots of trigonometric function evaluations

• 28x speedup over CPU, 180 GFLOPs

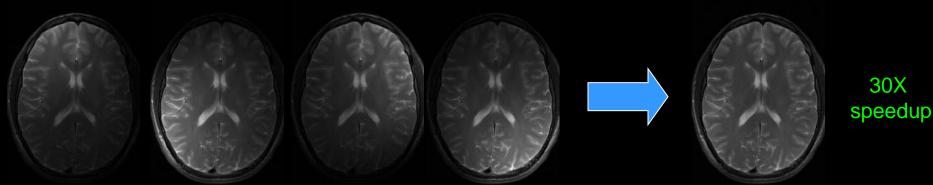
#### Summary of F<sup>H</sup>d Results



S. Stone et al., J Parallel Distrib Comput 68:1307-1318, 2008.

# Non-Fourier MR Reconstruction

- Parallel Imaging
  - Data acquired with multiple spatially diverse sensors



#### Field inhomogeneity correction



## <u>Summary</u>

- Common MRI computations have been accelerated by orders-of-magnitude using GPUs
  - Enables more practical use of advanced reconstruction algorithms to reduce scan time/image artifacts
  - Key primitives: 3D convolution, 3-D histogram, sparse/Toeplitz matrix-vector multiplication, sparse CG solver, (I) FFT
- Current challenge: To develop a common, modular framework for GPU reconstruction of MR data (and other imaging modalities)
  - Single framework for multi-core CPUs and many-core GPUs
  - Automatic tuning and selection for each primitive

#### • Future work:

- Continued optimization, scaling of reconstruction algorithms
- GPU implementation of MR parameter estimation
- Support for integration into production MRI pipelines

# Further Reading

#### <u>Journal</u>

• S. S. Stone, *et al.* "Accelerating Advanced MRI Reconstructions on GPUs." *Journal of Parallel and Distributed Computing* 68:1307-1318, 2008.

#### <u>Conferences</u>

- W.-m. W. Hwu, *et al.* "Accelerating MR Image Reconstruction on GPUs." *Proc IEEE ISBI*, Boston, 2009, pp. 1283-1286.
- S. S. Stone, *et al.* "Accelerating Advanced MRI Reconstructions on GPUs." *Proc ACM Computing Frontiers*, Ischia, 2008, pp. 261-272.
- J. P. Haldar, *et al.* "Fast MR Image Reconstruction using Graphics Processing Units." *Proc ISMRM*, Toronto, 2008, p. 1493.
- S. S. Stone, *et al.* "How GPUs Can Improve the Quality of Magnetic Resonance Imaging." *Proc GPGPU*, Boston, 2007.