

High Performance Medical Image Reconstruction. **What could go wrong?**

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What can go wrong?

- **algorithm selection and parameterization**
 - non-convex objective
 - non-convergence (ill-conditioned)
 - insufficient signal versus noise
- **implementation**
 - transcription errors
 - non-deterministic parallelization
 - model does not capture physics (including units)

COCONUT

Code CONstructing User Tool

- DSLs embedded in Haskell
 - ***strong*** types (detect errors in model and code)
- *high level*
 - start with objective function
 - differentiate symbolically
 - simplify expressions
 - recognize recurrence relations
 - generate code
- verification
 - SIMD & distributed parallelization (private and shared-memory)
 - *linear-time* verification via **AVOp model**
- algorithm selection and experiment design
 - homotope models to maintain convexity
 - optimize SNR and bound maximum noise (**SSV metric**)
 - scale regularization to ensure required convergence

Models

define variables

```
> let x = var3d (16,16,16) "x"  
> let y = var3d (16,16,16) "y"  
> ft (x +: y)  
(FT((x(16,16,16)+:y(16,16,16))))
```

define an objective function to minimize

$$\|ft(x + iy)\|_2$$

and take it's derivative

```
>> diff (mp ["X"]) (norm2 (ft (x +: y)))  
((((Re(FT((d(X[16] [16] [16])+:0.0[16,16,16])))).(Re(FT((X[16] [16] [16])+:Y[16] [16] [16]))))))  
+ ((Re(FT((d(X[16] [16] [16])+:0.0[16,16,16])))).(Re(FT((X[16] [16] [16])+:Y[16] [16] [16]))))))  
+ (((Im(FT((d(X[16] [16] [16])+:0.0[16,16,16])))).(Im(FT((X[16] [16] [16])+:Y[16] [16] [16]))))))  
+ ((Im(FT((d(X[16] [16] [16])+:0.0[16,16,16])))).(Im(FT((X[16] [16] [16])+:Y[16] [16] [16]))))))))
```

more complicated models

- e.g., **conservation of mass** in material transport models is a function

`massConservation (ThreeD vx, ThreeD vy, ThreeD vz)`
`= norm2 (conv3Zip1ZM com vx vy vz)`

where

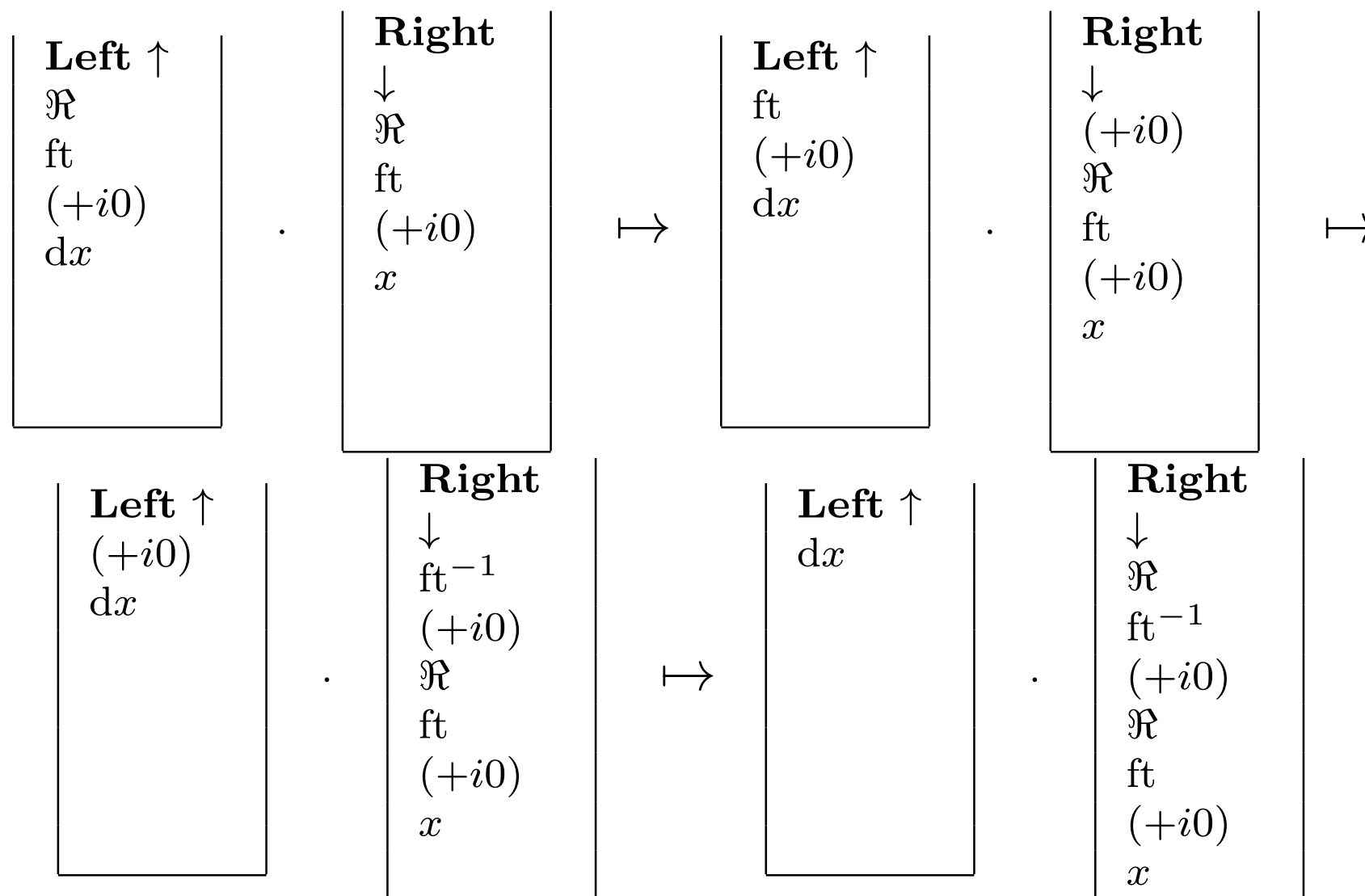
$$\begin{aligned} \text{com } (vX, vY, vZ) = & (vX[1, 0, 0] - vX[-1, 0, 0]) \\ & + (vY[0, 1, 0] - vY[0, -1, 0]) \\ & + (vZ[0, 0, 1] - vZ[0, 0, -1]) \end{aligned}$$

what was that $d(X)$, and where's ∇f ?

- $d(X)$ is a vector of differential forms
- comes from *implicit* derivative
- we can extract ∇f by simplifying

simplify rules

- commute $d(x)$ to LHS of dot
- move linear operators to RHS via adjoint



simplification DSL

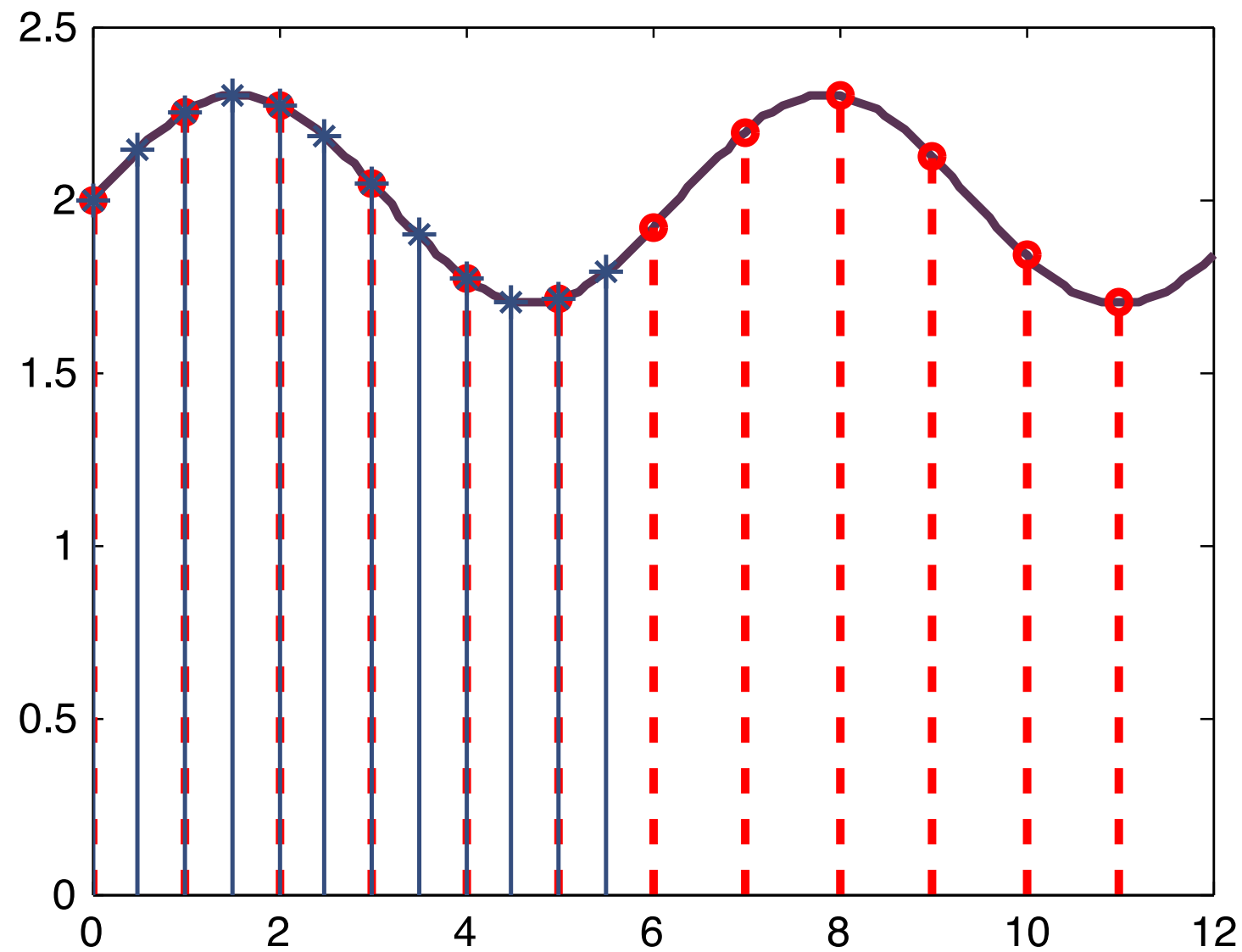
$[x * . (y * . z) \mid . \sim \sim > (x * y) * . z$
 $, 0 * x \mid . \sim \sim > 0$
 $, 0 * . x \mid . \sim \sim > 0$
 $, ft (invFt z) \mid . \sim \sim > z$
etc.

- natural algebraic simplification rules
- distinguishes scaling and multiplication

Type Checking (|| work)

- with stronger typing, compilers find more mistakes
- in C, all dynamic arrays (float*) look alike
- others check size and dimension
- we can do better

Arrays of Samples



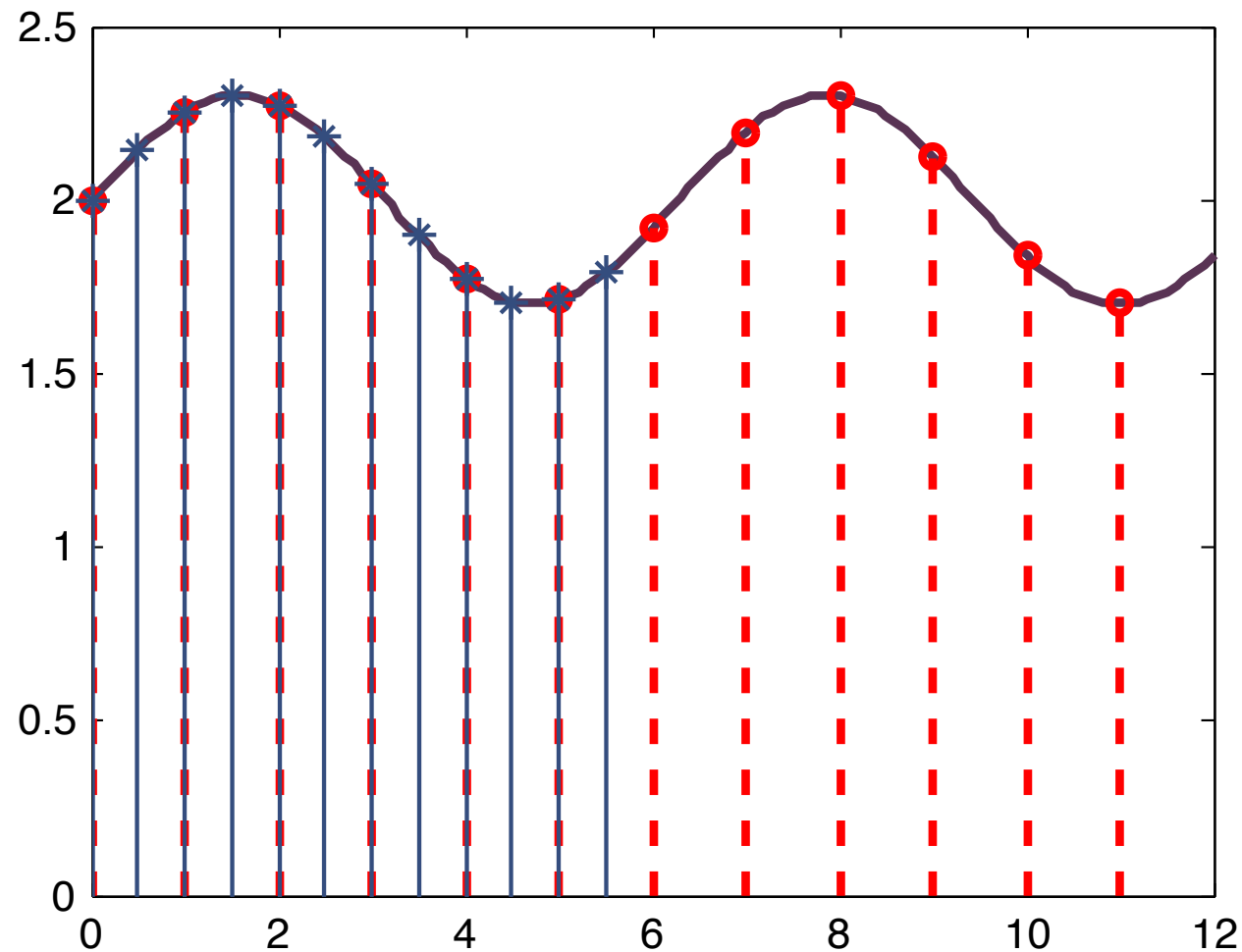
Arrays of Samples

- number of samples
- dimension
- frame of reference
- resolution (including units)
- units of measurement

Formalization

- in Haskell types define
 - physical units
 - array size and dimension
 - *frame of reference*

```
canalSample2  ::  Discretization1D  (F  "CanalFrame" )  
                                         (NAT  12)  
                                         [ float | 0.02 | ]  
                                         Meter  
                                         [ Double ]
```



- array sizes match, so try to add them ...

No **instance** for

(Add

(Discretization1D

(F "CanalFrame")

(NAT 12)

(FLOAT 'Pos **1** ('E_ 2))

(SIUnit ('M 1) ('S 0) ('Kg 0) ('A 0) ('Mol 0) ('K 0))

[**Double**])

(Discretization1D

(F "CanalFrame")

(NAT 12)

(FLOAT 'Pos **2** ('E_ 2))

(SIUnit ('M 1) ('S 0) ('Kg 0) ('A 0) ('Mol 0) ('K 0))

[**Double**])

Type Inference Across Linear Operations

- e.g., Fourier Transforms easy to break
 - Nyquist Sampling Theorem, etc.
- frame of reference
- resolution (including units)
- units of measurement

Classy Proofs

- Properties are Classes

```
class FT a b | a → b, b → a where  
  ft :: a → b  
  invFt :: b → a
```

```
class MultD3 f2 f1 f0 e2 e1 e0 g2 g1 g0 |  
  f2 f1 f0 e2 e1 e0 → g2 ,  
  f2 f1 f0 e2 e1 e0 → g1 ,  
  f2 f1 f0 e2 e1 e0 → g0 where
```

Proofs are Instances

```

instance (
  AssertDualFrames frame1 frame2, Frame frame1, Frame frame2,
  IsFloat stepSize2, IsFloat stepSize1, ToFloat numSamp ~ numSampF,

  and encode the Nyquist criterion:

  MultNZ stepSize1 numSampF t0,
  MultNZ t0 stepSize2 t1,
  t1 ~ FLOAT Pos 1 (E 0)

) ⇒
FT (Discretization1D frame1 numSamp stepSize1 rangeU [Complex Double])
   (Discretization1D frame2 numSamp stepSize2 rangeU [Complex Double])
  where
    ft (Discretization1D x) = (Discretization1D $ FFT.fft x)
    invFt (Discretization1D x) = (Discretization1D $ FFT.ifft x)

instance (
  Times f0 e0 p00h p00l, Times f1 e0 p10h p10l,
  Times f2 e0 D0 p20l, Times f0 e1 p01h p01l,
  Times f1 e1 D0 p11l, Times f2 e1 D0 D0,
  Times f0 e2 D0 p02l, Times f1 e2 D0 D0,
  Times f2 e2 D0 D0,
  Add3 p00h p10l p01l c1h c1l,
  Add5 p20l p01h p11l p02l c1h D0 c2l)
⇒ MultD3 f2 f1 f0 e2 e1 e0 c2l c1l p00l where

```


Readable Errors

- type inference fails on modelling errors
- ghc *loves* to throw up thousand-line errors
- we tamed it

```
No instance for (DualUnits  
  (SIUnit ('M 1) ('S 0) ('Kg 0) ('A 0) ('Mol 0) ('K 0))  
  (SIUnit ('M 0) ('S 0) ('Kg 0) ('A 0) ('Mol 0) ('K 0)))
```

Multi-Core = ILP Reinvented

Instruction Level Parallelism	Multi-Core Parallelism
CPU	Chip
Execution Unit	Core
Load/Store Instruction	DMA
Arithmetic Instruction	Computational Kernel
Register	Buffer / Signal

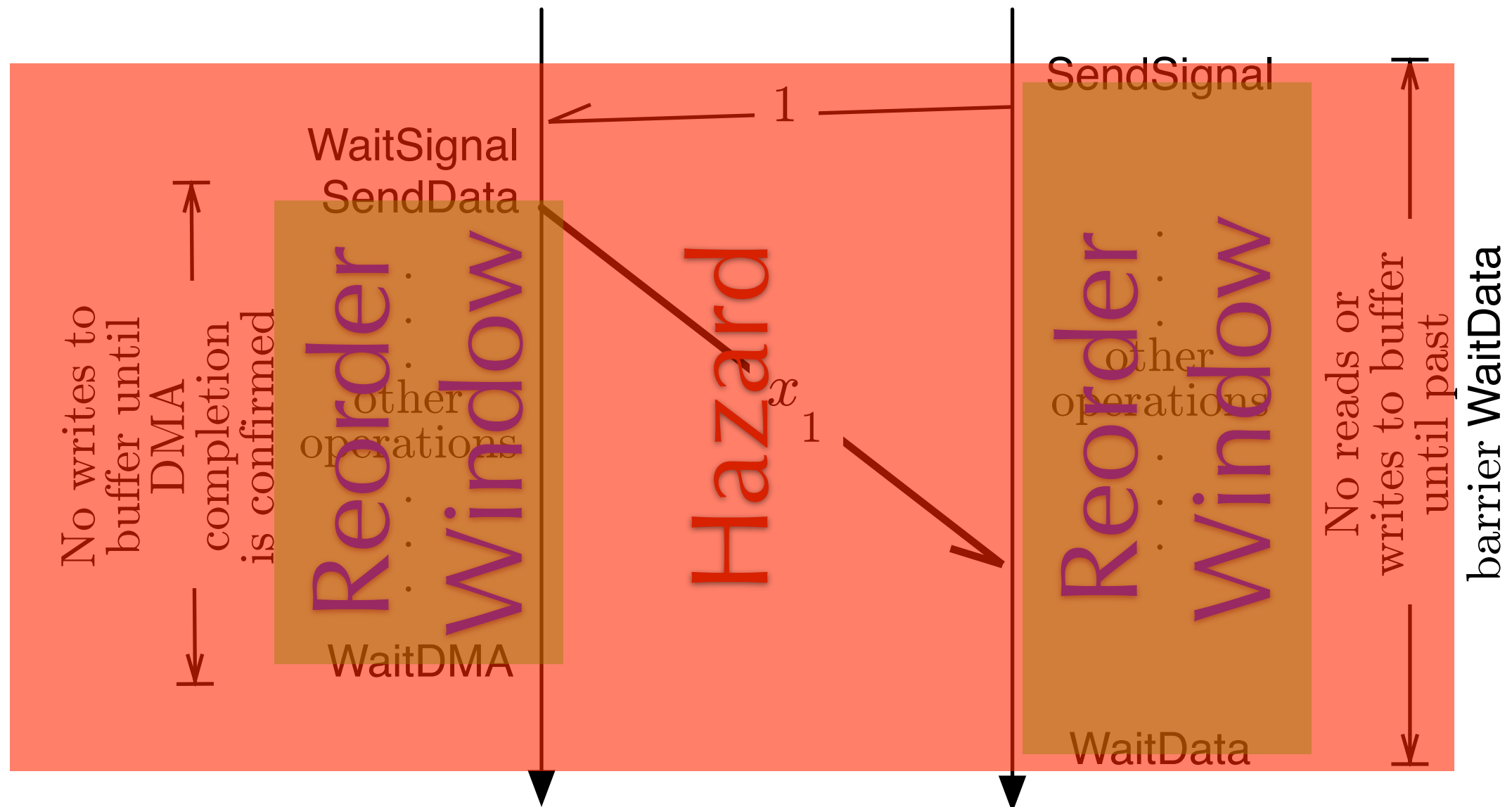
The Catch: Soundness

- on CPUs hardware maintains OOE
 - instructions execute out of order
 - hardware hides this from software
 - ensures order independence
- in our *Multi-Core virtual CPU*
 - compiler inserts synchronization
 - soundness up to software
 - uses asynchronous communication

Asynchronous

- no locks
 - locking is a multi-way operation
 - a lock is only local to one core
 - incurs long, unpredictable delays
- use asynchronous messages
 - matches efficient hardware

Async Signals



Multi-Core Language

AVOps

Computation <i>operation bufferList</i>	do a computation with local data
SendData <i>localBuffer remoteBuffer tags</i>	start DMA to send local data off core
WaitData <i>localBuffer tag</i>	wait for arrival of DMAed data
WaitDMA <i>tag</i>	wait for locally controlled DMA to complete
SendSignal <i>core signal</i>	send a signal to distant core
WaitSignal <i>signal</i>	wait for signal to arrive
Loop <i>n π body</i>	<i>body; π(body); $\pi(\pi(\text{body}))\dots$</i>

locally Sequential Program

index	core 1	core 2	core 3
1	SendSignal $s \rightarrow c2$	long computation	SendSignal $s \rightarrow c2$
2			
3		WaitSignal s	
4		computation	
5			
6		WaitSignal s	

- total order for instructions
 - easier to think in order
- send precedes wait(s)

NOT *sequential*

index	core 1	core 2	core 3
2	SendSignal $s \rightarrow c2$		
5			SendSignal $s \rightarrow c2$
	<i>second signal overlaps the first, only one registered</i>		
1		long computation	
3		WaitSignal s	
4		computation	
	<i>no signal is sent, so the next WaitSignal blocks</i>		
6		WaitSignal s	

- can execute out of order

does NOT imply *order independent*

index	core 1	core 2	core 3
1		long computation	
5			SendSignal $s \rightarrow c2$
3		WaitSignal s	
4		computation using wrong assumptions	
2	SendSignal $s \rightarrow c2$		
6		WaitSignal s	

Linear-Time Verification

- must show
 - results are independent of execution order
 - no deadlocks
- need to keep track of all possible states
- linear in time = one-pass verifier
 - constant space
 - = max possible states at each instruction

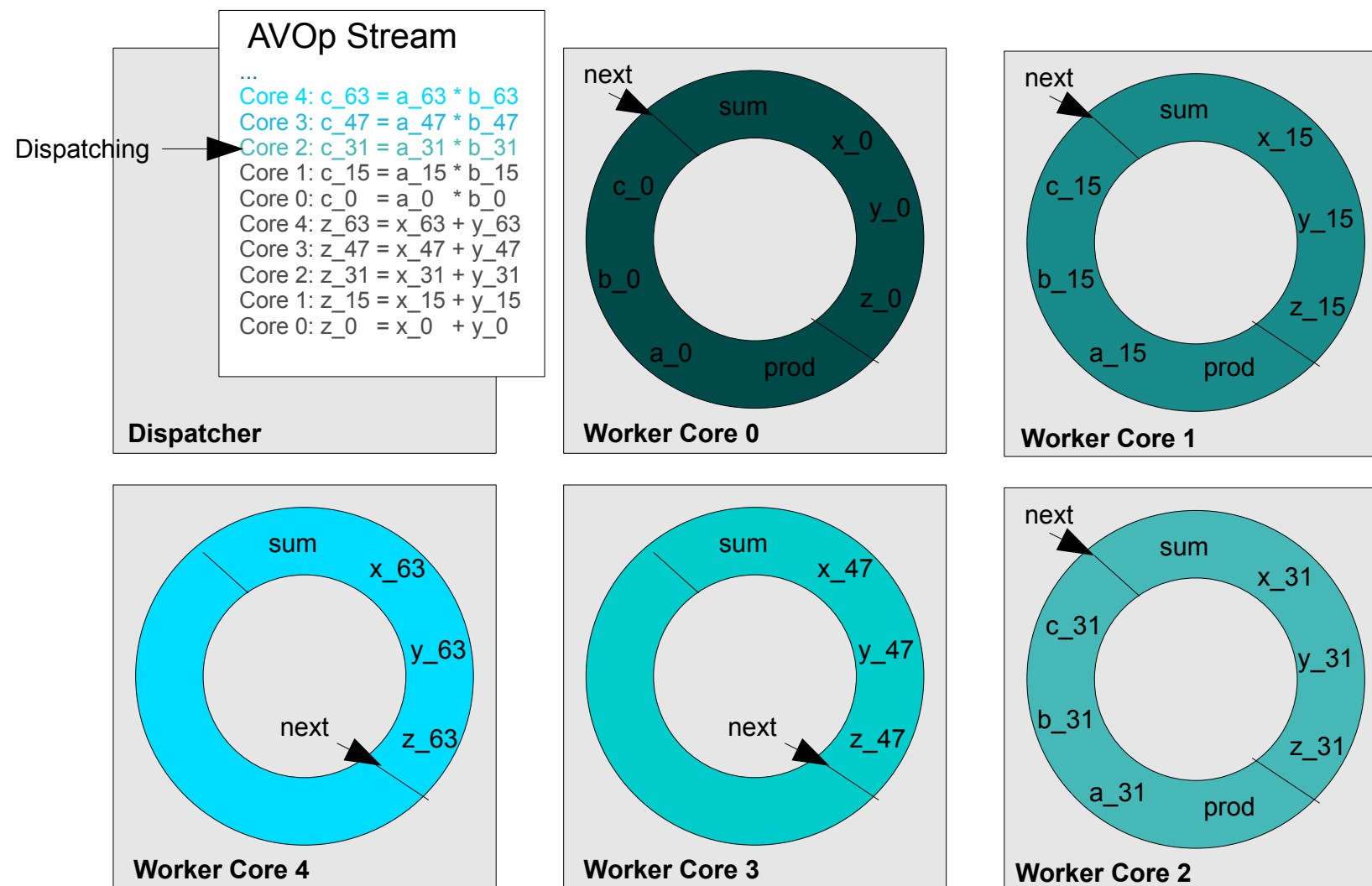
Impact

- no parallel debugging !!
- every optimization trick used for ILP can be adapted

But...

- assumes you have signals
 - what about shared memory?
 - still lock-free synchronization?
 - let's try it on x86

Single-Reader, Single-Writer AVOp Ring Buffers (SRSWARB)



Limit Hazards

- AVOps can only conflict if they can fit in the Ring Buffers at the same time
 - on each core, AVOps are sequential, therefore safe
 - reads on different cores are safe
 - while write AVOp is on a core, check that no other core reads or writes

Nest Step

- signals across nodes
- SRSWARB system for multicore
 - new system for GPU

Details ...

Michal Dobrogost, Christopher Anand, Wolfram Kahl, Verified Multicore Parallelism using Atomic Verifiable Operations, accepted for publication in *Multicore Technology: Architecture, Re-configuration and Modeling*, Muhammad Yasir Qadri and Stephen J. Sangwine (eds), CRC Press., 2013, 107–151.

Christopher Kumar Anand, Wolfram Kahl, Synthesising and Verifying Multi-Core Parallelism in Categories of Nested Code Graphs; “Process Algebra for Parallel and Distributed Processing (Algebraic Languages in Specification-Based Software Development)”, eds. Michael Alexander and William Gardner, Chapman and Hall/CRC, 2008, 3–45.

Jessica L M Pavlin and Christopher Kumar Anand, Symbolic Generation of Parallel Solvers for Inverse Imaging Problems, CAS-14-05-CA.

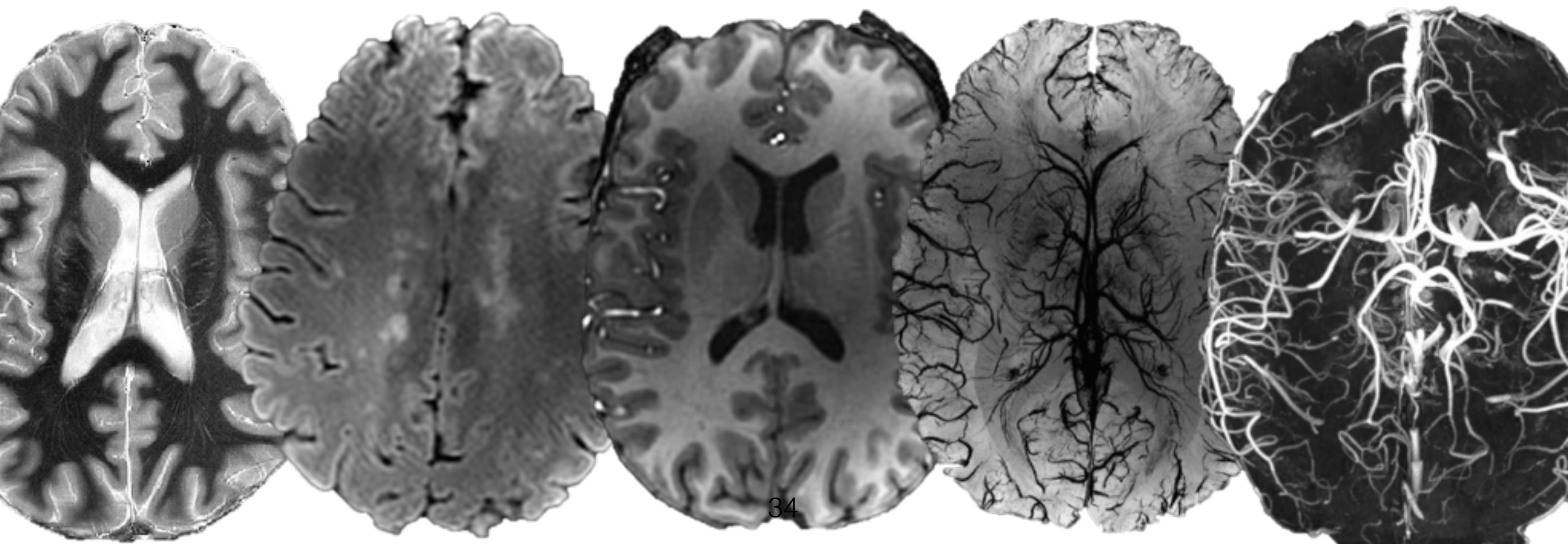
Maryam Moghadas, **Yuriyy Toporovskyy**, Christopher Kumar Anand, Type-Safety for Inverse Imaging Problems, CAS-14-04-CA.

It Works !

- used to generate, from the objective function, a multi-core (shared memory) image reconstruction software for **parallel Magnetic Resonance Imaging** for AllTech Medical Systems America

MRI

- Image values & tissue contrast depends on how the experiment is conducted
- Qualitative and quantitative

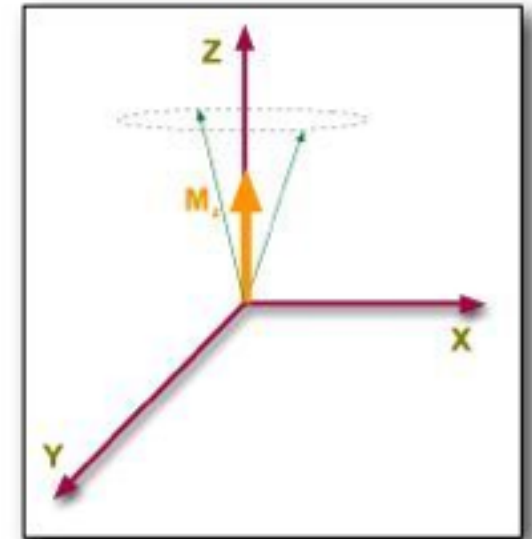
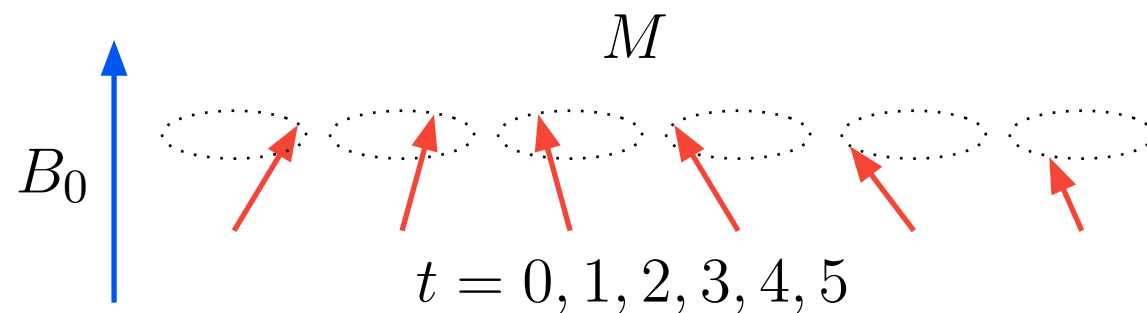


Quantum Mechanical Foundation of MRI

slide contents tunnelled away

Signal and Contrast

- Sample magnetization M - vector



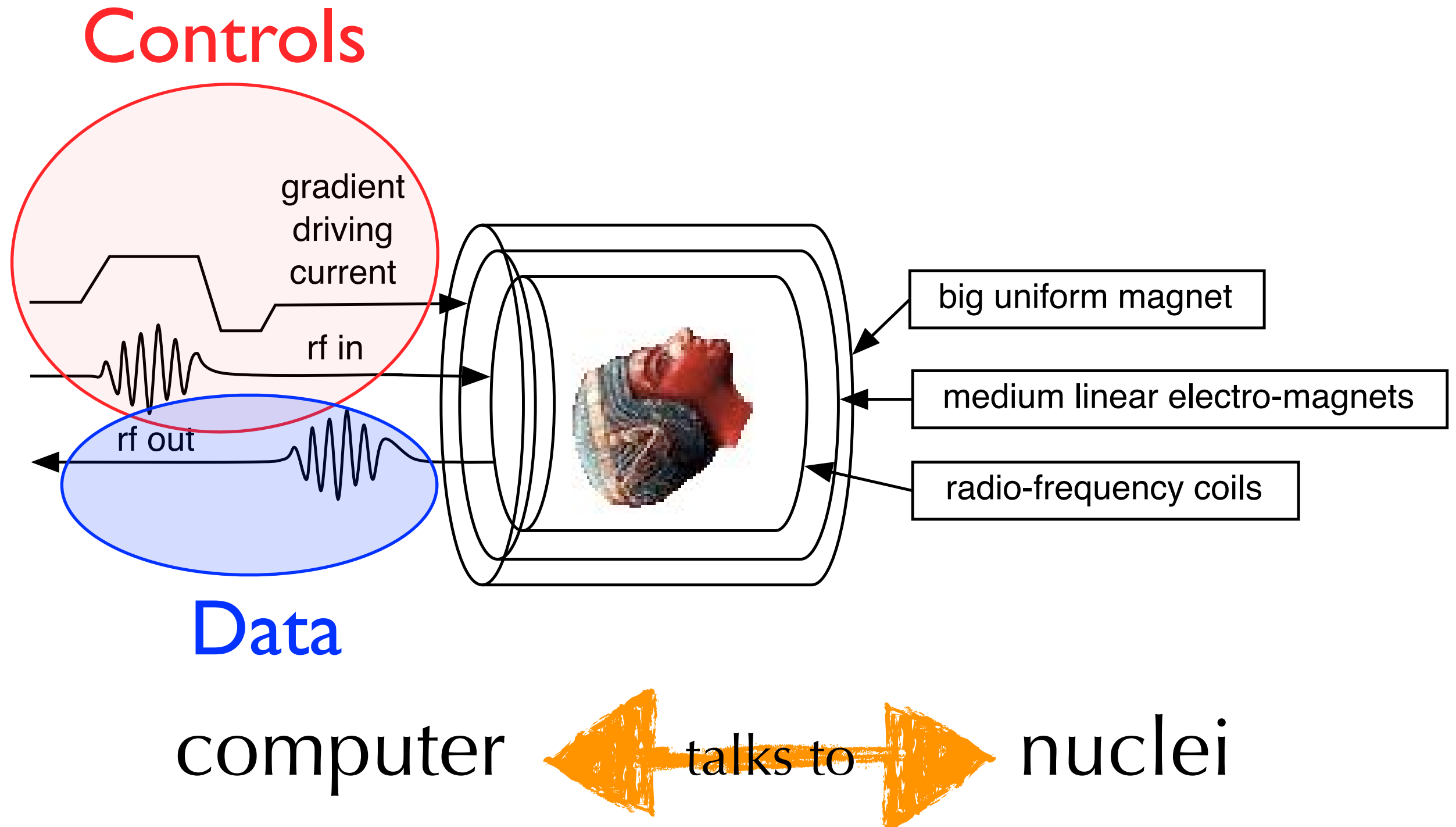
- Evolves in time \longrightarrow Bloch Equation

$$\frac{dM}{dt} = M \times B_0$$

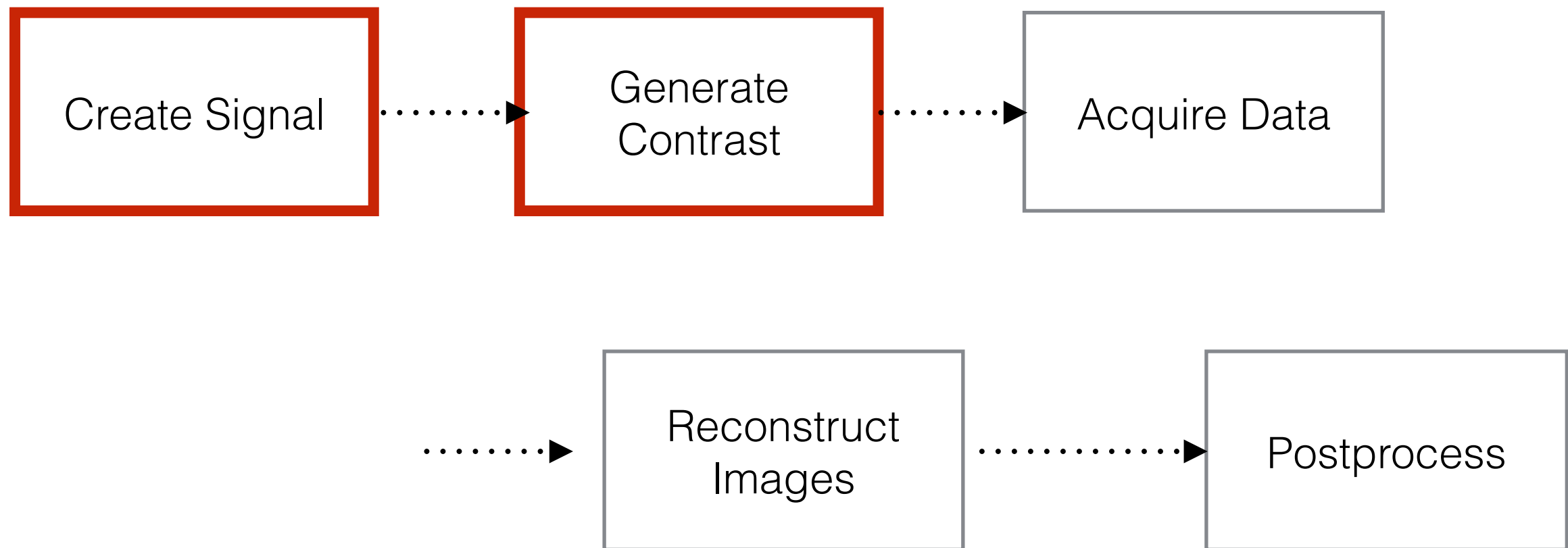
- Controllable by application of RF and magnetic fields:

$$\frac{d}{dt} \begin{pmatrix} M_x \\ M_y \\ M_z \end{pmatrix} = \begin{pmatrix} -R_2 & -\Omega^0 & bi \\ \Omega^0 & -R_2 & -br \\ -bi & br & -R_1 \end{pmatrix} \begin{pmatrix} M_x \\ M_y \\ M_z \end{pmatrix} + \begin{pmatrix} 0 \\ 0 \\ R_1 \times M_e \end{pmatrix}$$

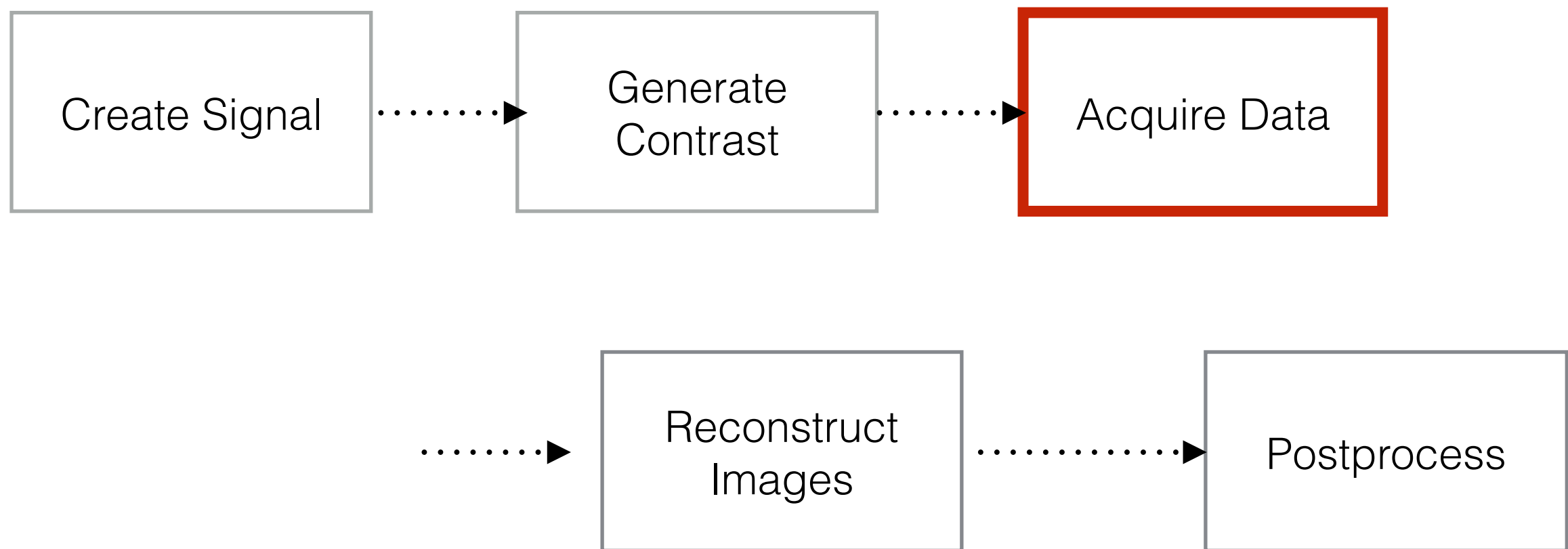
Controls and Data



MRI Experiment

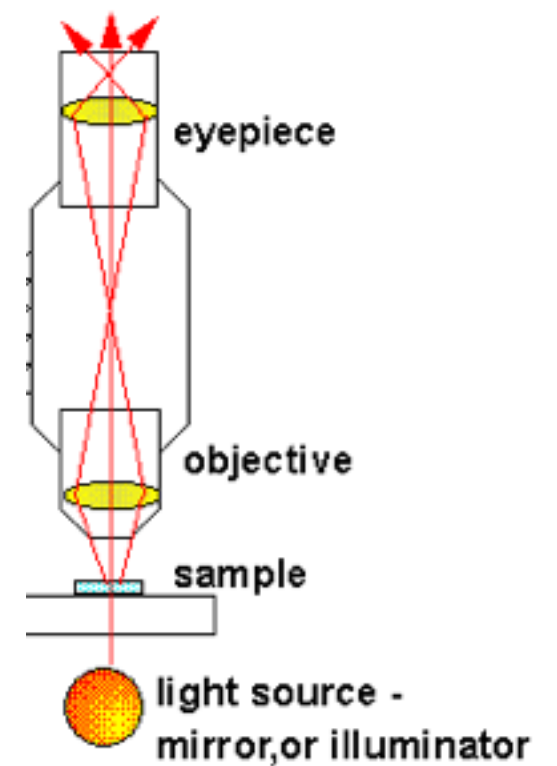
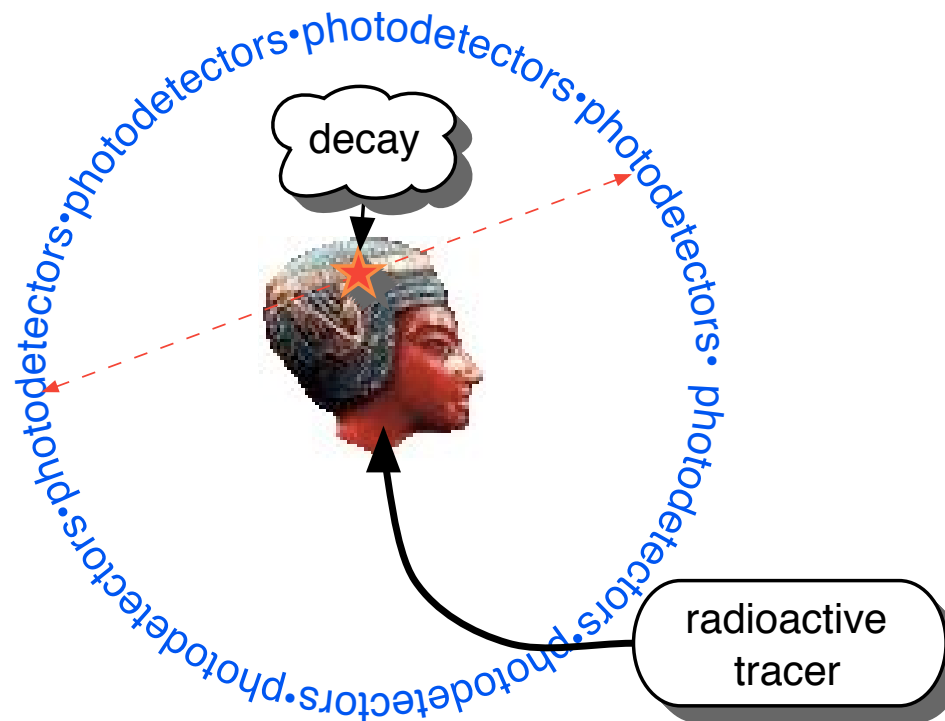
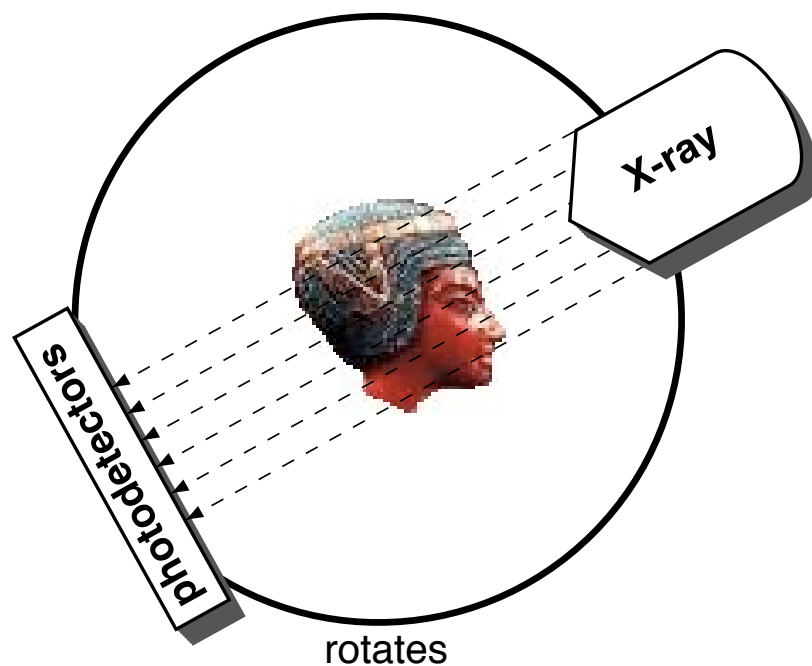


MRI Experiment



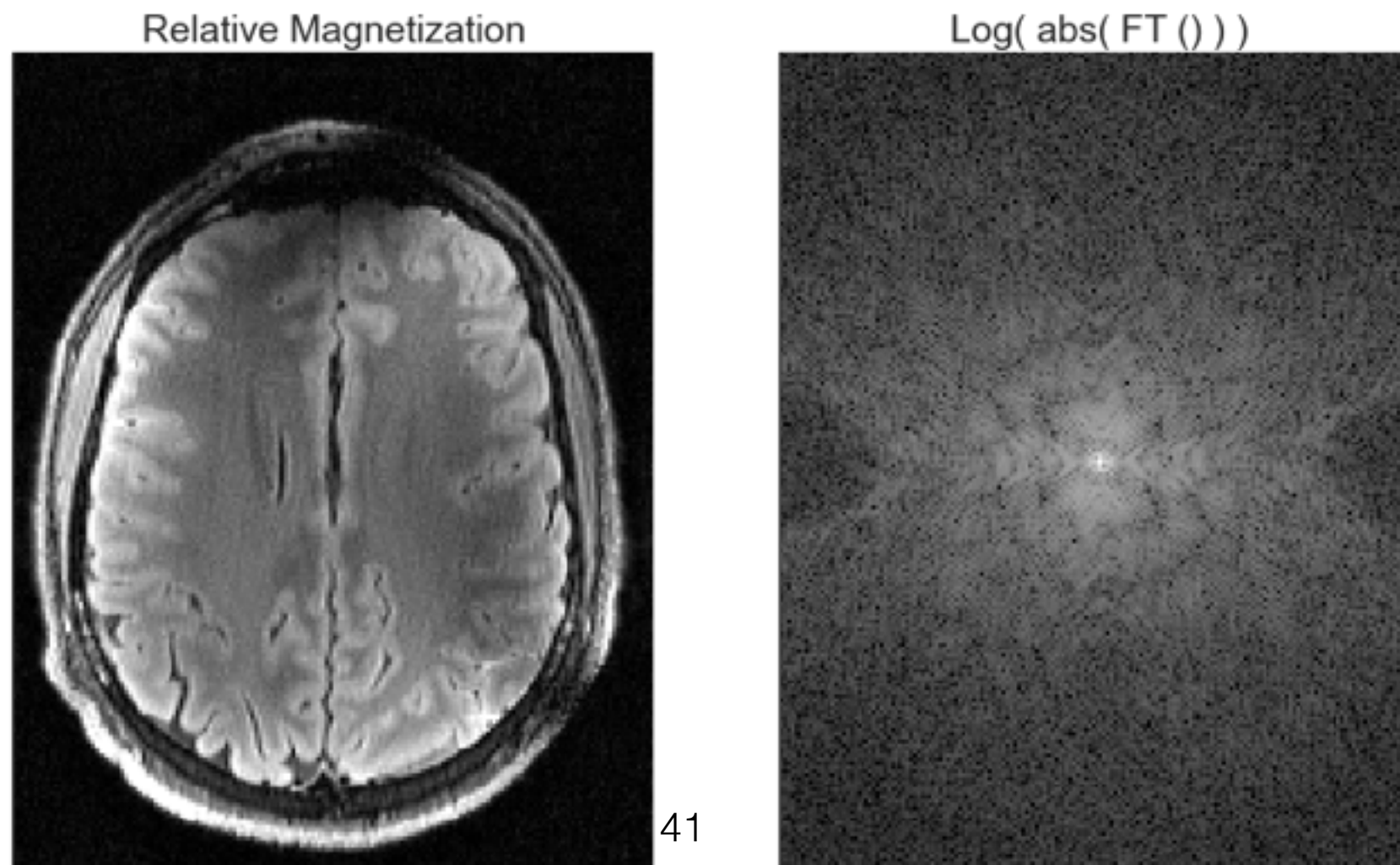
Data Acquisition

- Unlike x-ray/CT, PET, or Microscopy, we **don't** capture projections or reflections of the incident waves



K-Space

- Via some Nobel prize winning work:
- Modulate phase of magnetization in space & measure DFT. — frequency domain

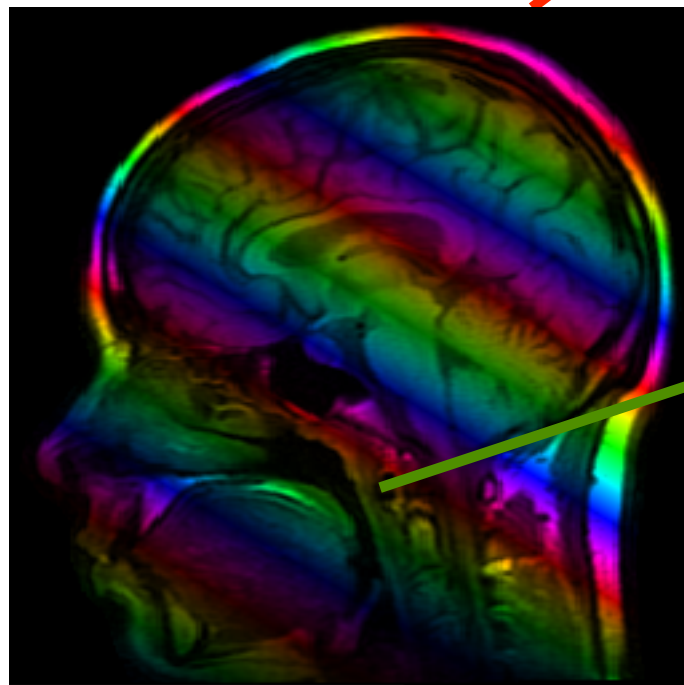


sampled
Fourier
Transform

k-space

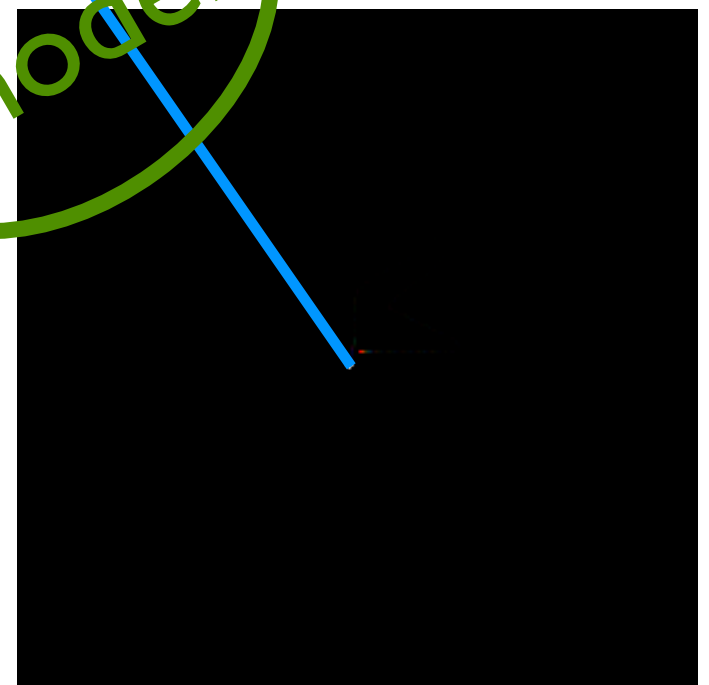
linear phase
variation

$$s(t) = \int_{\mathbb{R}^3} e^{i\langle x, k(t) \rangle} \rho(x) dx,$$



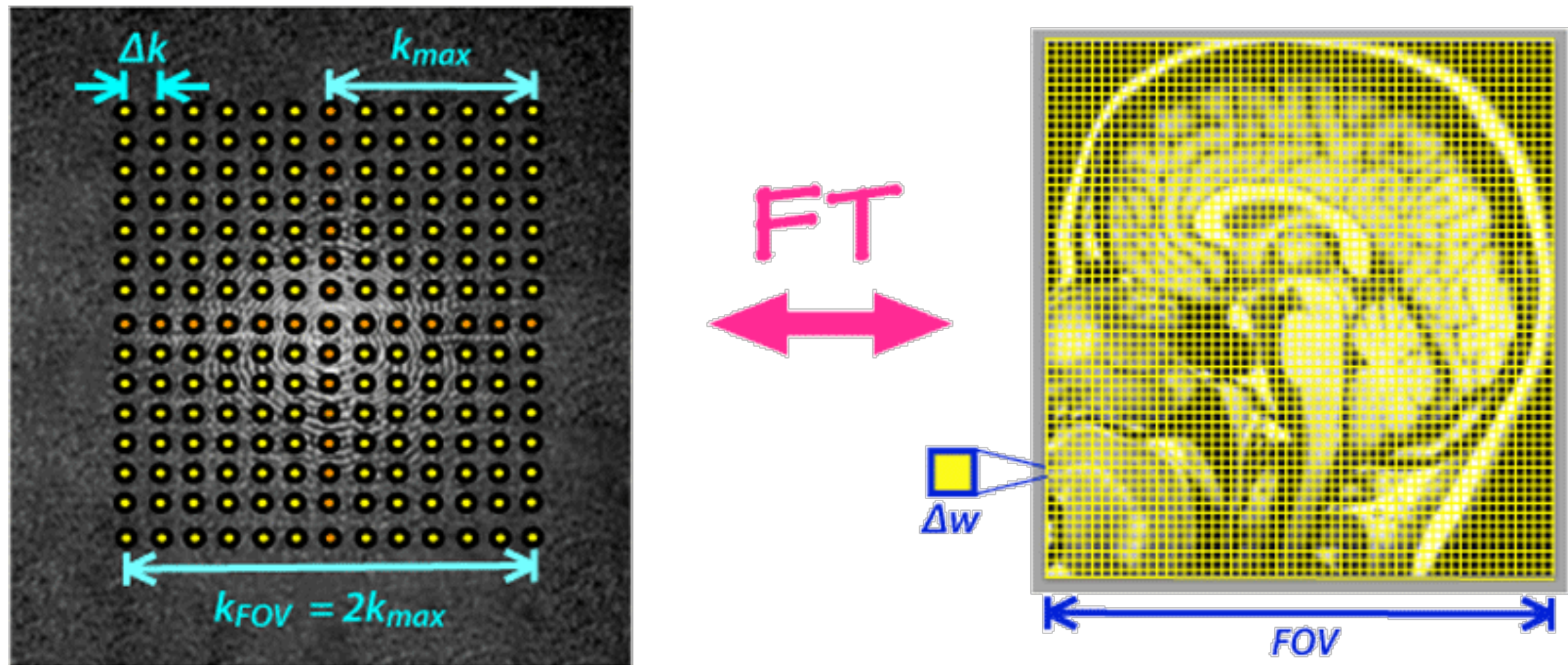
✧ each sample
point gives
spatial
information

model



K-space properties

- Nyquist, my friend. Sample rate \sim kspace spacing

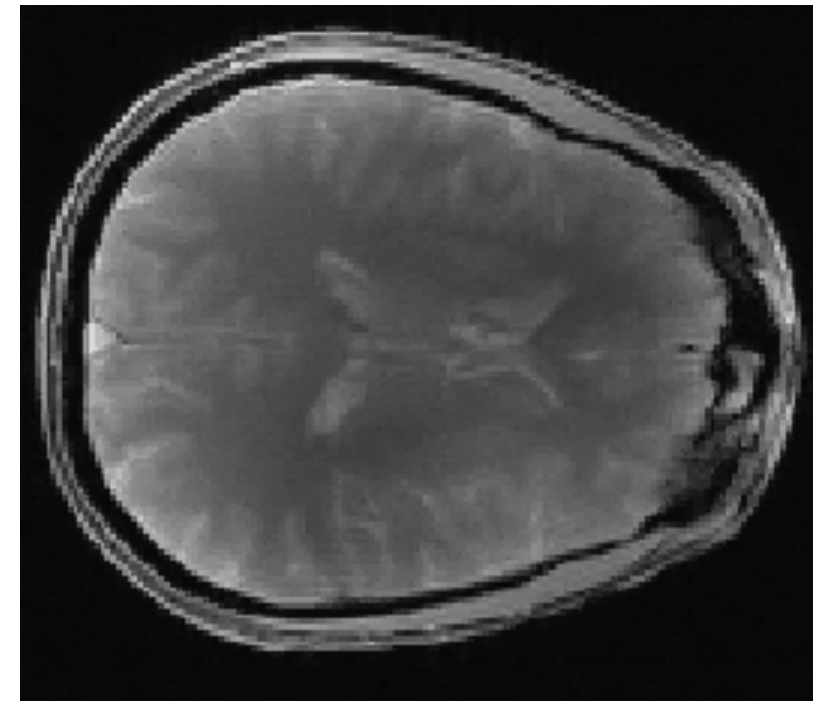


$$\Delta k = 1 / FOV$$

$$\Delta w = 1 / k_{FOV}$$

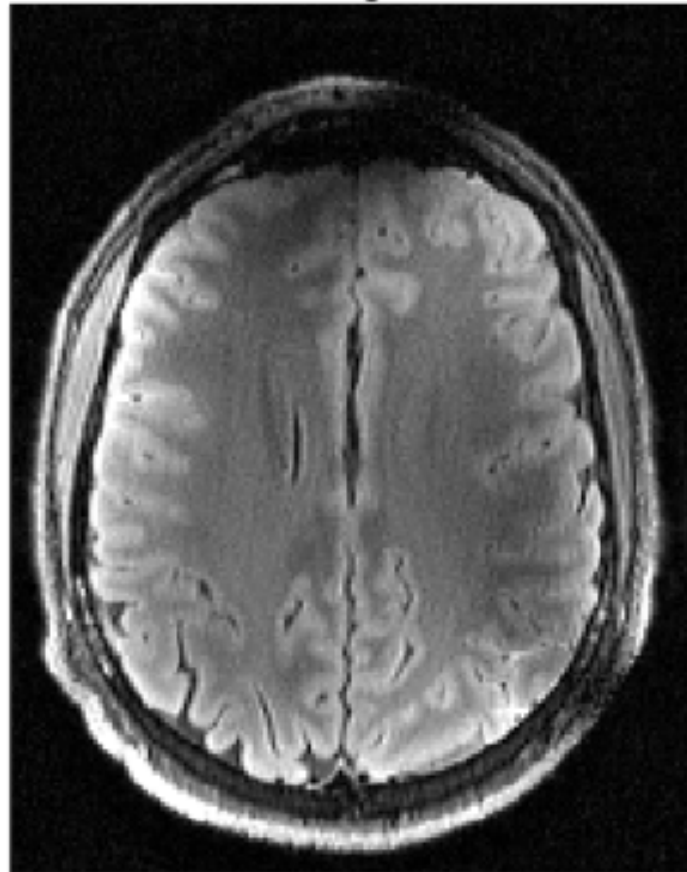
Receiver Elements

- In addition to phase modulation for k-space sampling, data is acquired via multiple receivers
 - spatial coverage
 - improved signal
 - spatial encoding

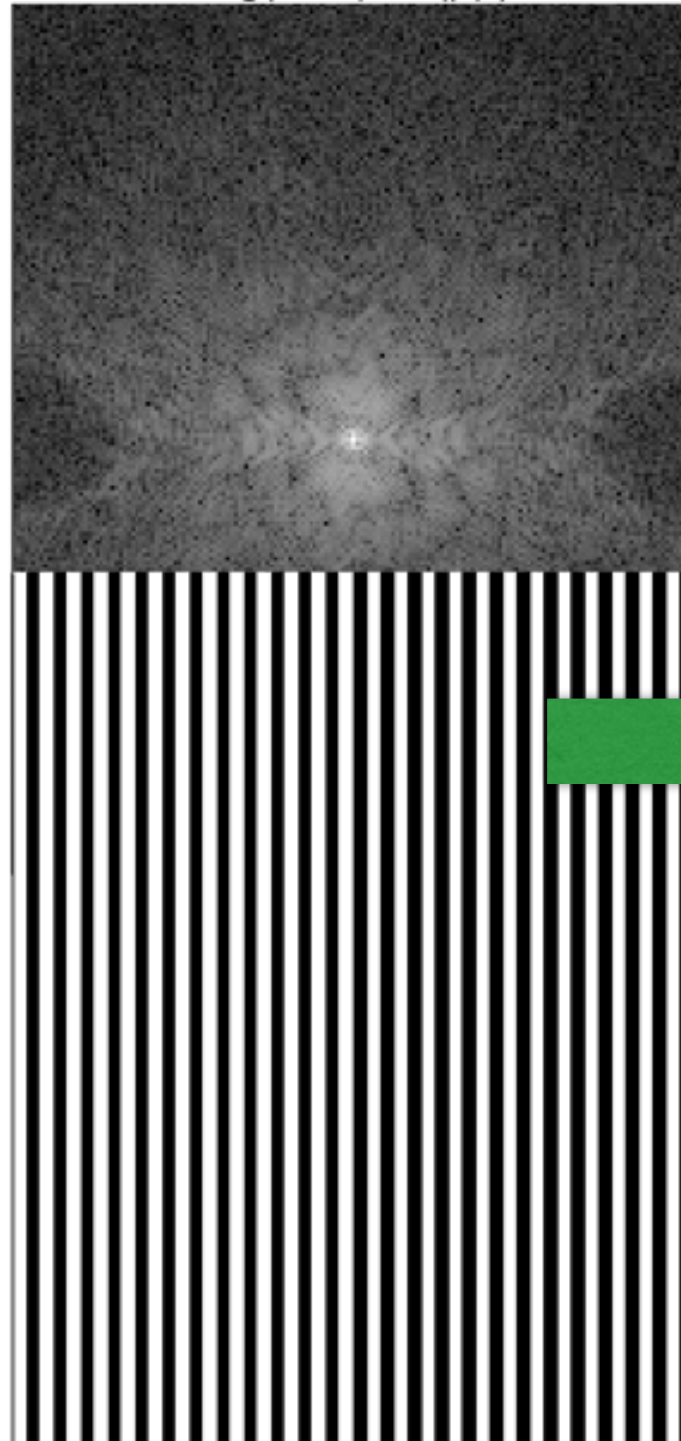


Parallel Imaging By Example

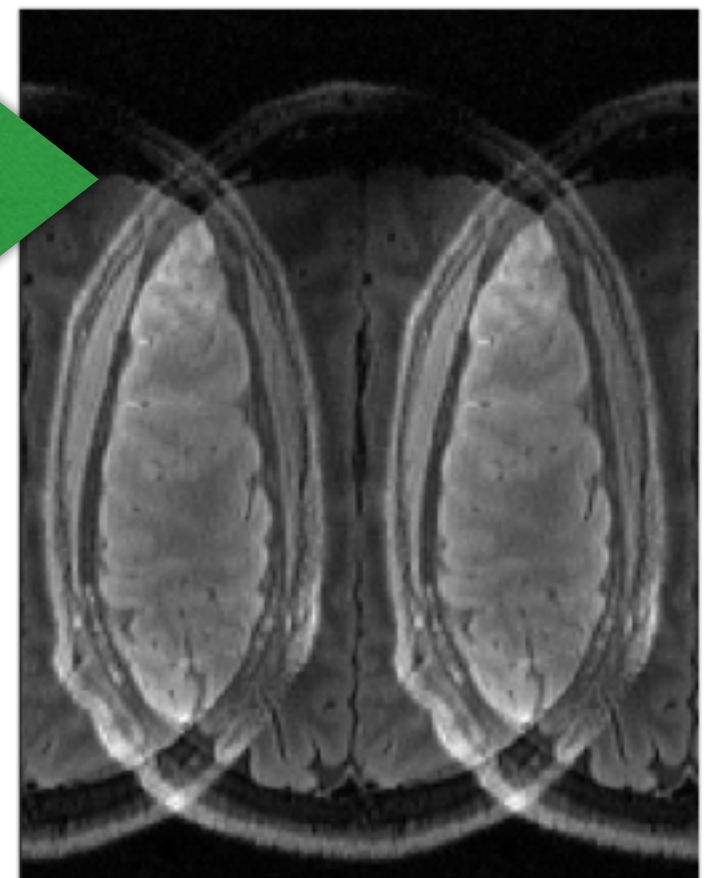
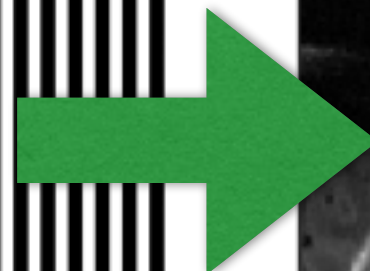
Relative Magnetization



$\text{Log}(\text{abs}(\text{FT}()))$



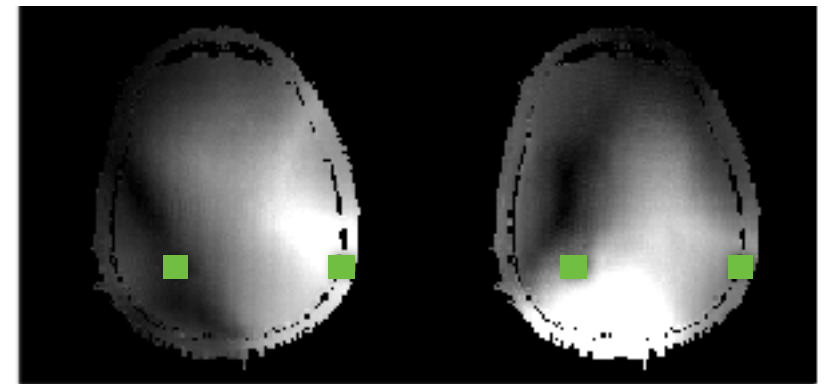
Aliasing &
FOV Reduction!



Original formulation:
SENSE

Parallel Imaging by Example

- How do we recover the underlying magnetization?



- Recall: multiple receivers

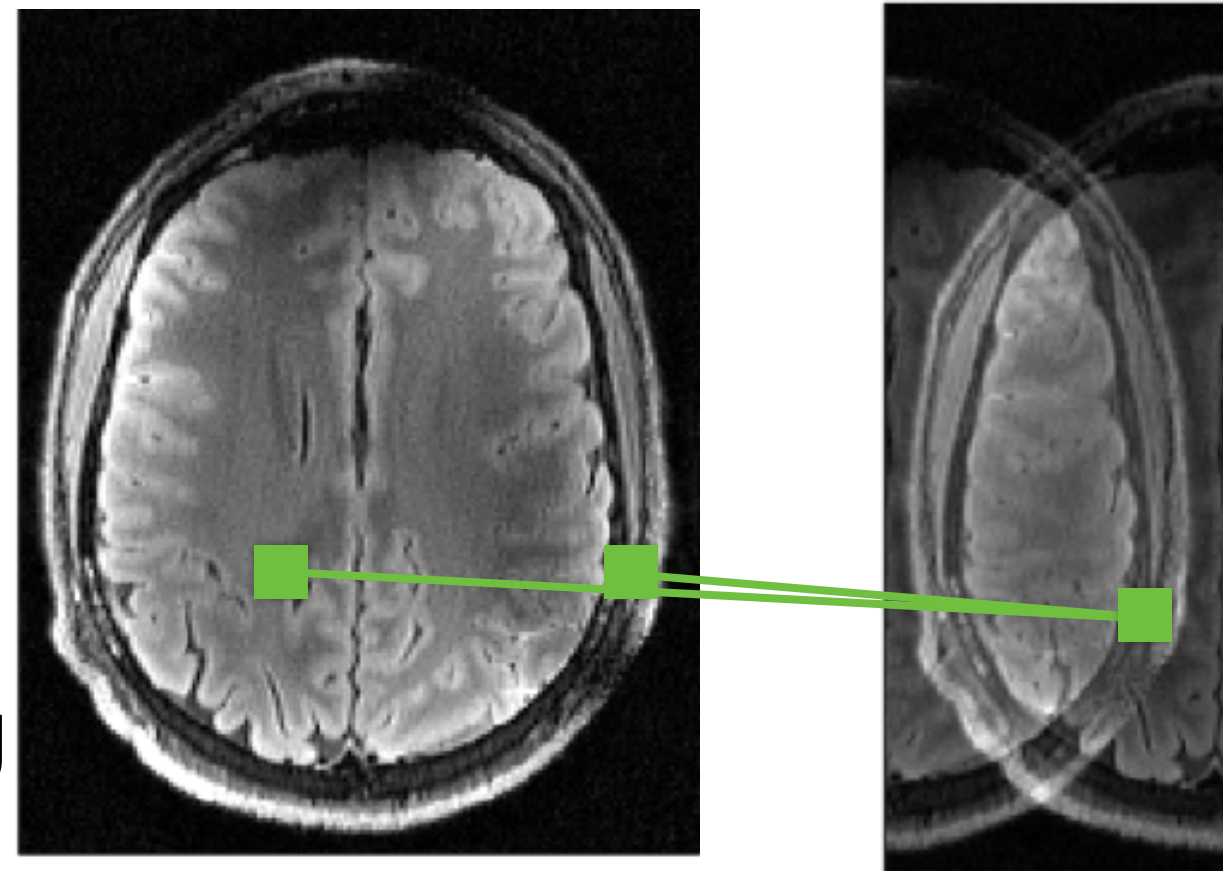
- Consider two overlapping voxels and receivers a & b

- Signal from a and b ~

$$s_a = c_a \cdot m_1 + c_a \cdot m_2$$

$$s_b = c_b \cdot m_1 + c_b \cdot m_2$$

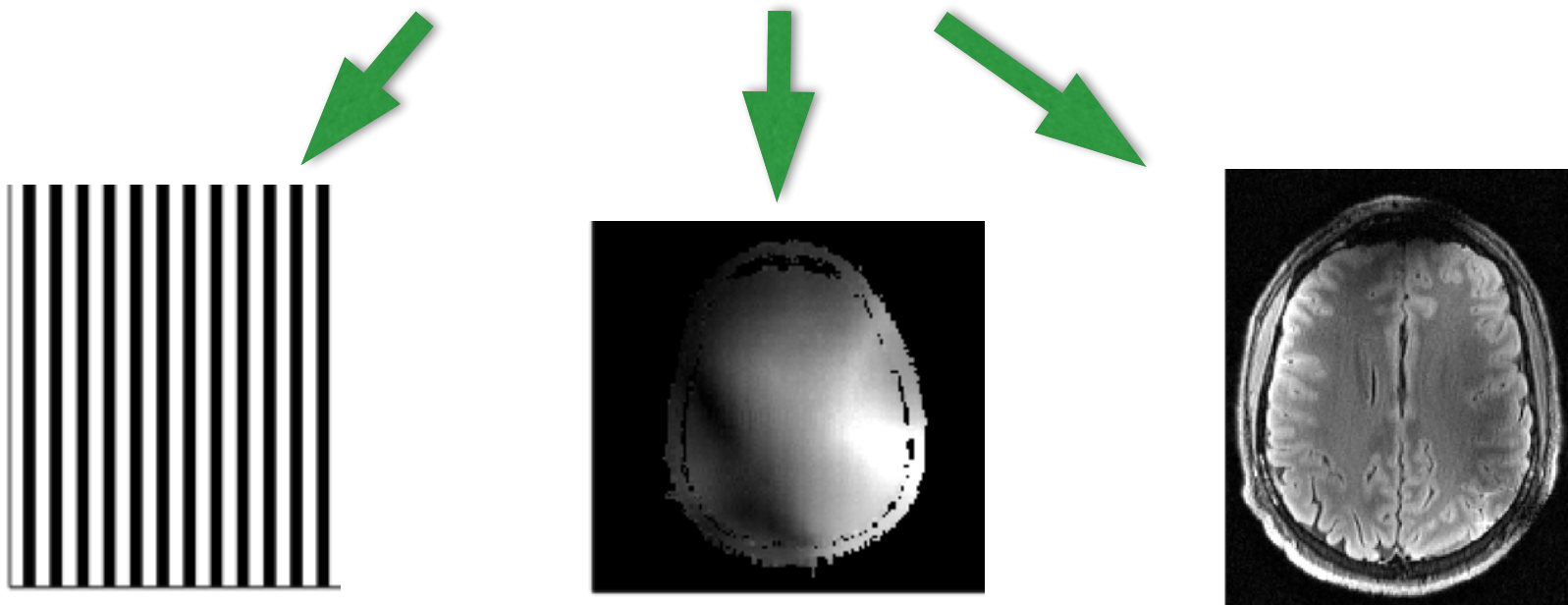
- Only works for nice aliasing



Forward Problem

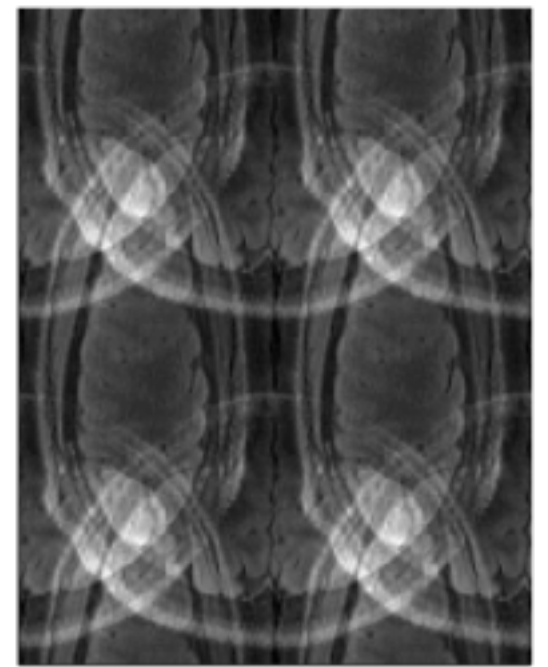
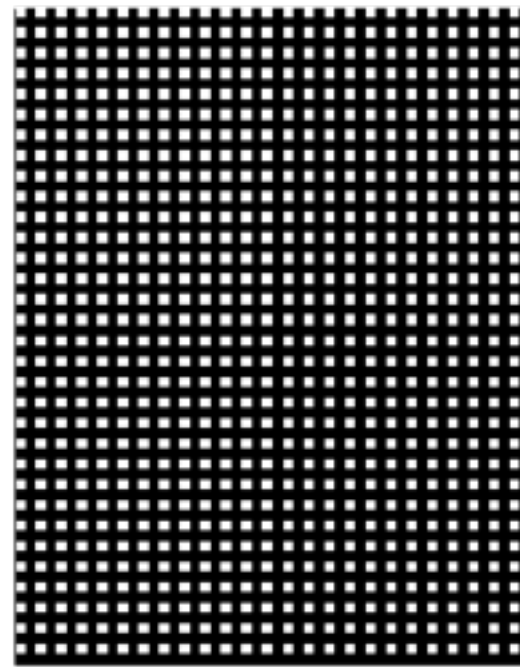
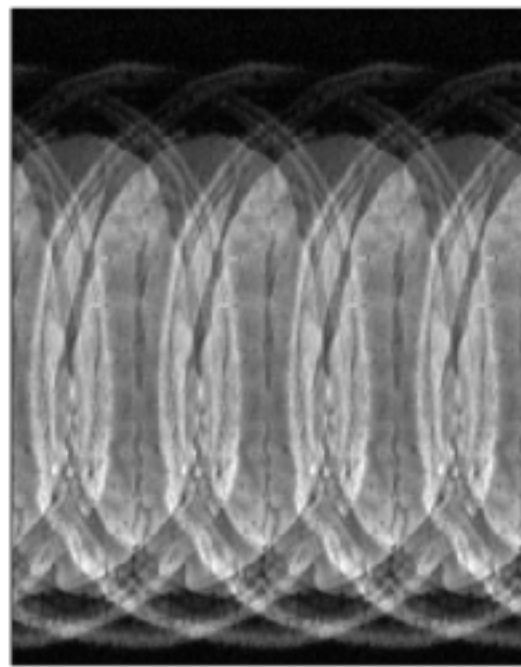
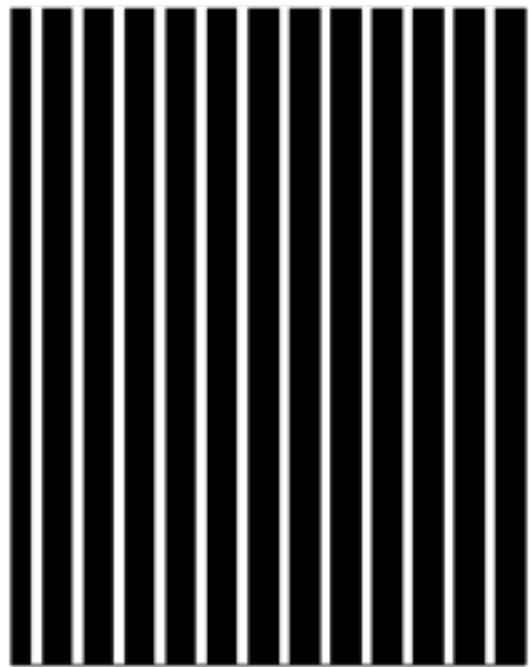
- Extending the previous example, more general image acquisition process:
- measurement =
k-encoding * FT * receive sensitivity * magnetization

$$M_i = k \cdot FT \cdot S_i \cdot \rho_0, \quad i = 1 \dots nC$$



Inverse Problem ?

- Reduction factor of 4:
Vastly different practical performance thanks to receiver profiles.

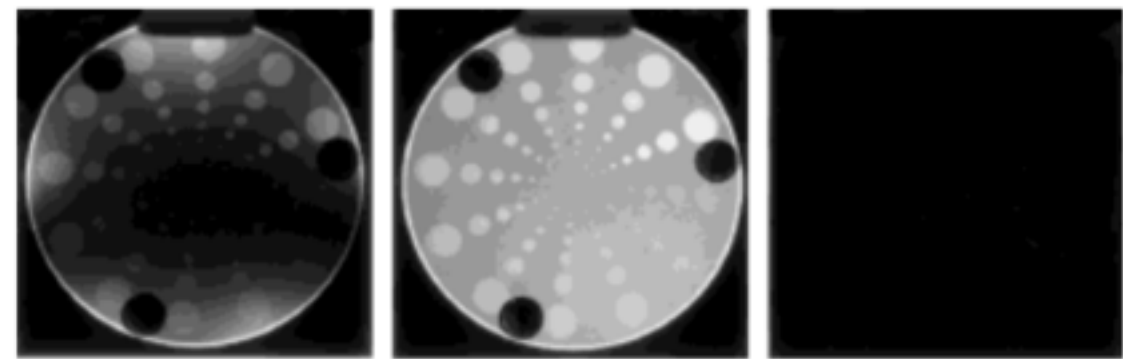


- *How do we measure which will be better before running the MRI?* **Expected reconstruction error!**

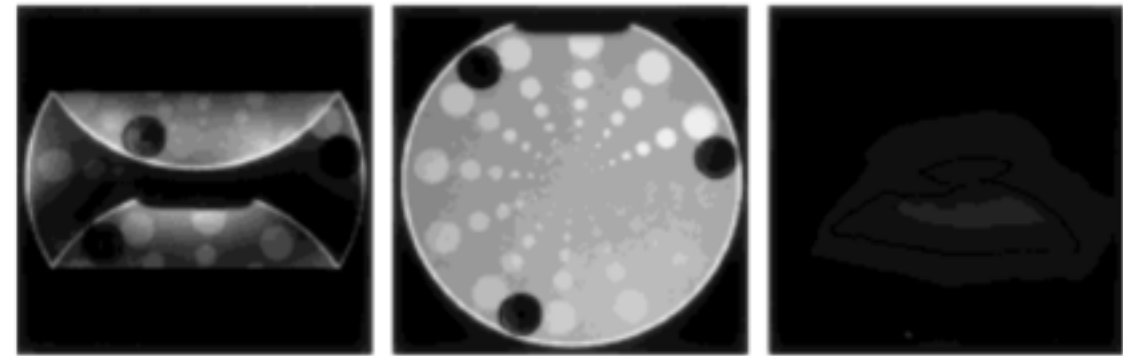
Geometry-factor

- Gold standard metric for assessing effect of under sampling patterns on the resultant image
- Estimates difficulty of unaliasing **pairs**/sets of spatial overlapping voxels
- n values. What is important? mean? max? 95%ile?

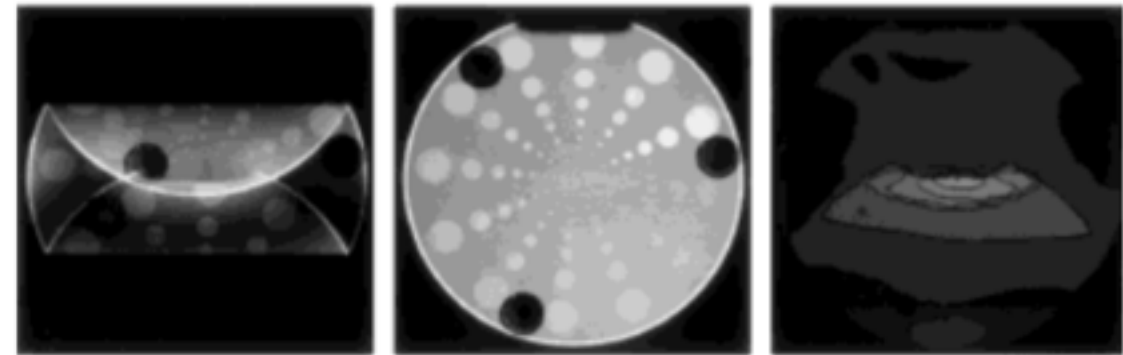
1.0



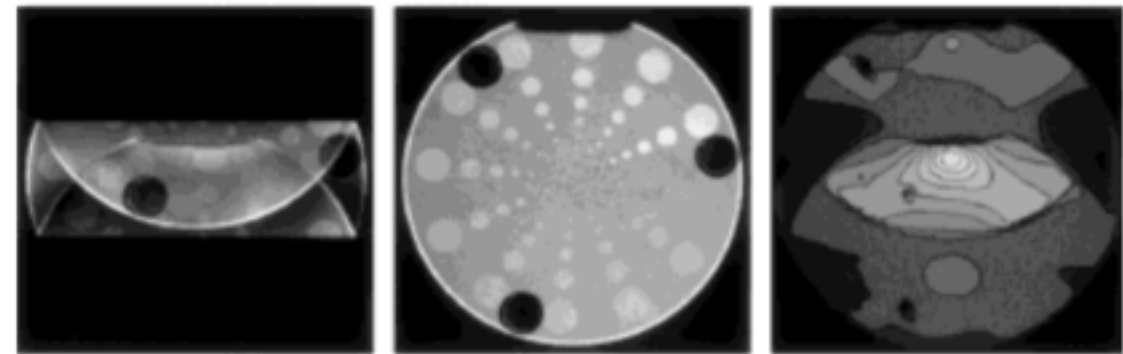
2.0



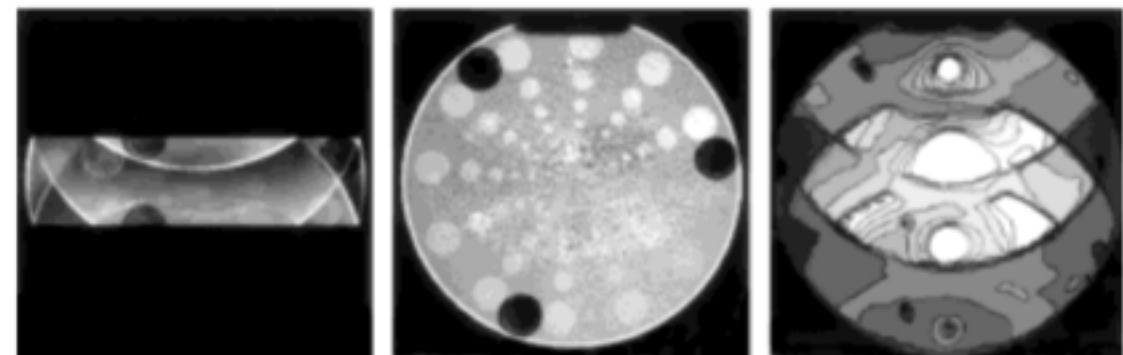
2.4



3.0

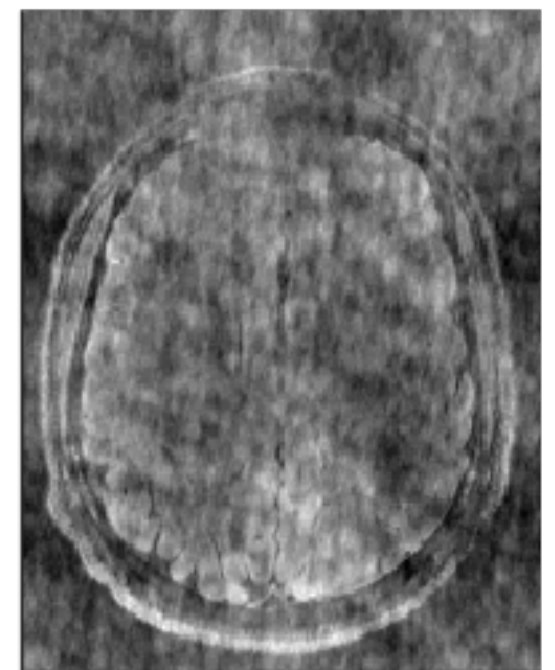
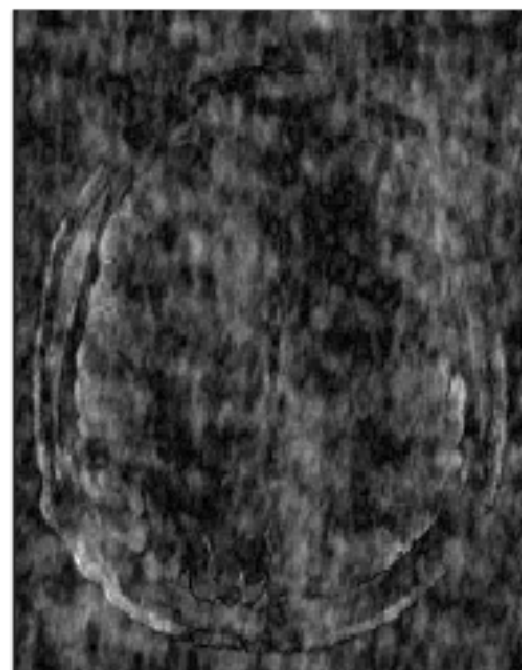
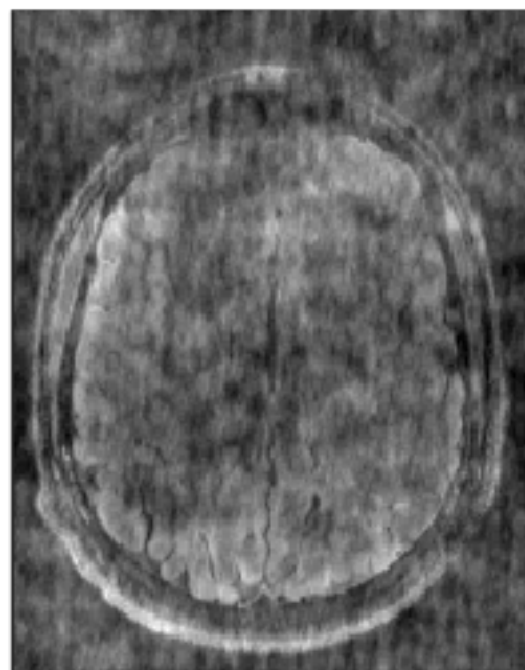
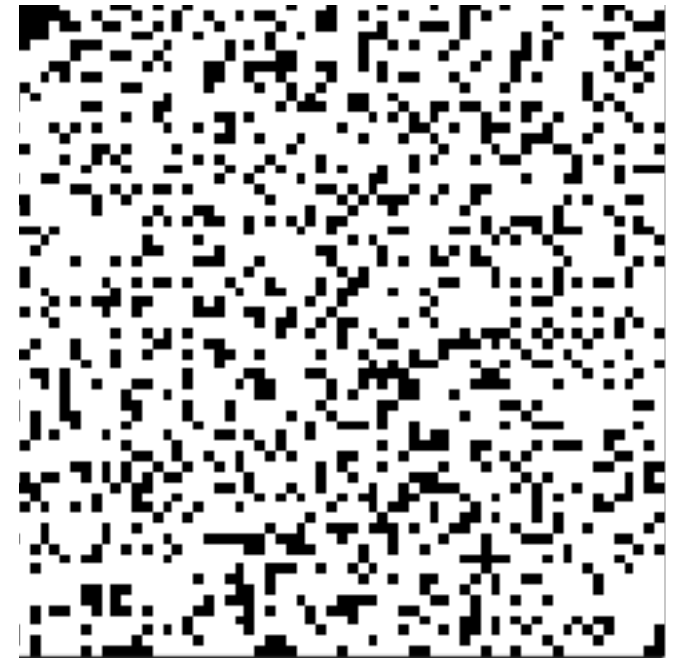


4.0



Dense aliasing

- For non-standard grids the aliasing patterns and G-factor are complicated, because signal can alias anywhere
- No simple DFT relation



Problem with Gold

- takes too long to mine (hours or days)
- bad for environment

New Approach

- Estimate largest and smallest *magnitude* singular values of M
$$M_i = k \cdot FT \cdot S_i \cdot \rho_0, \quad i = 1 \dots nC$$
- Physical interpretations
- Smallest singular value: **SSV** Metric
 - worst case noise amplification from reconstruction
- Largest singular value:
 - maximum of 1 with proper scaling.
- Large/small ~ condition #

SSV Metric Method

- M is rectangular!

Form normal system $M^H M$ and find

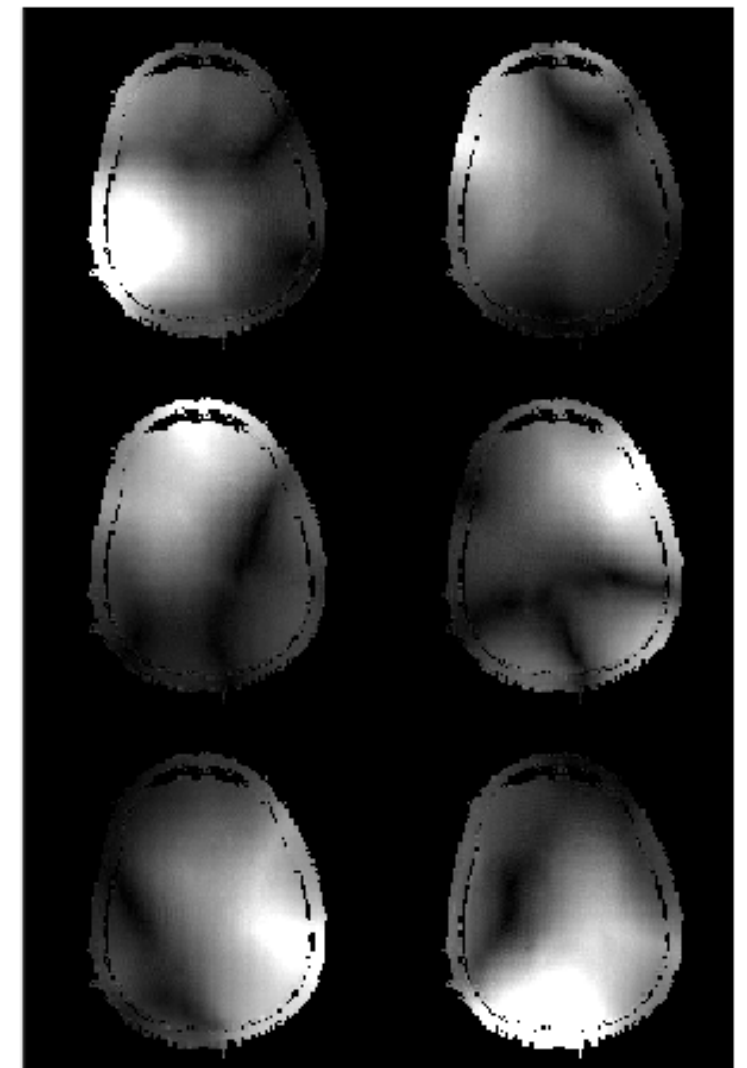
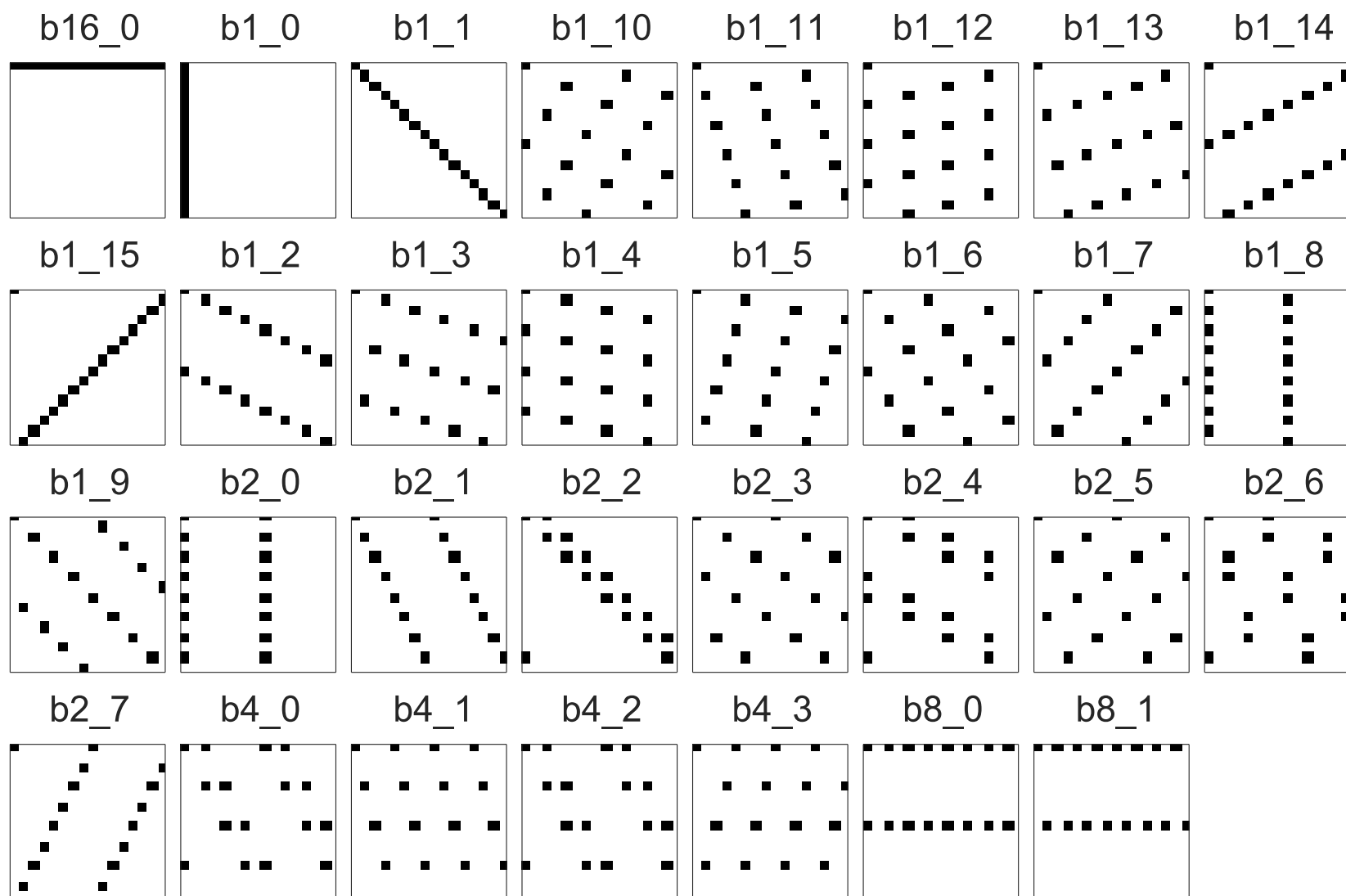
min/max eigenvalues $\rightarrow \lambda_{min} = \sigma_{min}^2$

$$M^H M = \sum_i^{N_{coils}} C_i^H \cdot iFT \cdot k^H \cdot k \cdot FT \cdot C_i$$

- $k^H k$ is really just zeroing of non-sampled locations
- Might affect convergence if SSV is small already
- Can include regularization: $(M^H M + R)\nu = \lambda\nu$
- ARPACK + FFTW

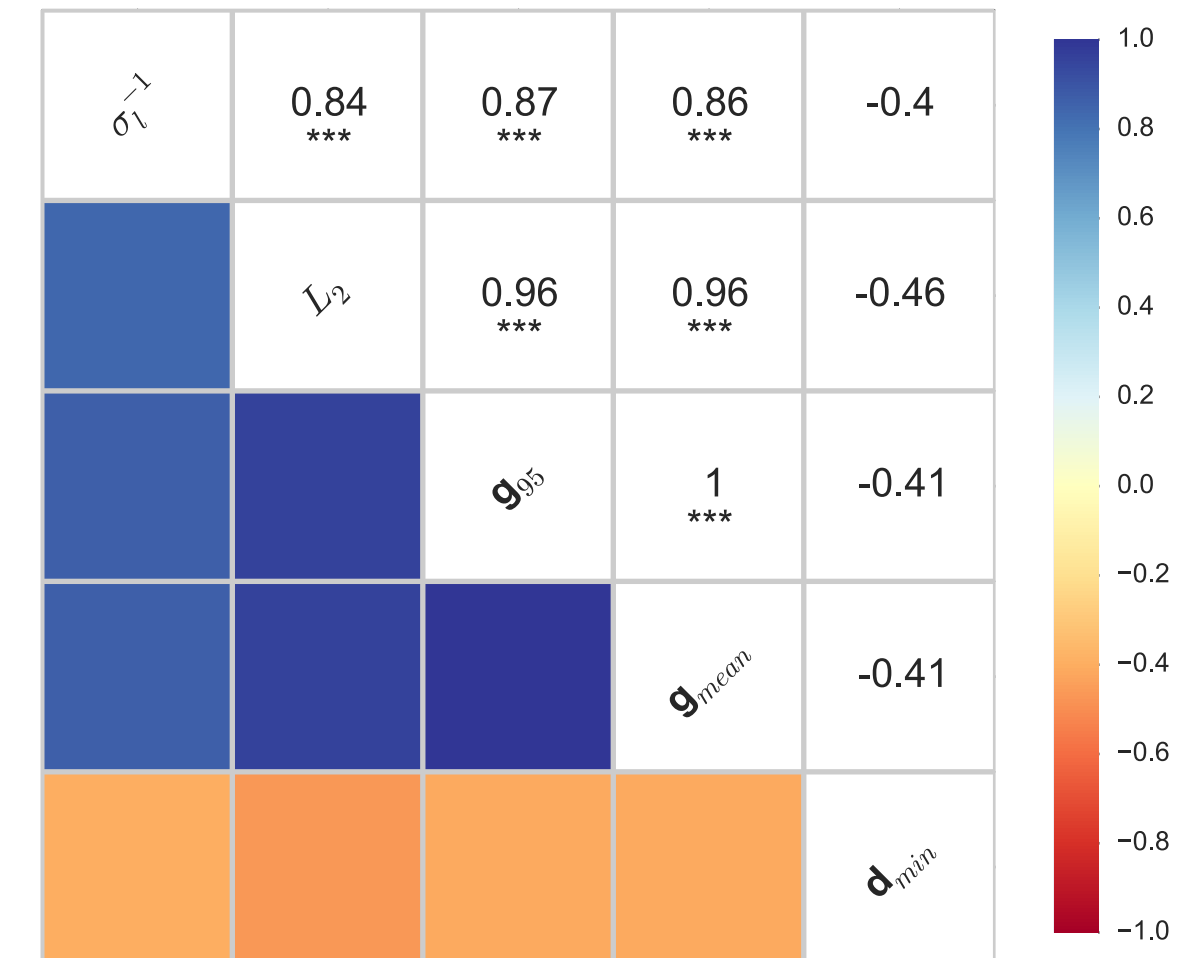
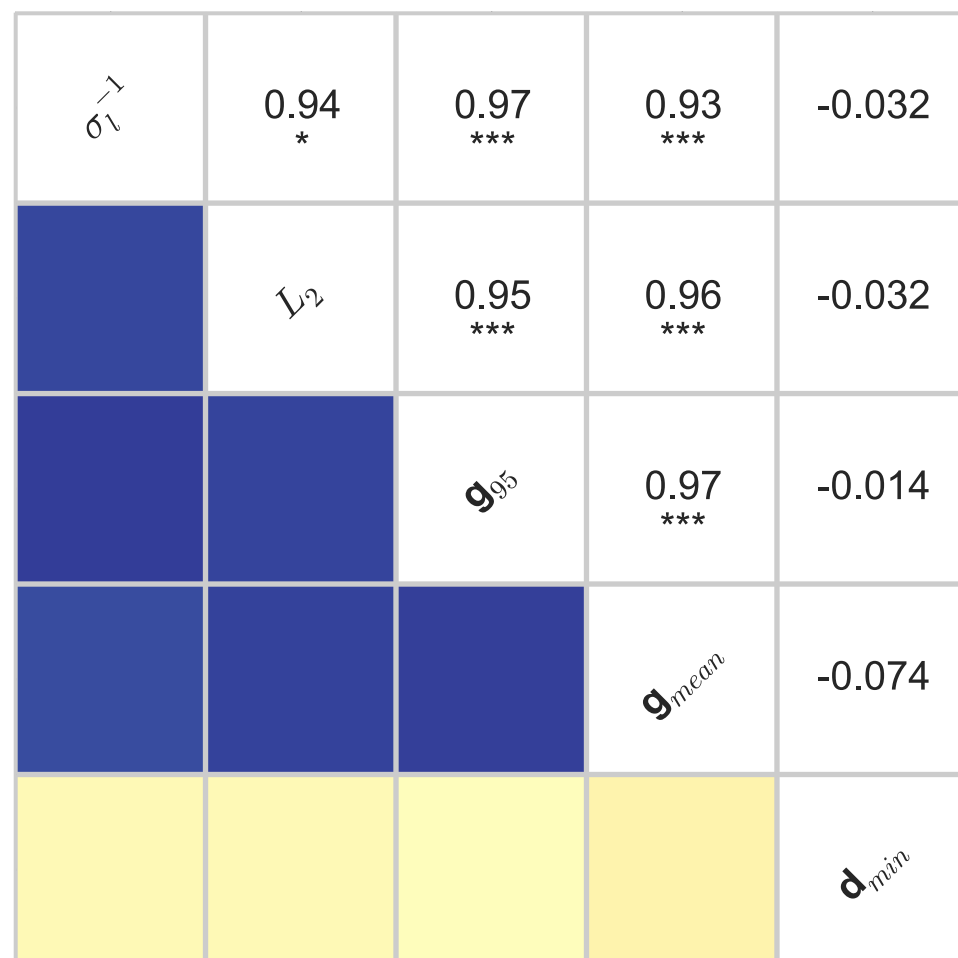
Assessing SSV

- Compute slow G-factor metrics, L2 error, and SSV for a family of well know sampling grids



Good as Gold

- Compute slow G-factor metrics, L2 error, and SSV for a family of well know sampling grids
- Get same ranking in a ***few seconds***

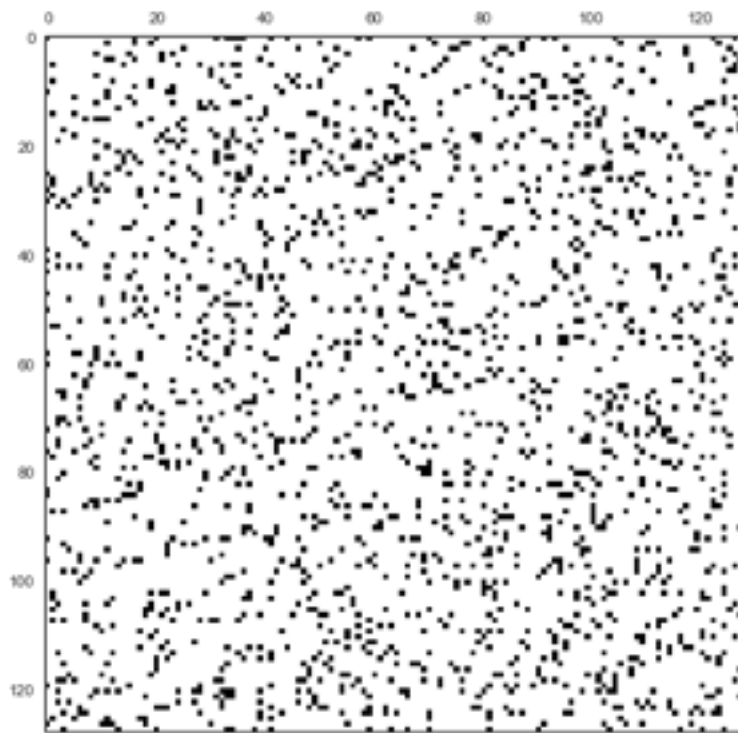


SSV Metric Results

- **SSV** ranks candidate sampling patterns very similarly to gold standard metrics
- orders of magnitude faster ~ seconds-minutes
- Doesn't require image values,
just **receivers** and **pattern**
- **Trade off: loss of spatial information that G-factor and simulation L2 error provides**
- **opens door for optimization!**

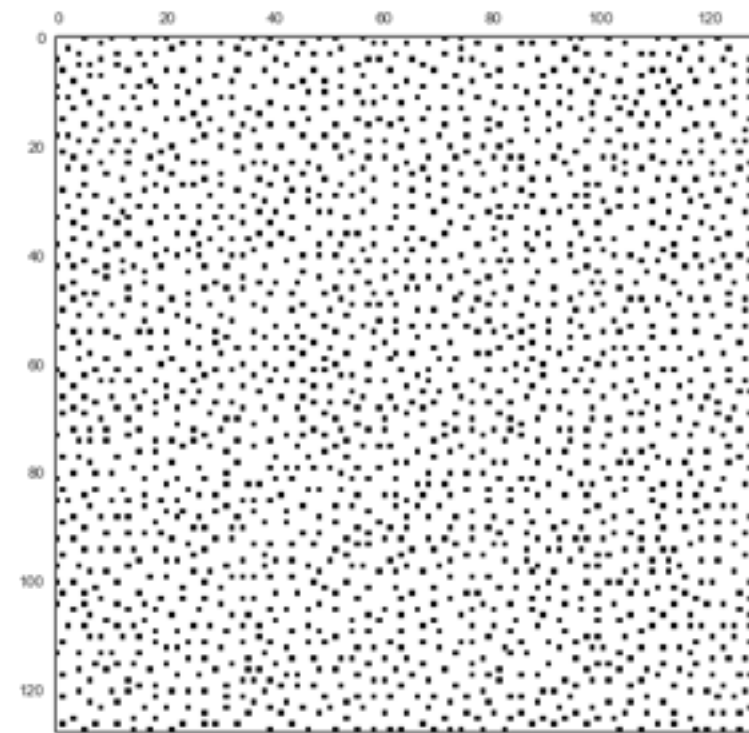
App 1: How Random?

- Combinatorial nightmare
- Random sampling \rightarrow incoherent aliasing “noise”
 \rightarrow CS techniques

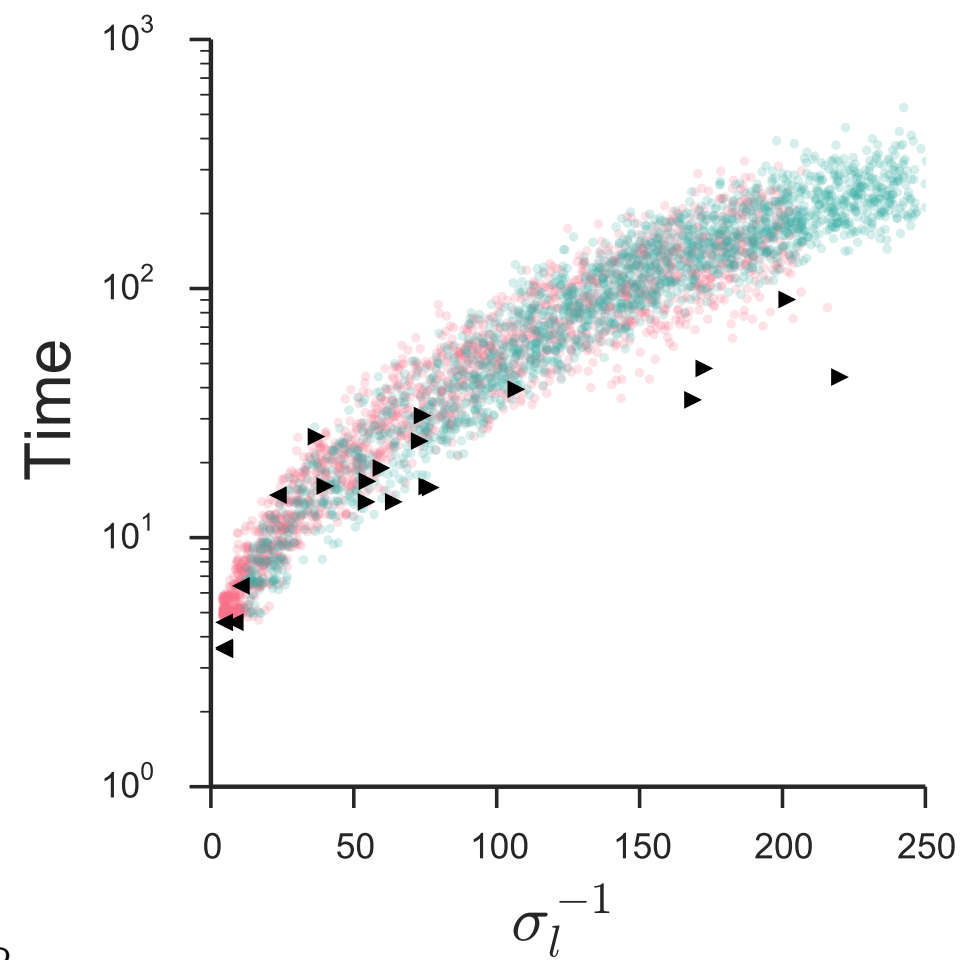
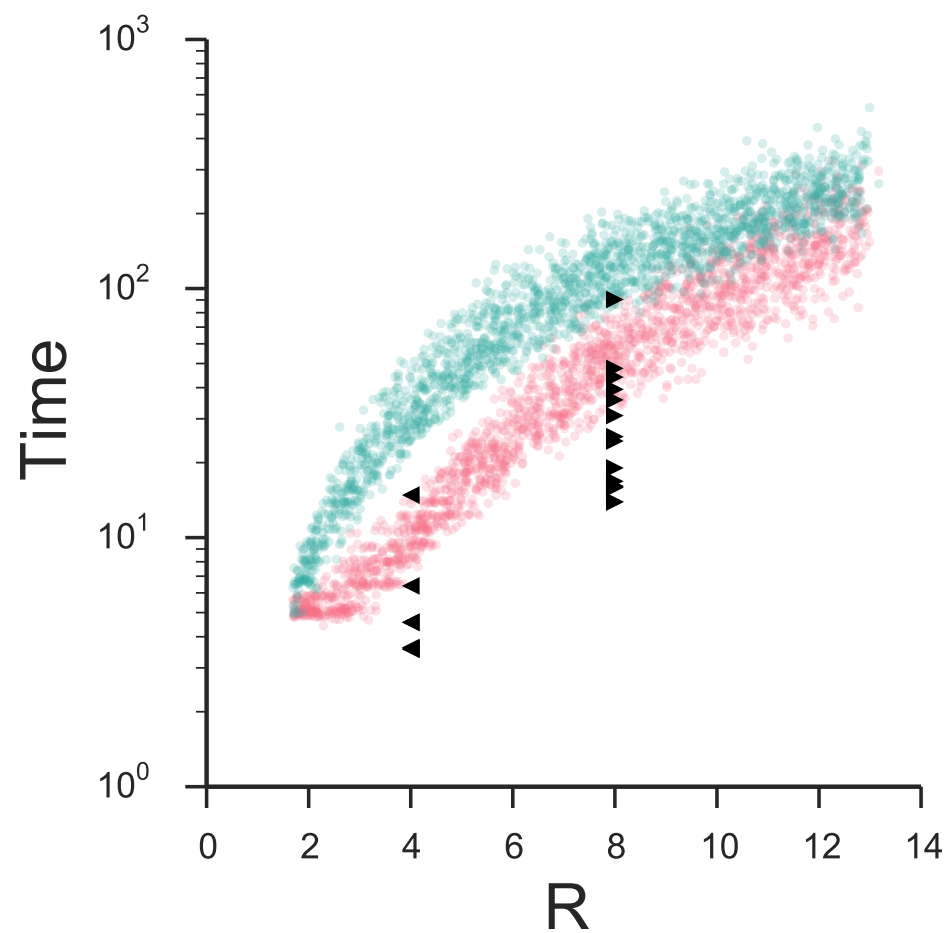
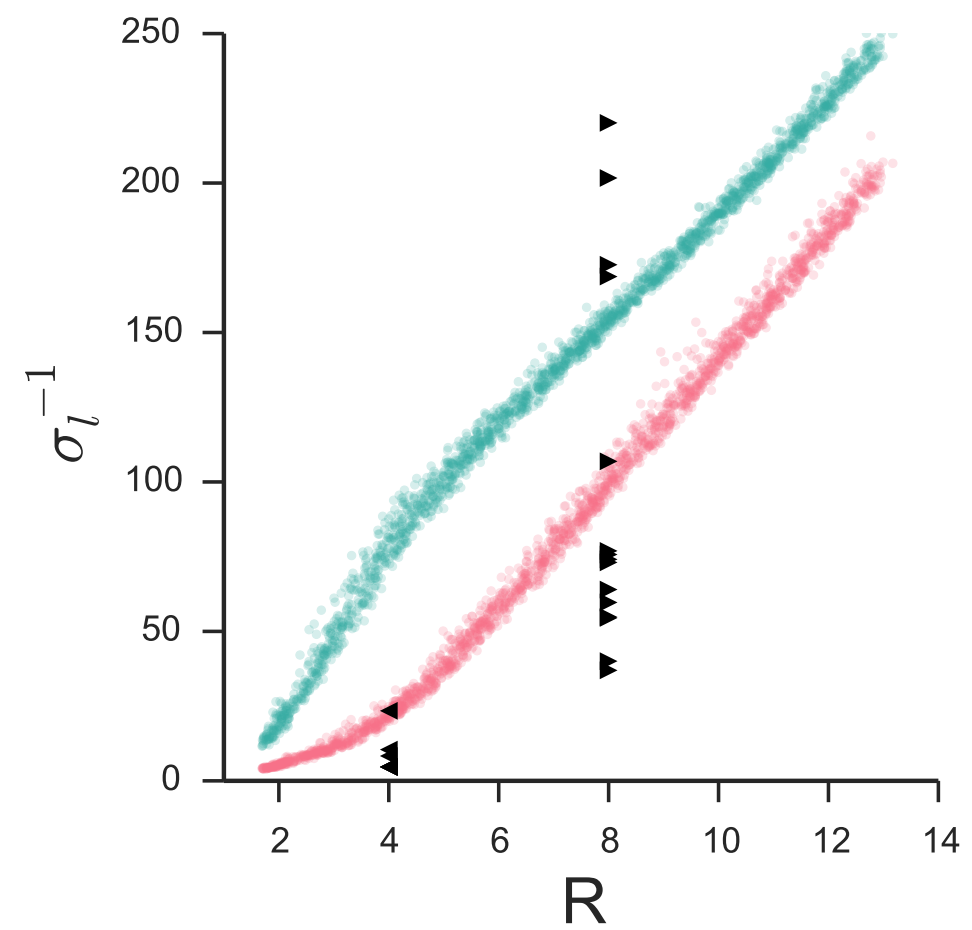
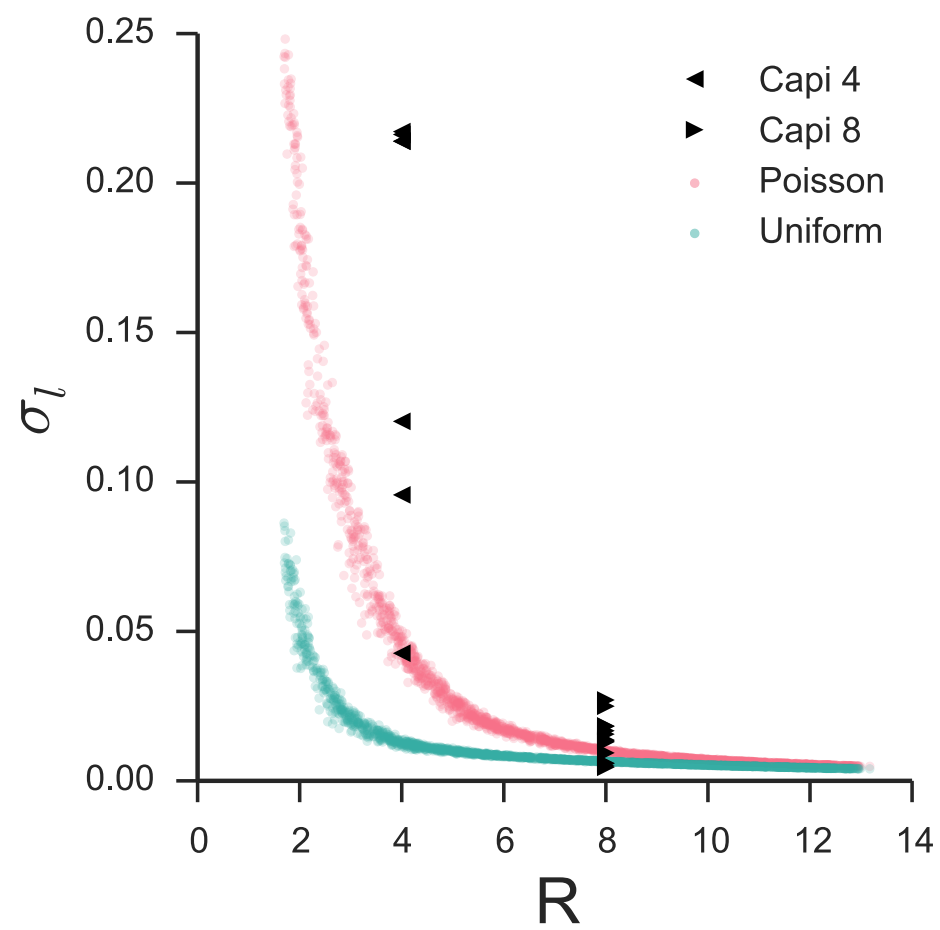


Uniform Random

$R=9$

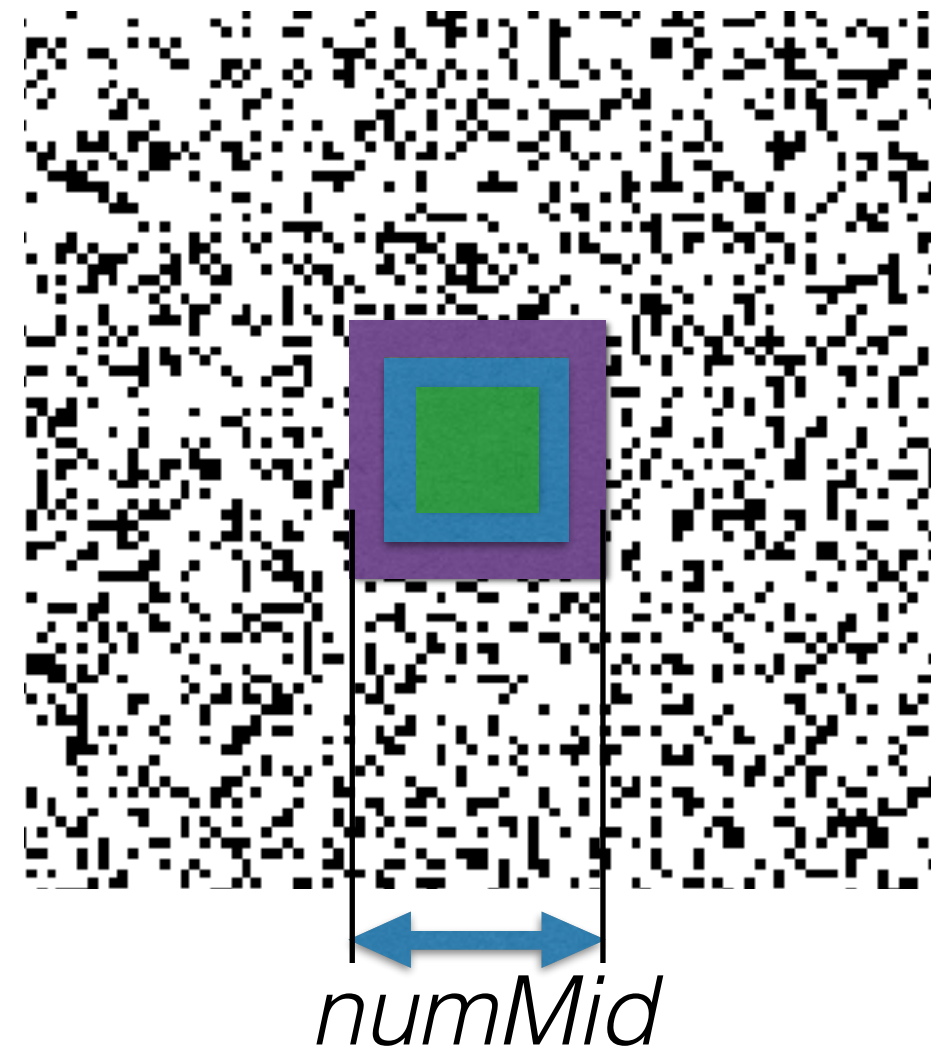


Poisson Disk



App 2: Why low k-space?

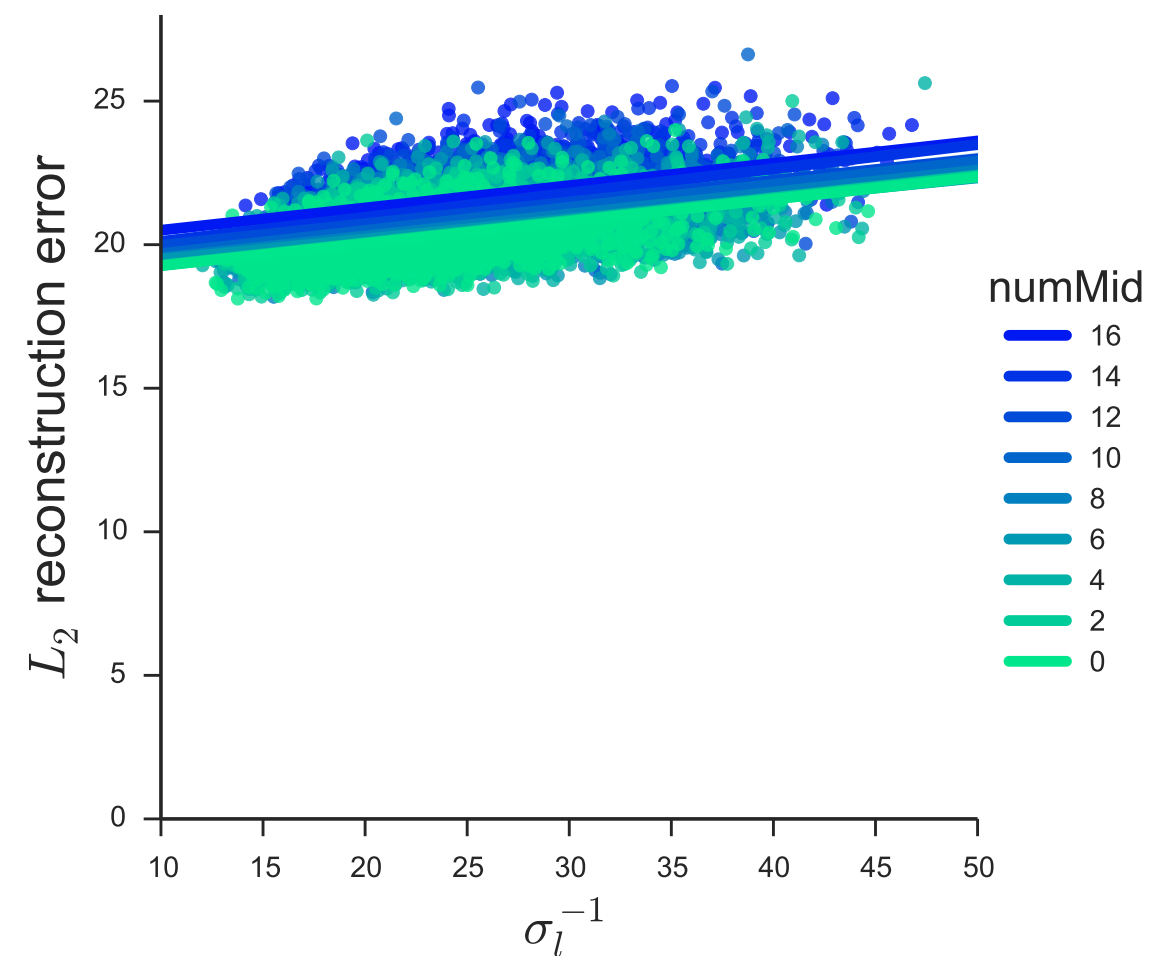
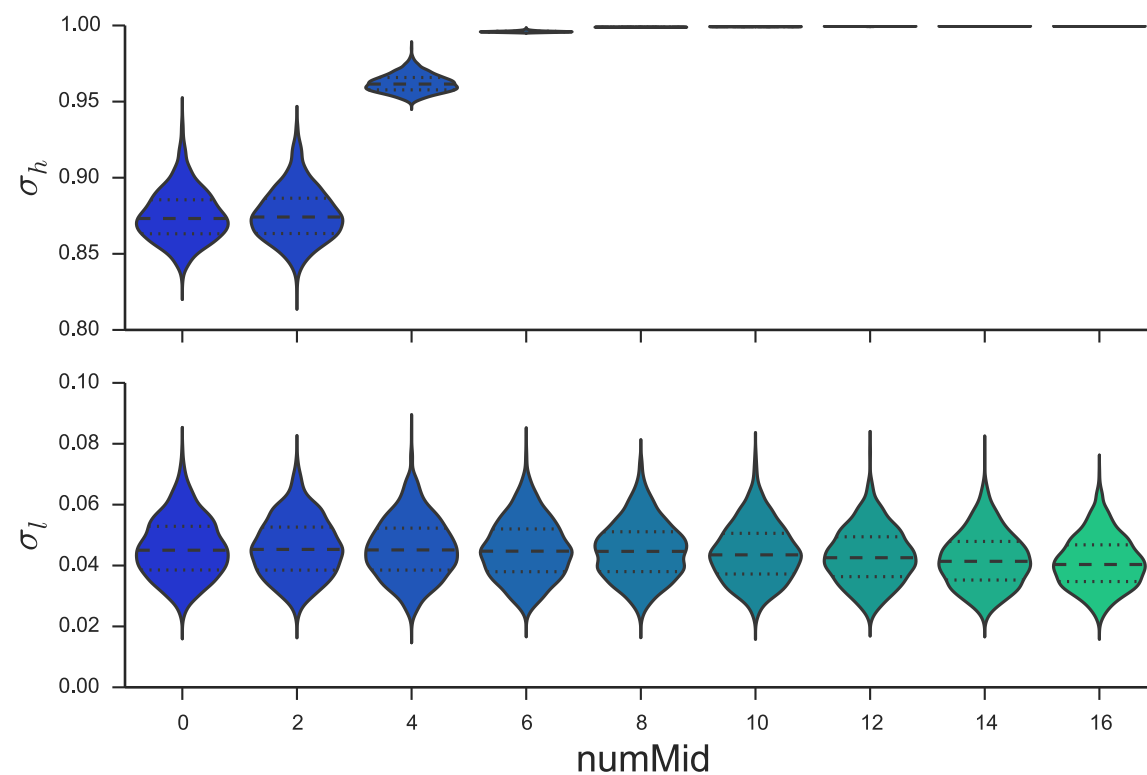
- folklore: always sample centre
- more “energy” at centre of k-space
- low res is easy—> compact coil support, so sample more at high k



- Test: Generate patterns with varying density at centre

Results:

- Centre sampling predominantly affects Largest singular value



SSV

- **SSV** is a *fast* metric for experiment designs
- demonstrated on:
 - 1) Assessing Random patterns
 - 2) Assessing effects of densely sampling k_0
- fast enough to use to optimize experiment design

Coconut Project

- *common sources of error fixable*
 - *transcription of model*
 - *wrong frame of reference*
 - *Heisenbugs*
 - *lots of little bugs left*
- *can effectively estimate SNR*
 - *use this to ensure statistical validity*

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Bonus: 3D sampling

- We said lines of k-space are “free”
- In 2D: we choose which lines to sample
- In 3D, we have a 2D transverse plane of possible locations

