



University of
Zurich ^{UZH}

ETH

Eidgenössische Technische Hochschule Zürich
Swiss Federal Institute of Technology Zurich

Accelerating Dynamic MRI

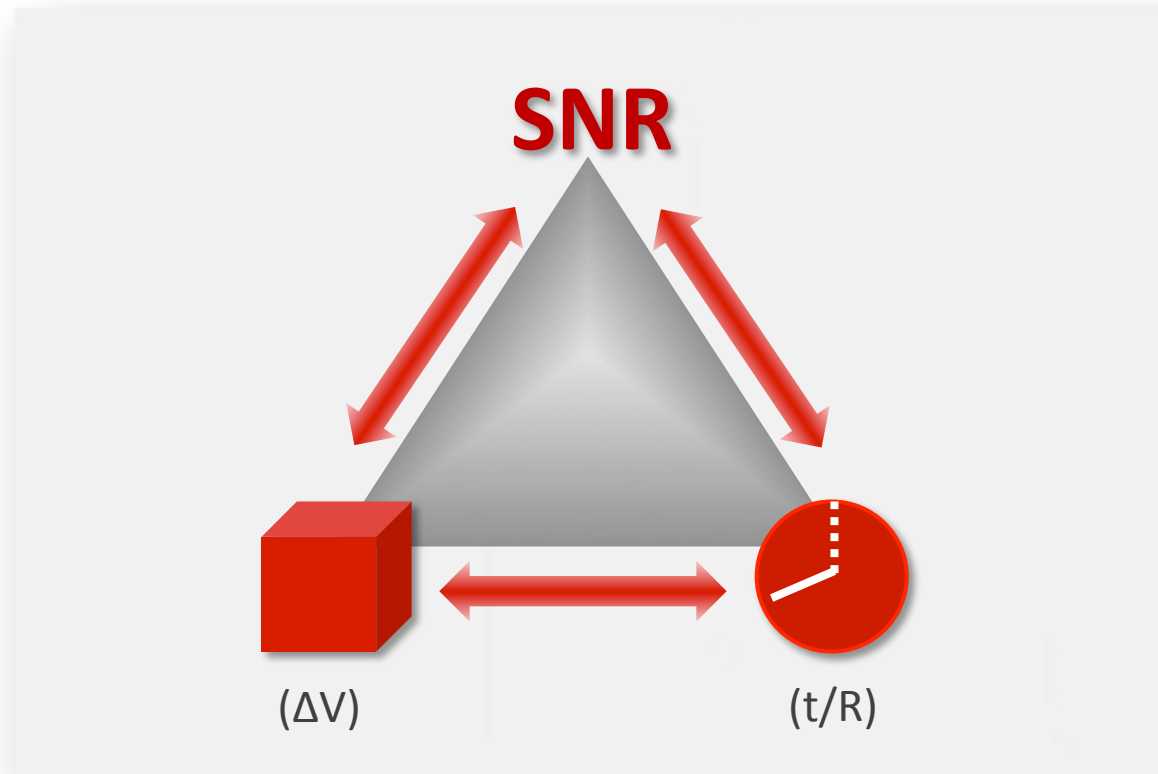
Uniform and random undersampling

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Institute for Biomedical Engineering, University and ETH Zurich
Biomedical Engineering and Imaging Sciences, King's College London



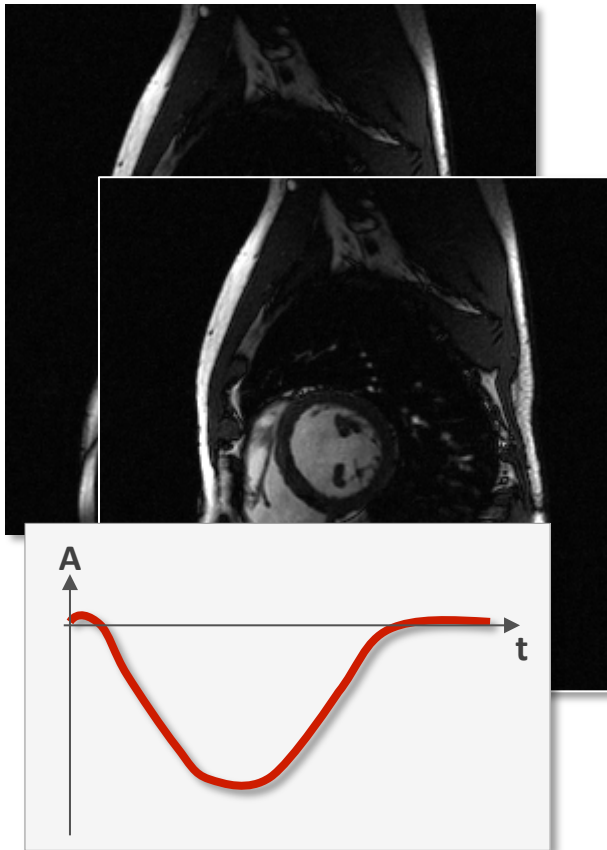
SNR – Speed



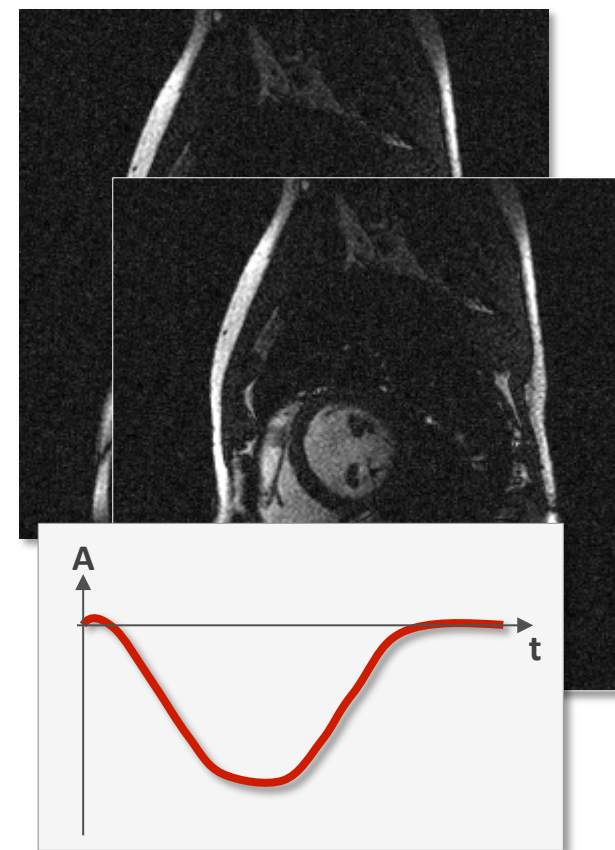
$$\text{SNR} \propto \Delta V \sqrt{t} / \sqrt{R}$$

SNR – Speed

SNR = 32



SNR = 8



Reconstruction

Encoding:

$$\vec{d} = E \vec{m} + \vec{\eta}$$

Decoding:

$$\vec{i} = F \vec{d}$$

| | MR Signal | Noise |
|-------|---|------------------------------------|
| Data | Encoding matrix E | Ψ |
| Image | Depiction FE ↓ Identity | $F\Psi F^H$ ↓ Minimum |

Reconstruction formulas

Constrain

$$FE = \text{Id}$$

Solutions

Pseudo-inverse:

$$F = (E^H E)^{-1} E^H$$

Optimum SNR inverse:

$$F = (E^H \Psi^{-1} E)^{-1} E^H \Psi^{-1}$$

Regularized solution:

$$R < N_c$$

$$F = (E^H \Psi^{-1} E + \lambda \Theta^{-1})^{-1} E^H \Psi^{-1}$$

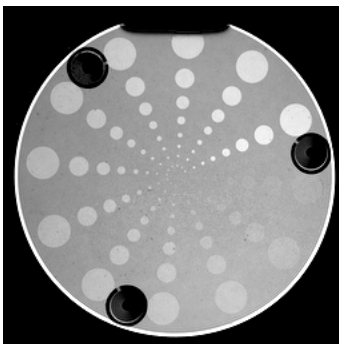
$$R > N_c$$

$$F = \Theta E^H (E \Theta E^H + \lambda \Psi)^{-1}$$

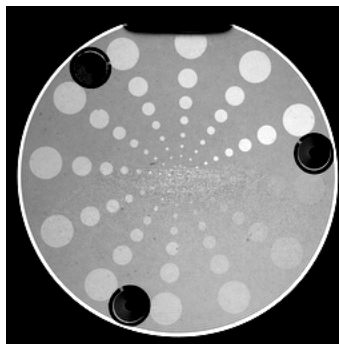
Parallel imaging – performance

Geometry factor

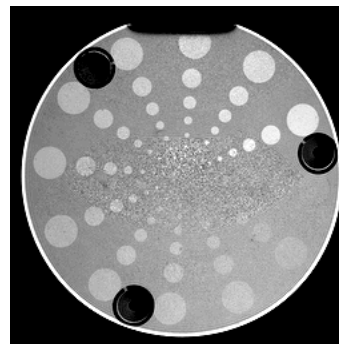
R = 2.0



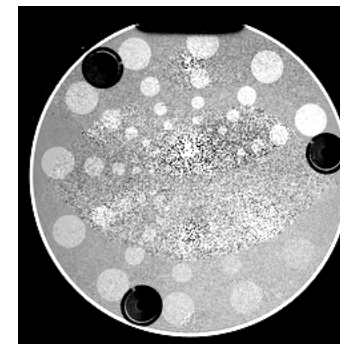
R = 2.4



R = 3.0



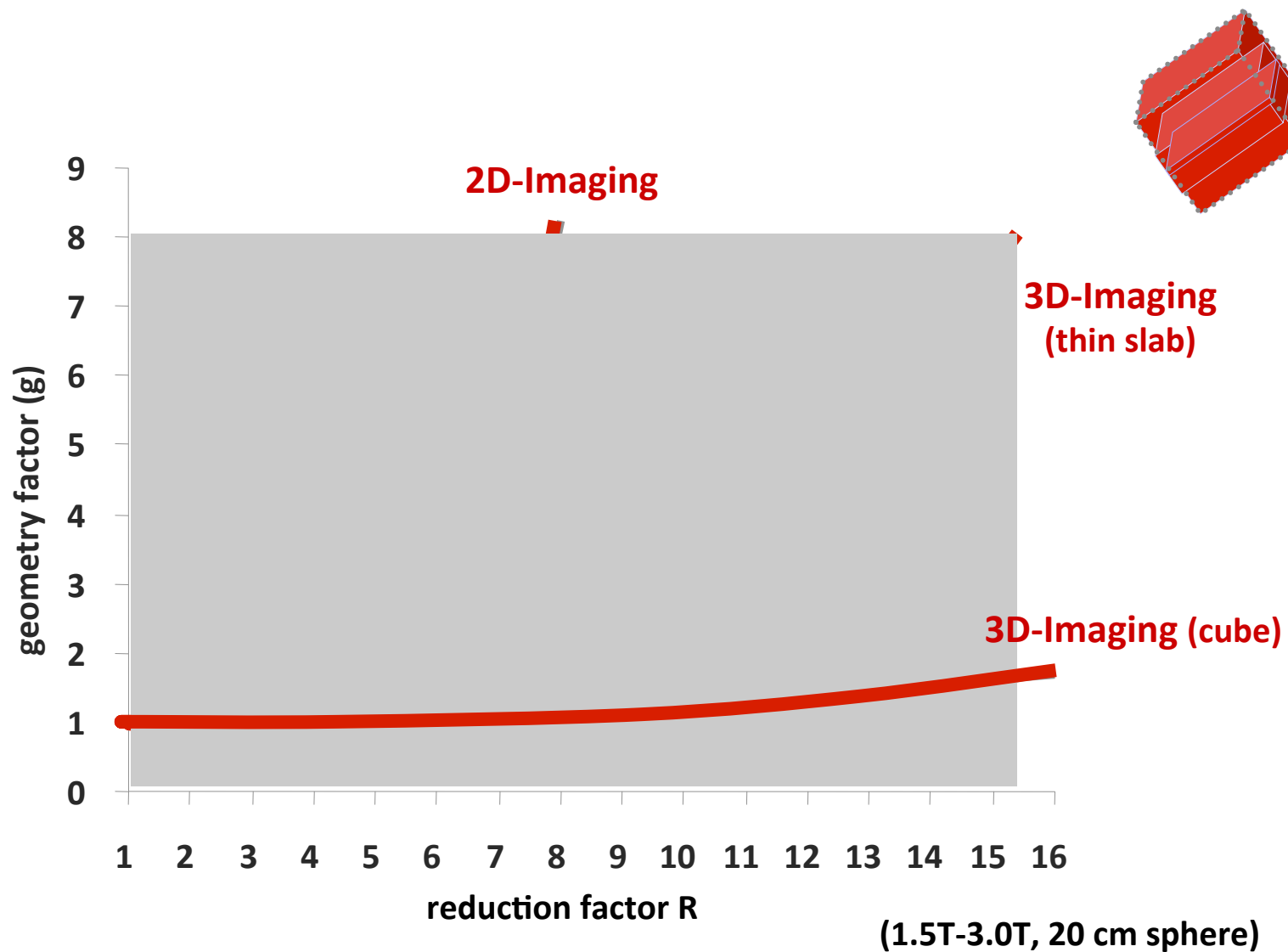
R = 4.0



$$\text{SNR}^{\text{SENSE}} = \frac{\text{SNR}^{\text{Full}}}{\sqrt{R \cdot g(\mathbf{x})}}$$

$$g(\mathbf{x}) = \sqrt{\left(\mathbf{E}^H \mathbf{E} \right)_{i,j} \left(\left(\mathbf{E}^H \mathbf{E} \right)^{-1} \right)_{i,j}} \geq 1$$

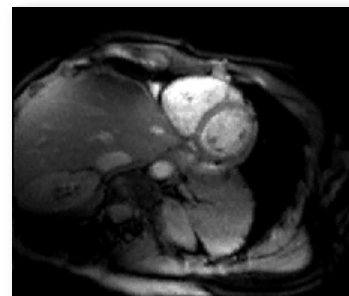
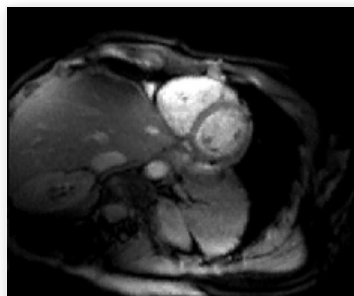
Parallel imaging – ultimate performance



Information redundancy

How much information is redundant?

- ▶ Transform data to a **sparser** coefficient space
- ▶ Find **model** that suitably **links** coefficients
- ▶ Retain **key** coefficients



Original: 16 bits/pixel

16 bits/pixel

Entropy rate¹⁾: **3.3** bits/pixel

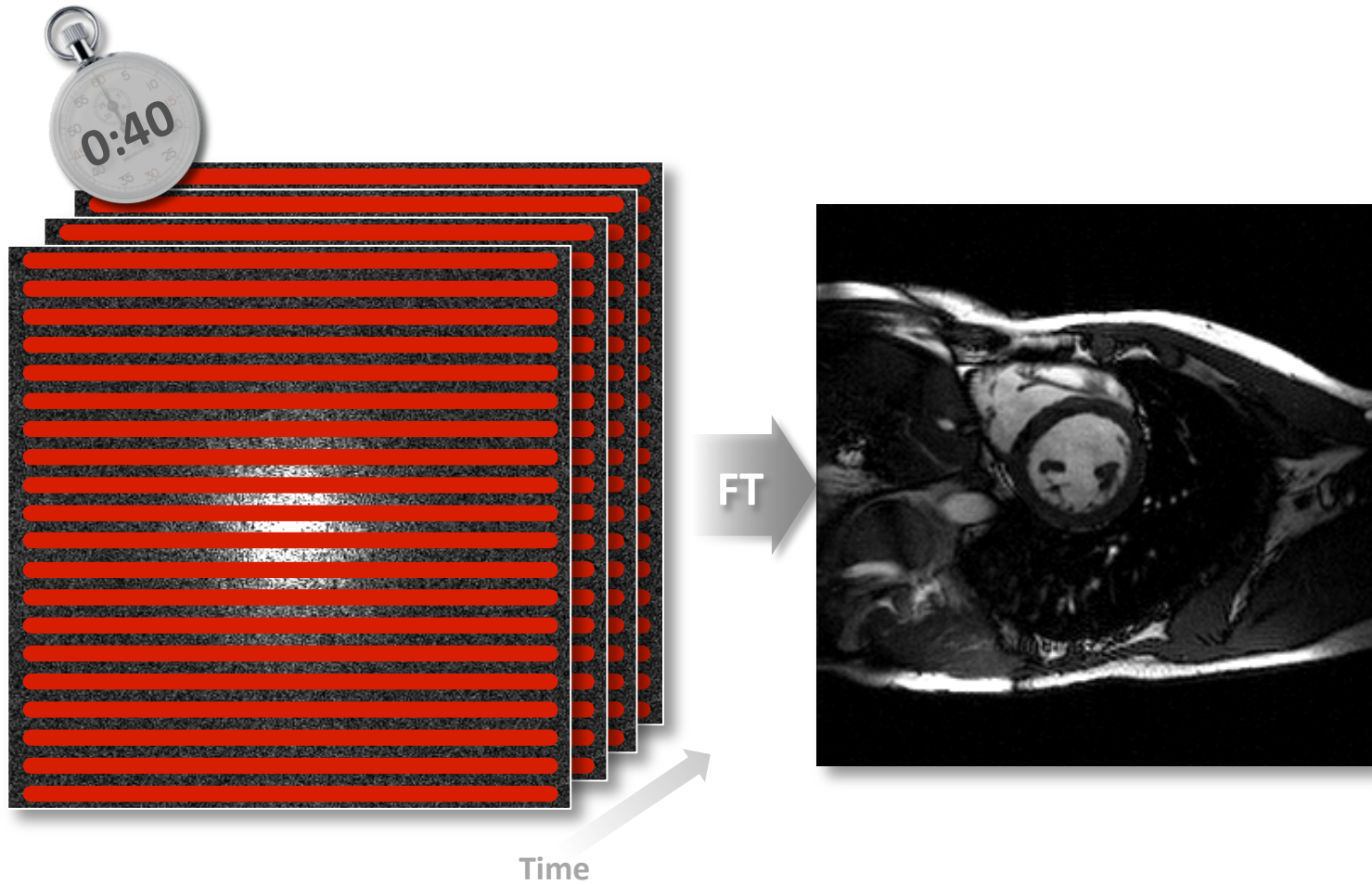
1.0 bits/pixel

“theoretical” R: **4.8**

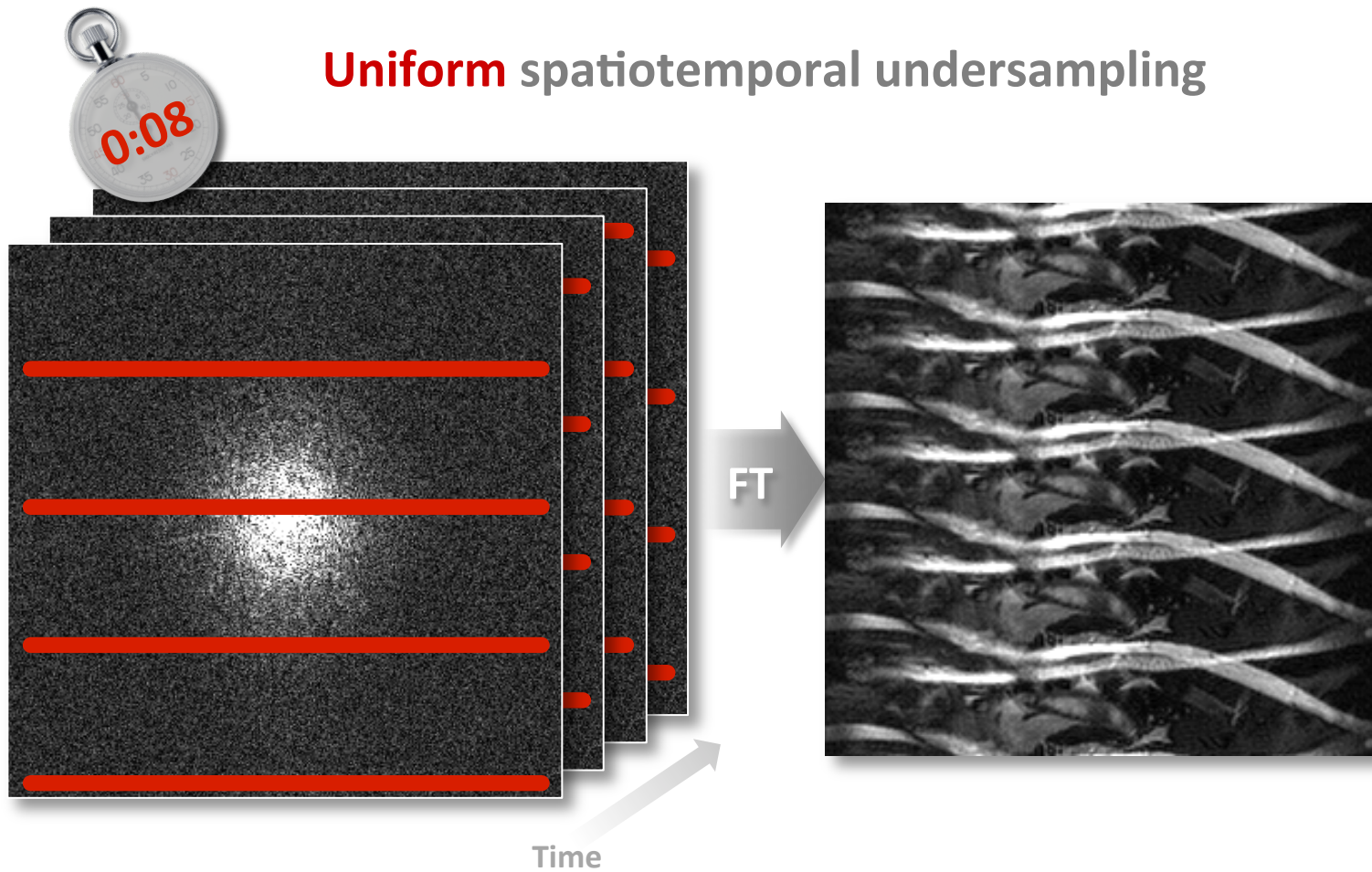
16.0

¹⁾ Cosine Transform, Huffman encoding

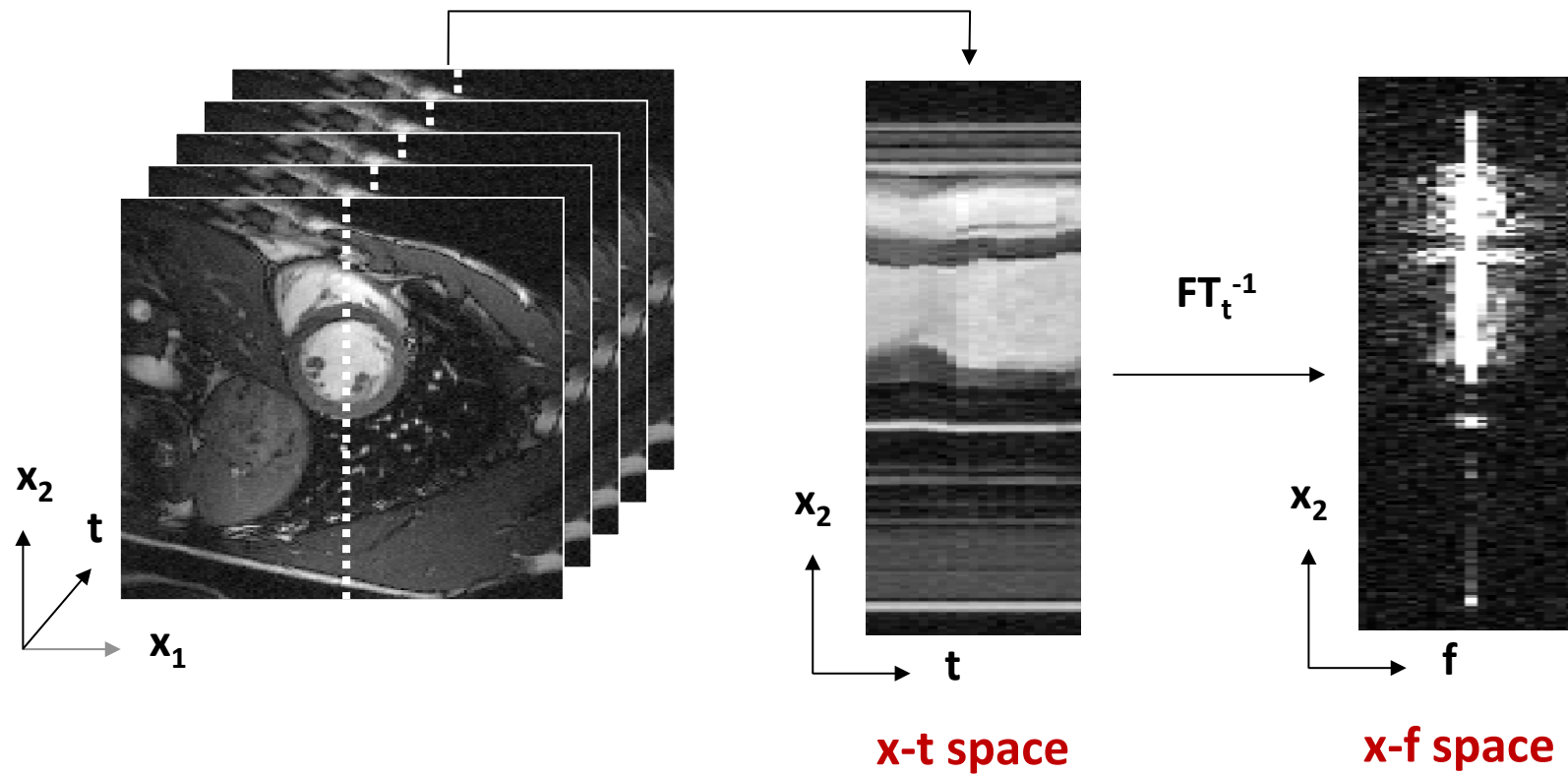
k-t undersampling



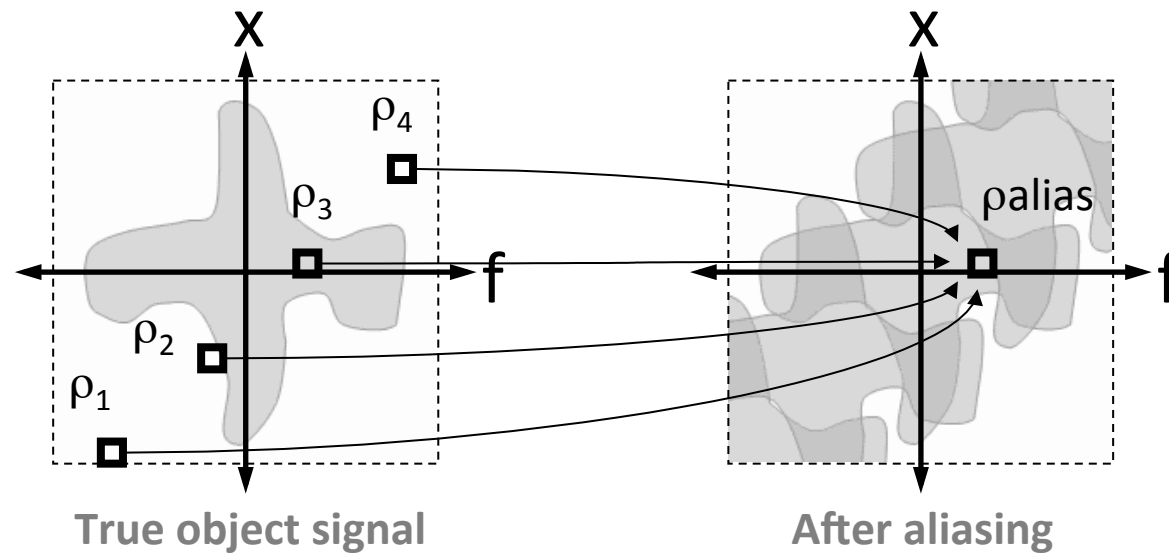
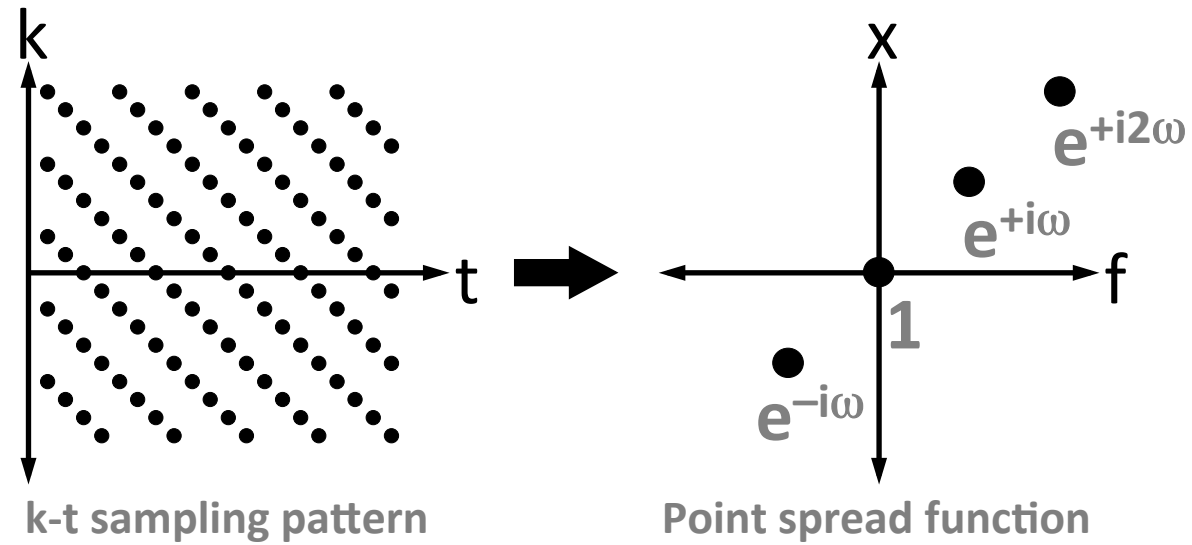
Uniform k-t undersampling



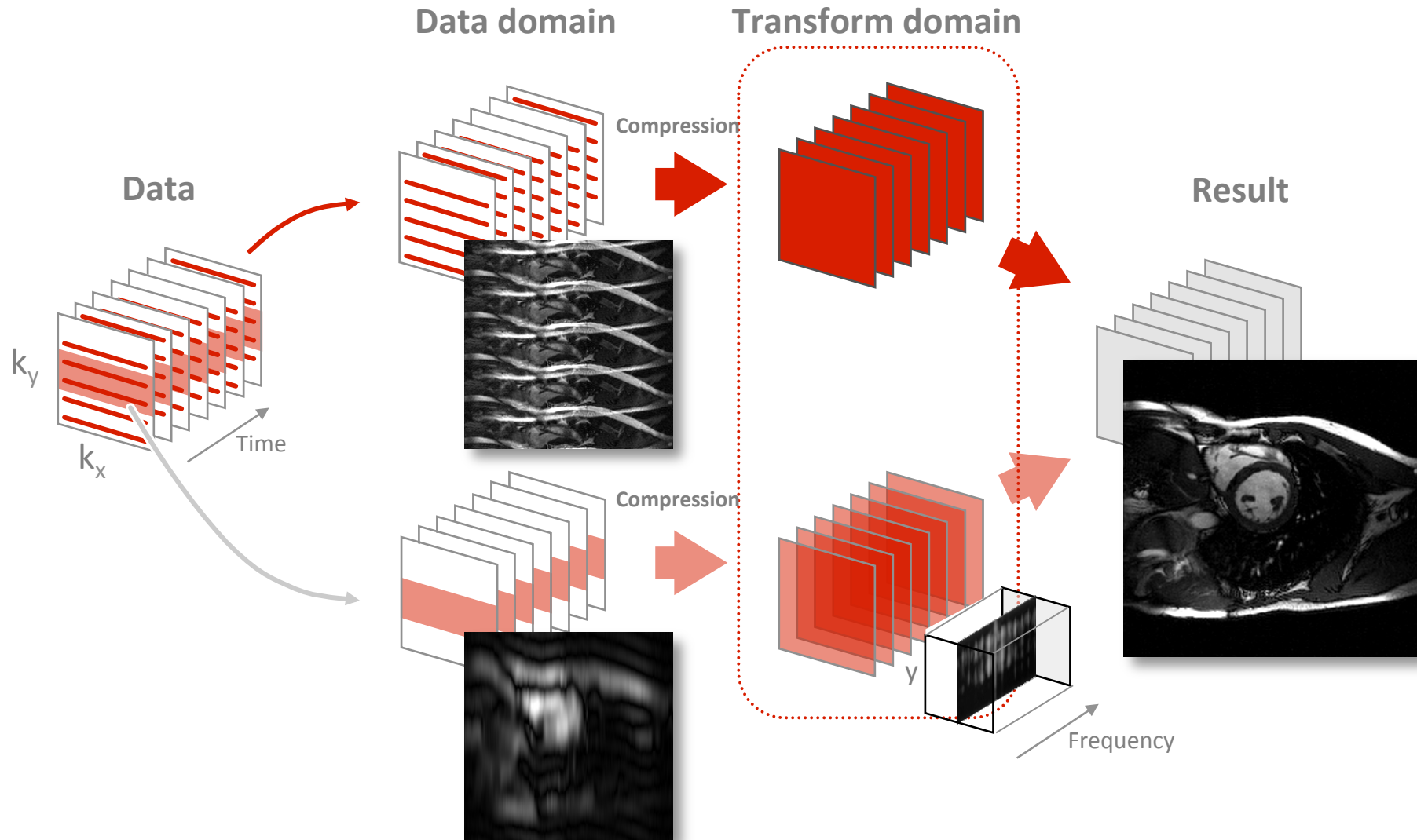
x-t \rightarrow x-f space



x-t \rightarrow x-f space

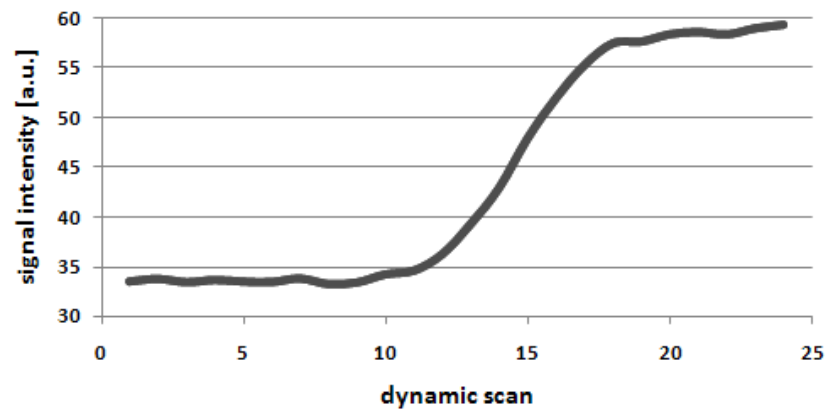
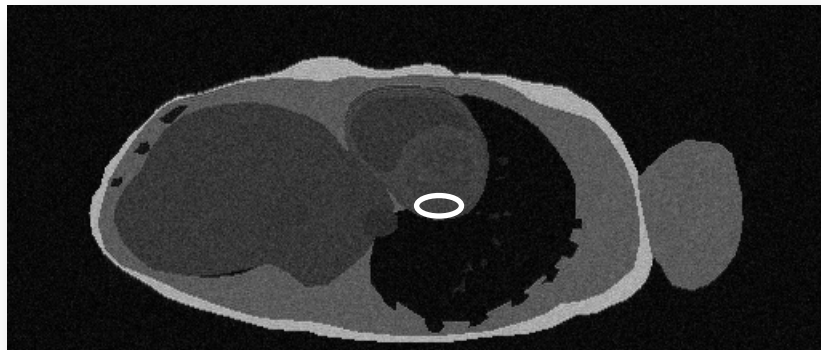


k-t BLAST / k-t SENSE

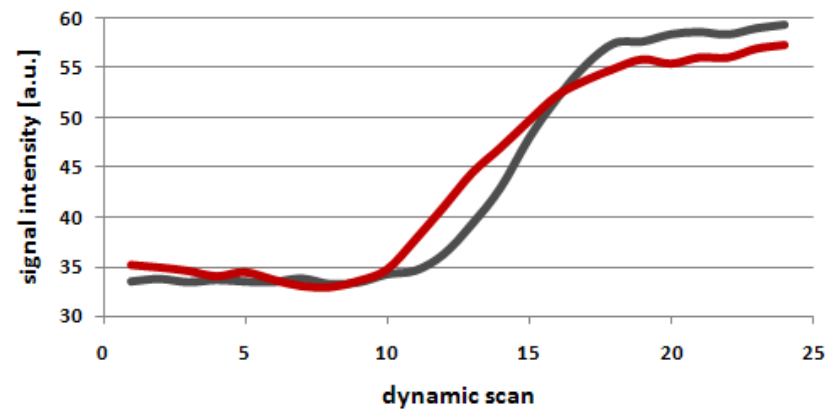
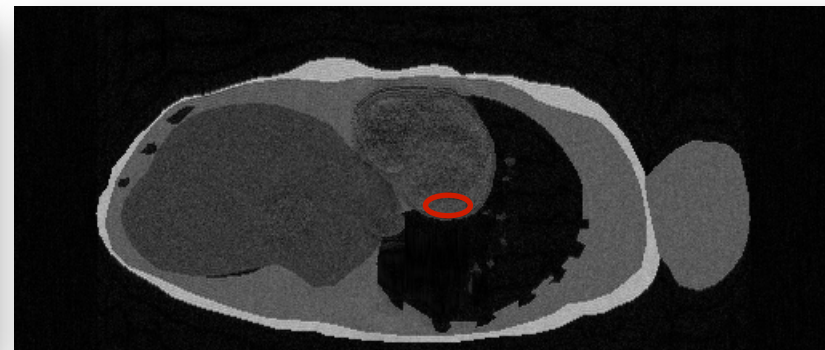


k-t SENSE – Temporal fidelity

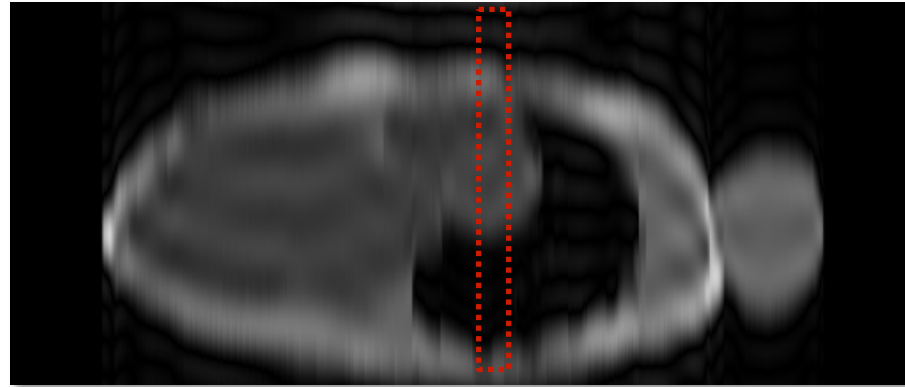
Reference



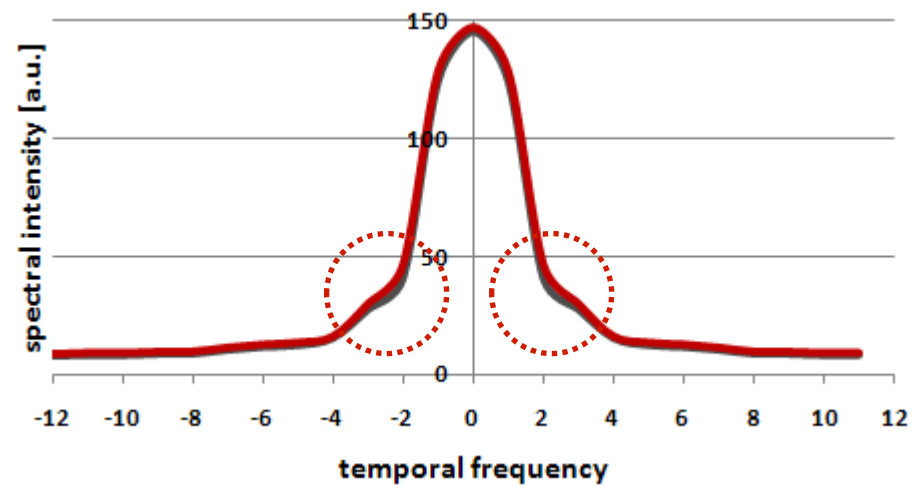
12x k-t SENSE



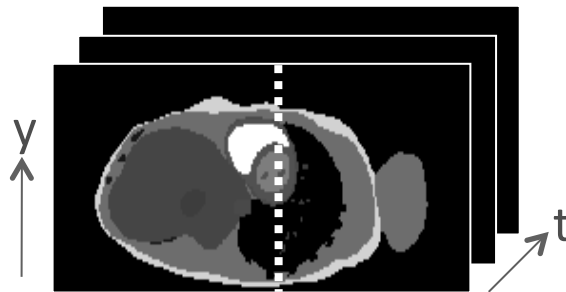
k-t SENSE – Training data



$$i = M^2 E^H (EM^2 E^H + \sigma^2)^+ m_{\text{alias}}$$

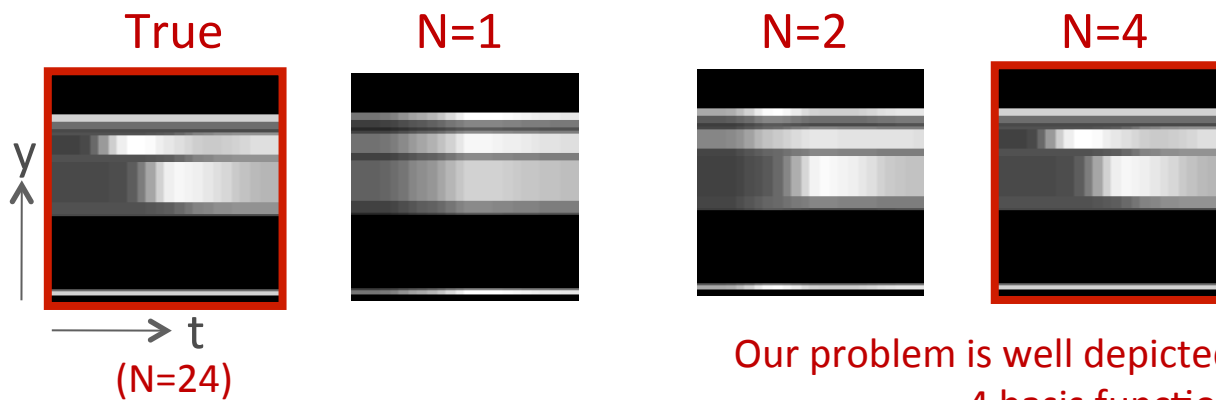


Reconsidering temporal signals

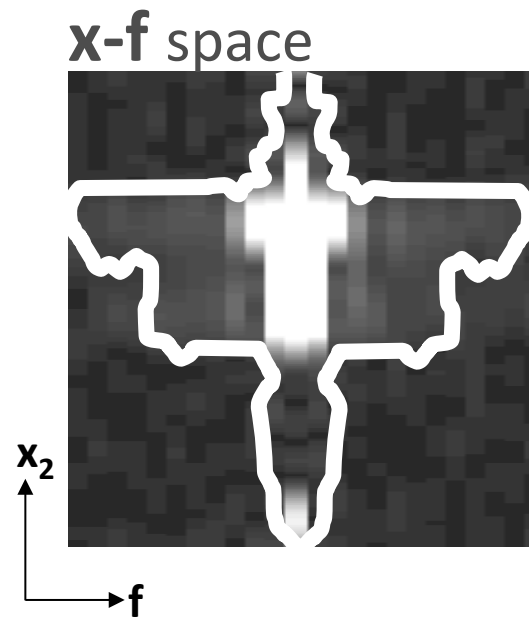
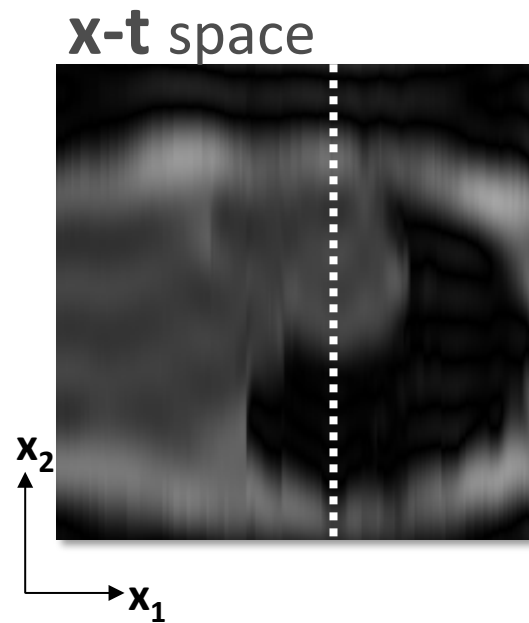


Spatial Weighting Basis functions

$$m(y,t) = UEV^H = \underbrace{W}_{\text{Spatial Weighting}} \underbrace{B}_{\text{Basis functions}} = \sum_{n=1}^N W(y,n)B(n,t)$$

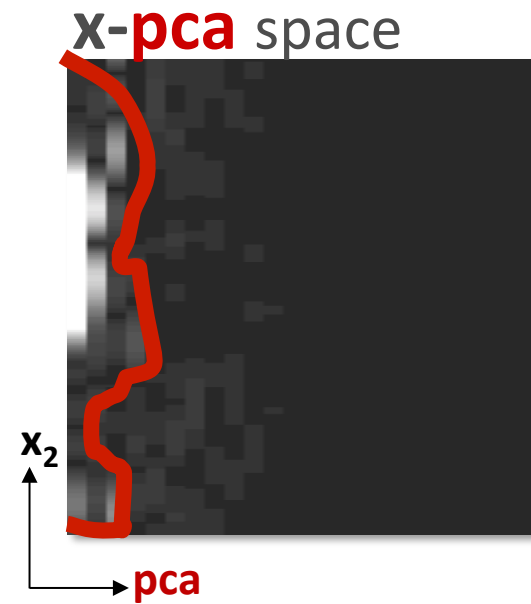


PCA space



48 frames

48 equations
in 48x12 unknowns

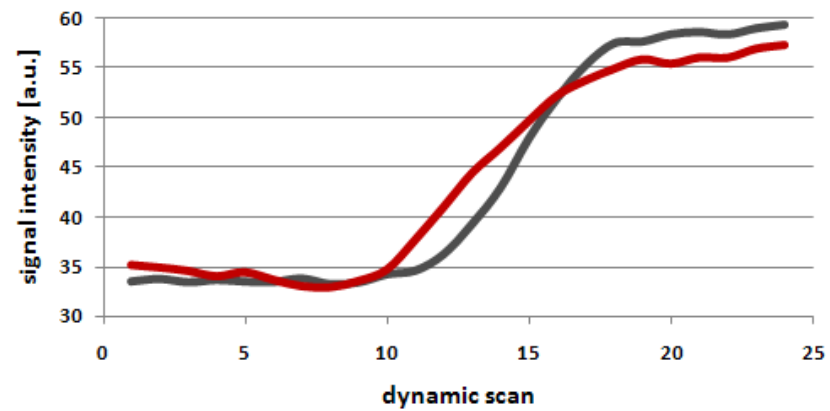
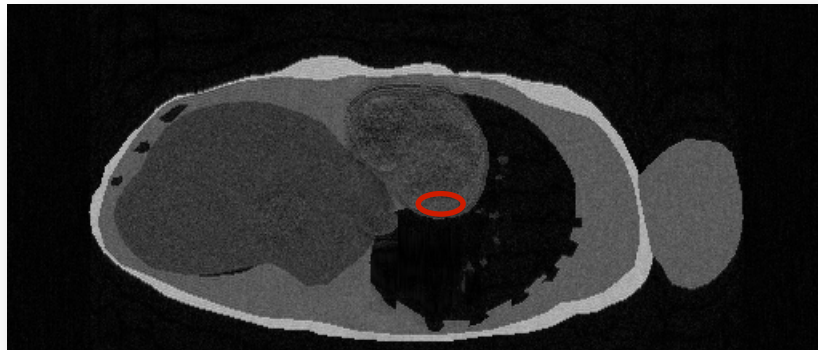


4 weights determine 48 frames

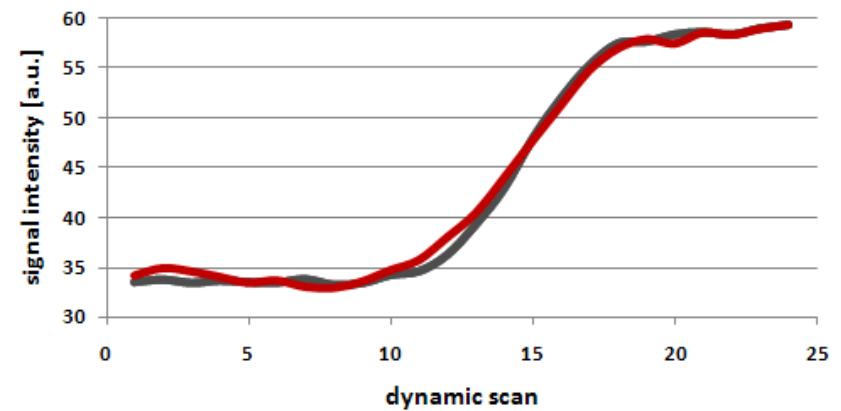
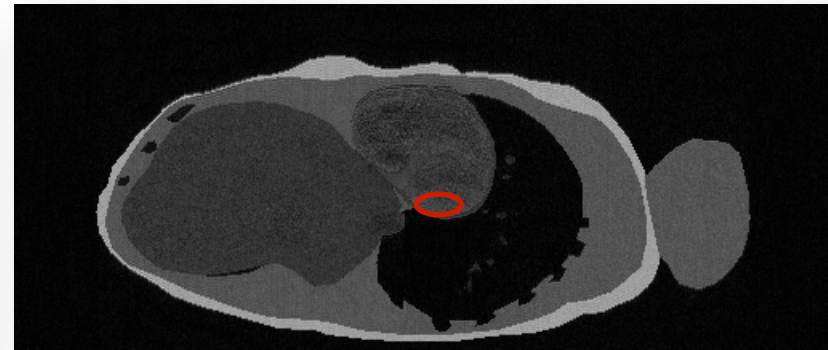
48 equations
in 4x12 unknowns

k-t SENSE vs k-t PCA

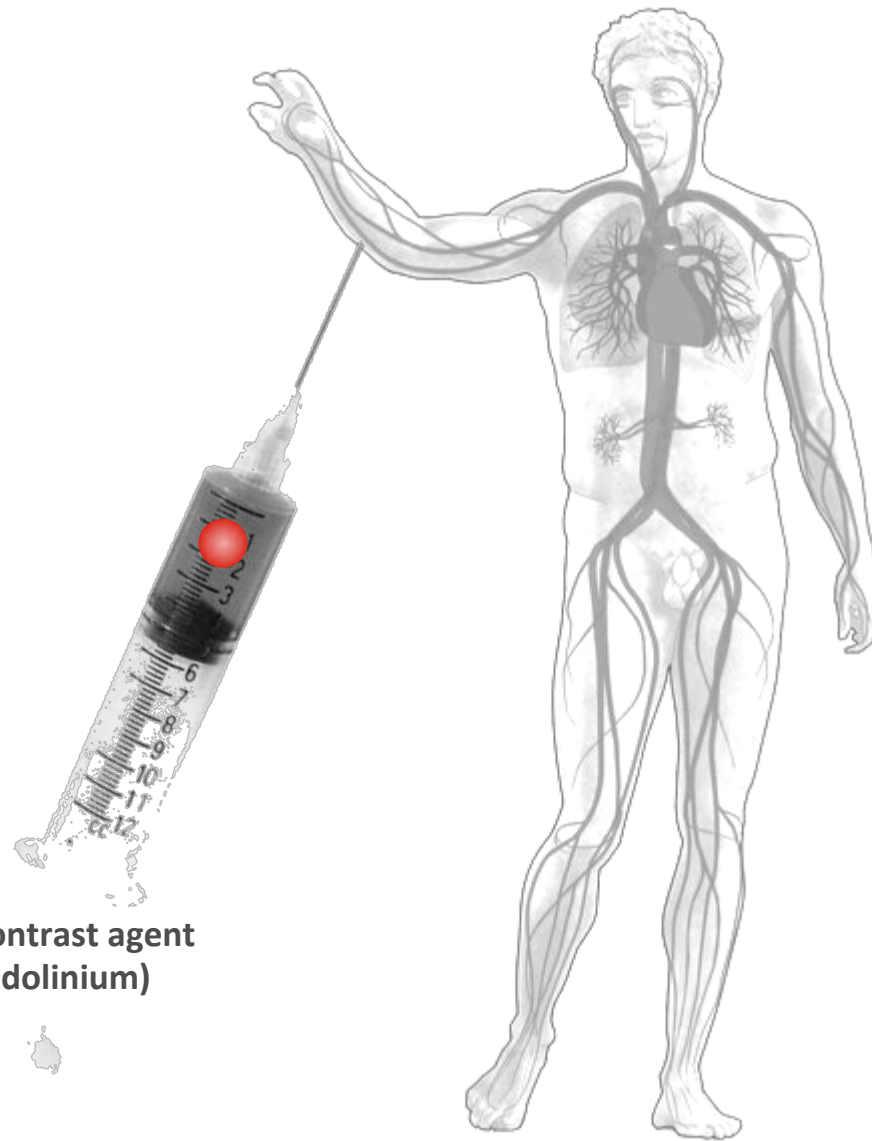
12x *k-t* SENSE



12x *k-t* PCA



Perfusion imaging

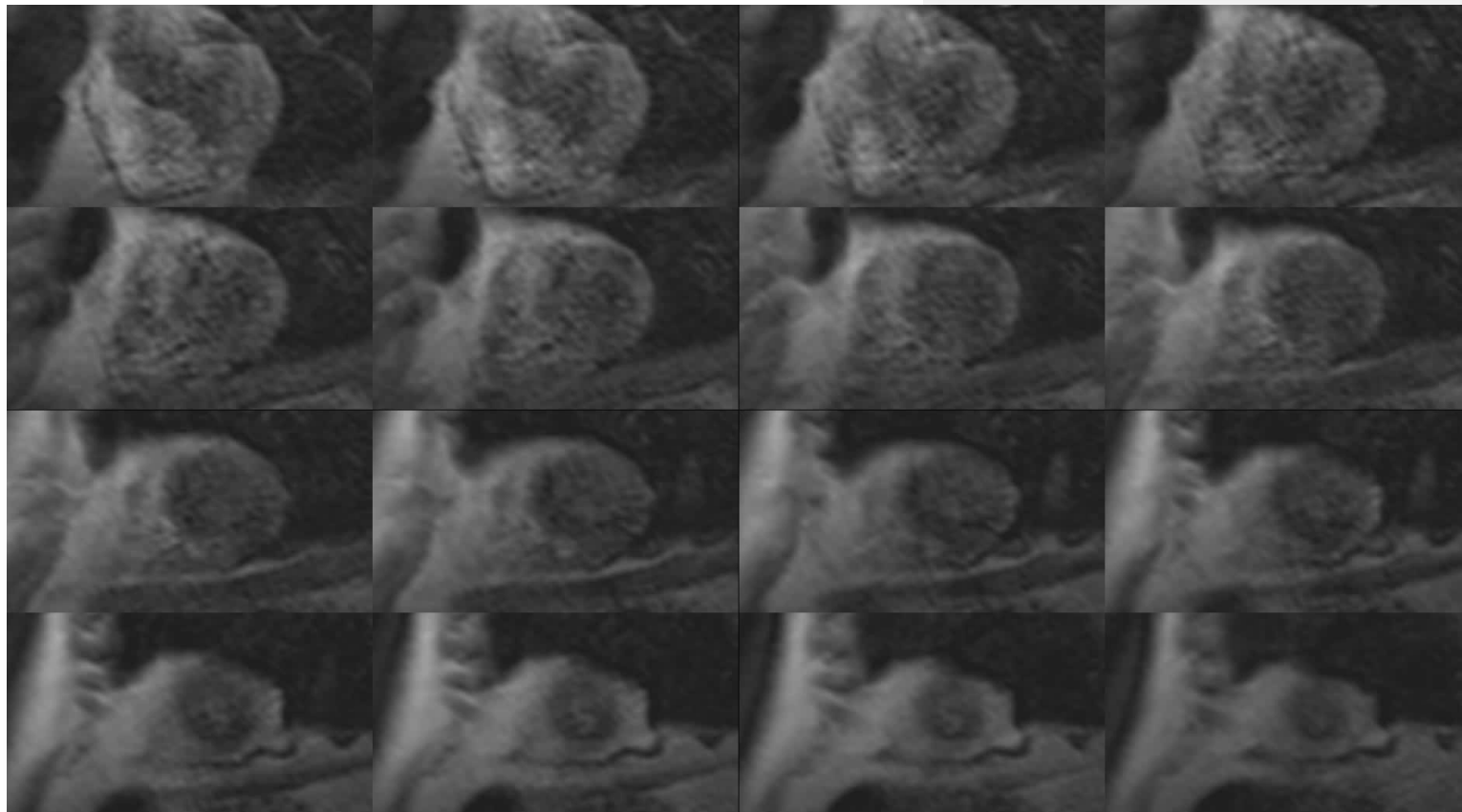


**MR Contrast agent
(Gadolinium)**

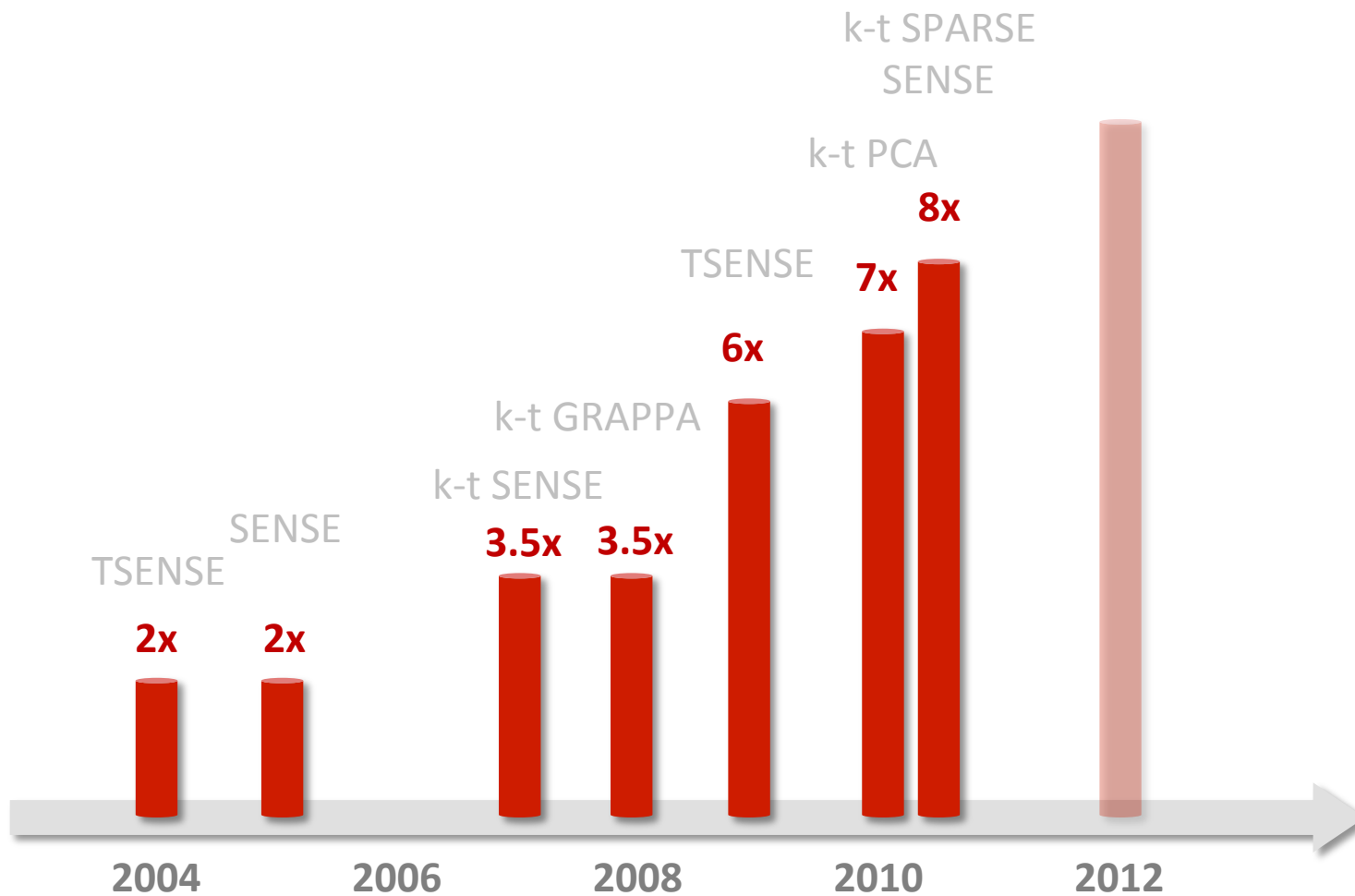
3D Perfusion imaging



10x 3D k-t PCA (2.2 x 2.2 mm²)



Speed-up – Perfusion imaging



Kellman P et al. MRM 2004

Plein S et al. MRM 2007

Nayak KS et al. JCMR 2008

Otazo R et al. MRM 2010

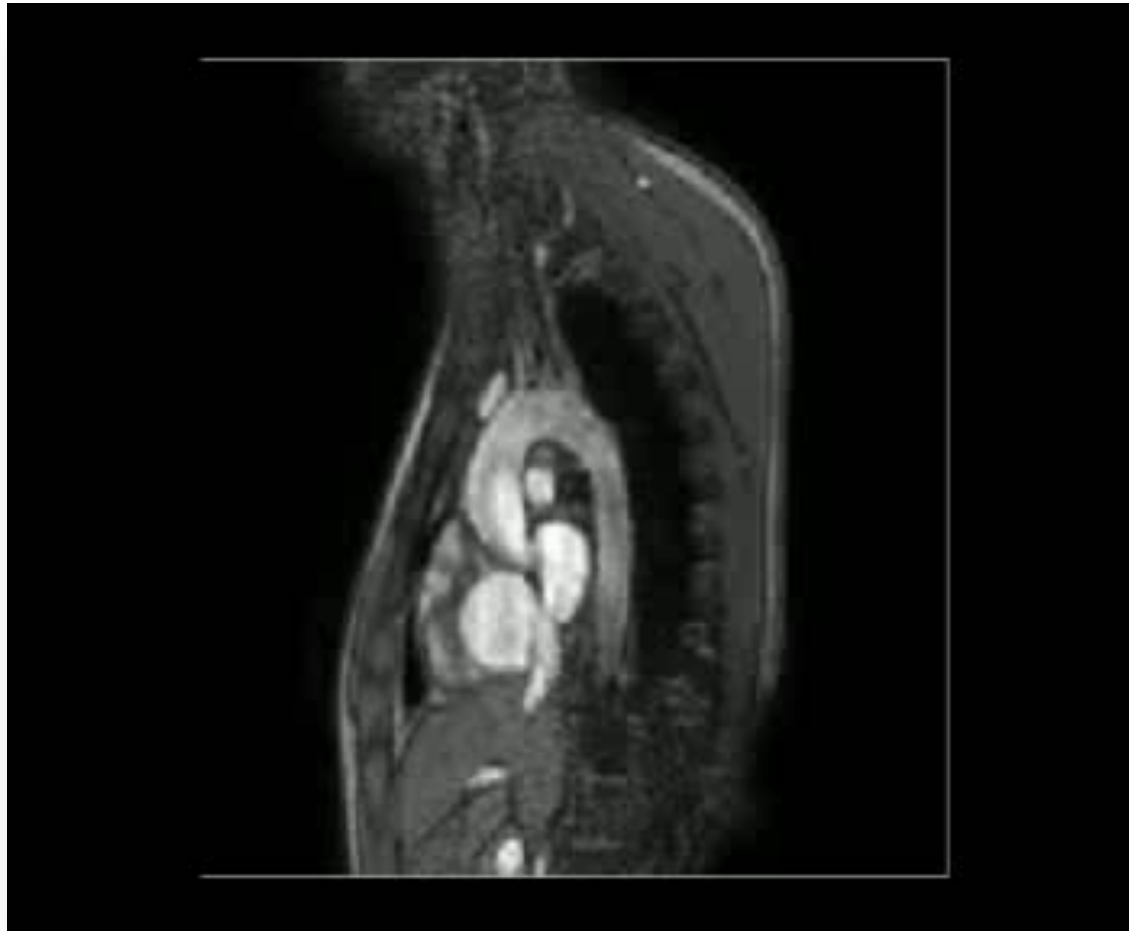
Plein S et al. Radiology 2005

Jung B et al. JMRI 2008

Vitanis V et al. MRM 2010

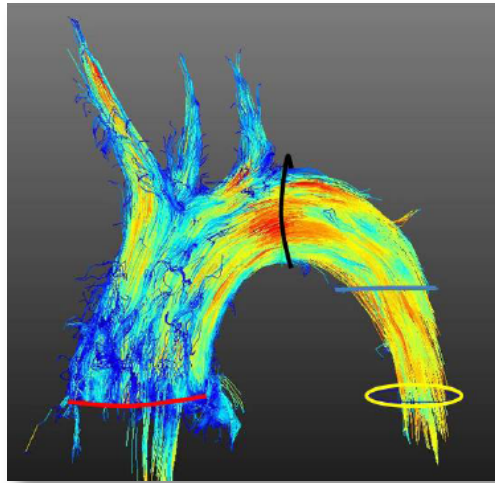
3D Blood flow quantification

8x 3D k-t PCA

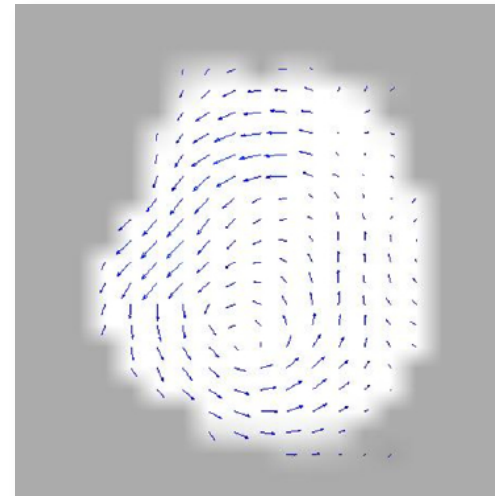
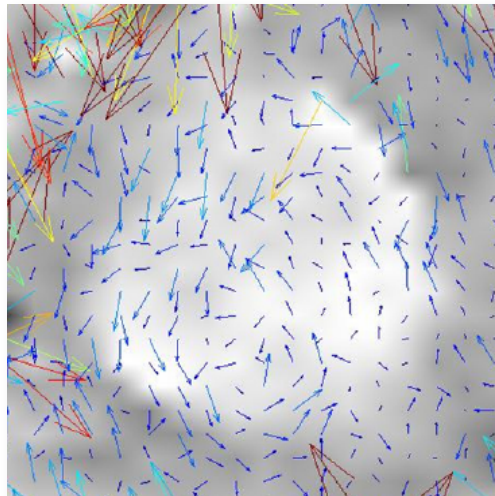
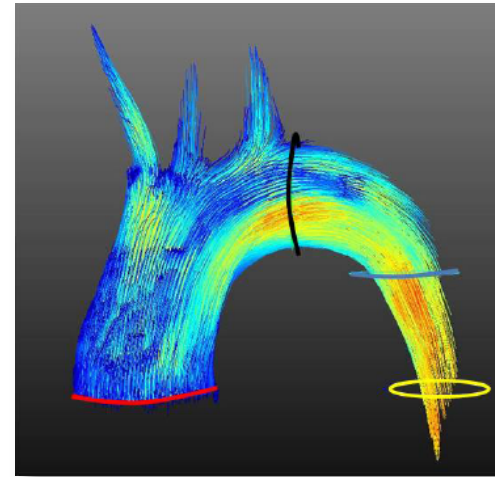


G. Crelier, GyroTools

Divergence-free constraint

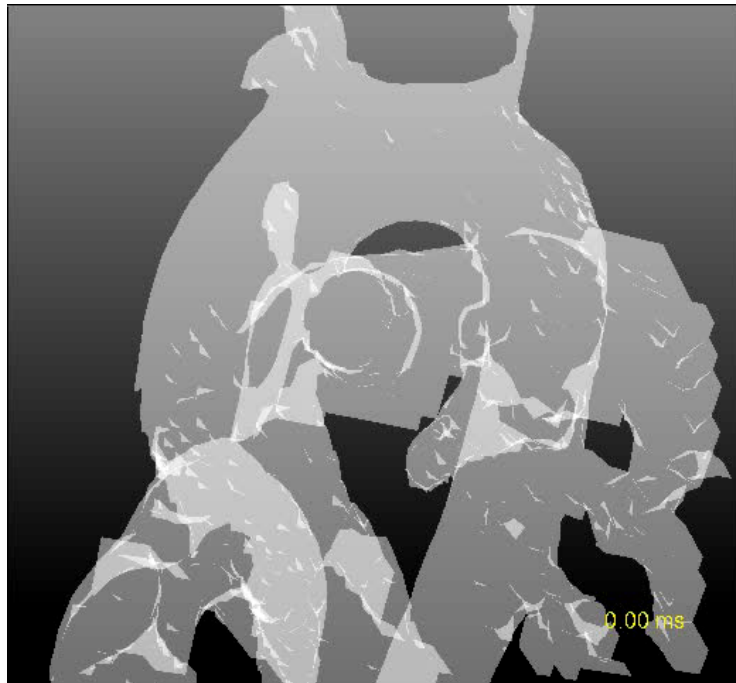


$$\text{div}(\mathbf{v})=0$$

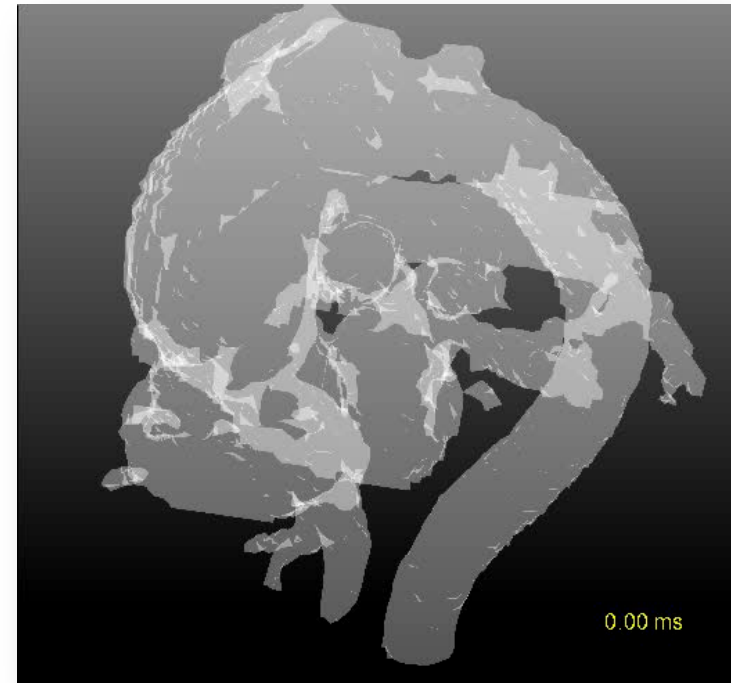


3D Blood flow quantification

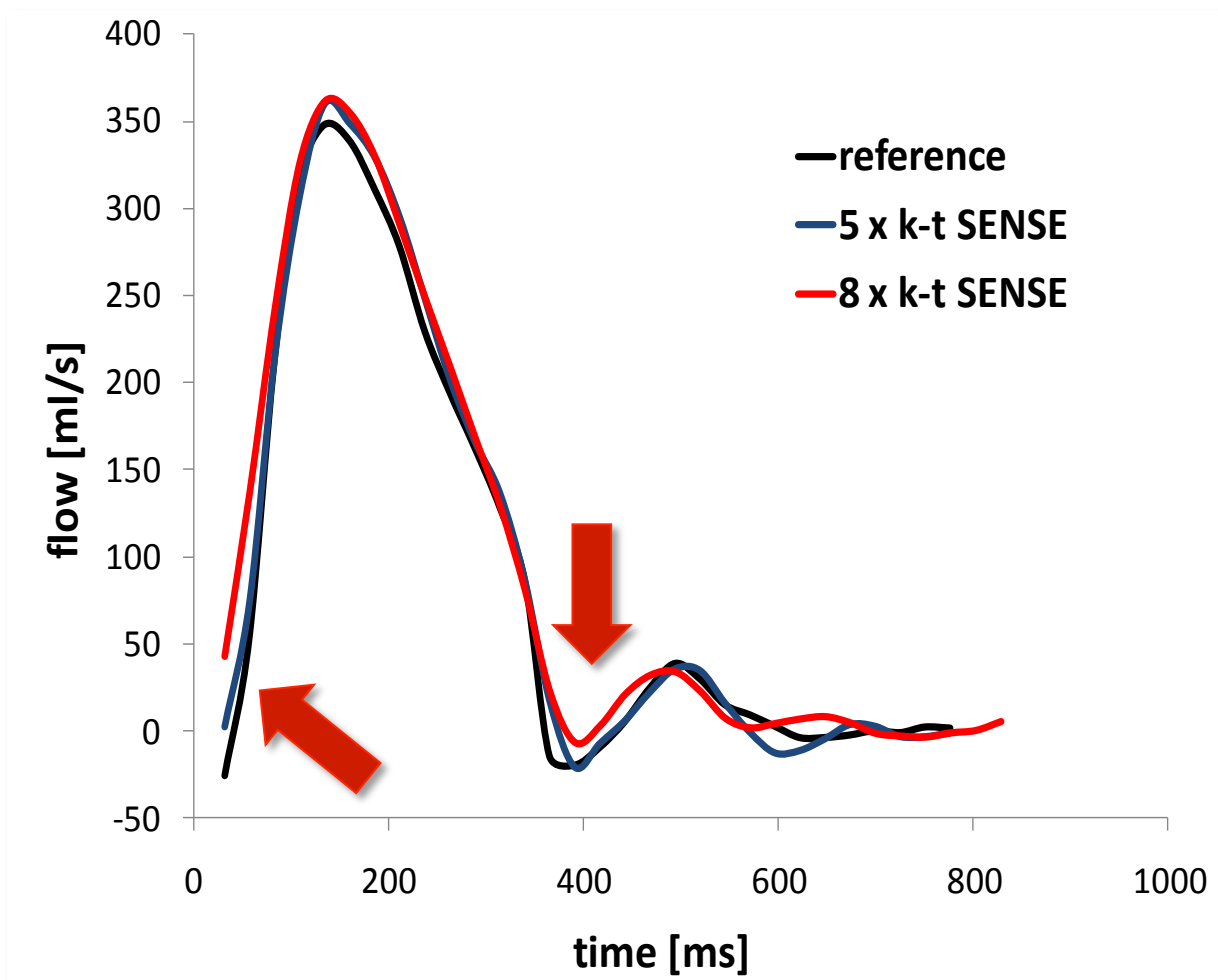
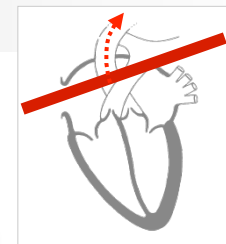
Healthy volunteer



Patient with dilated aorta



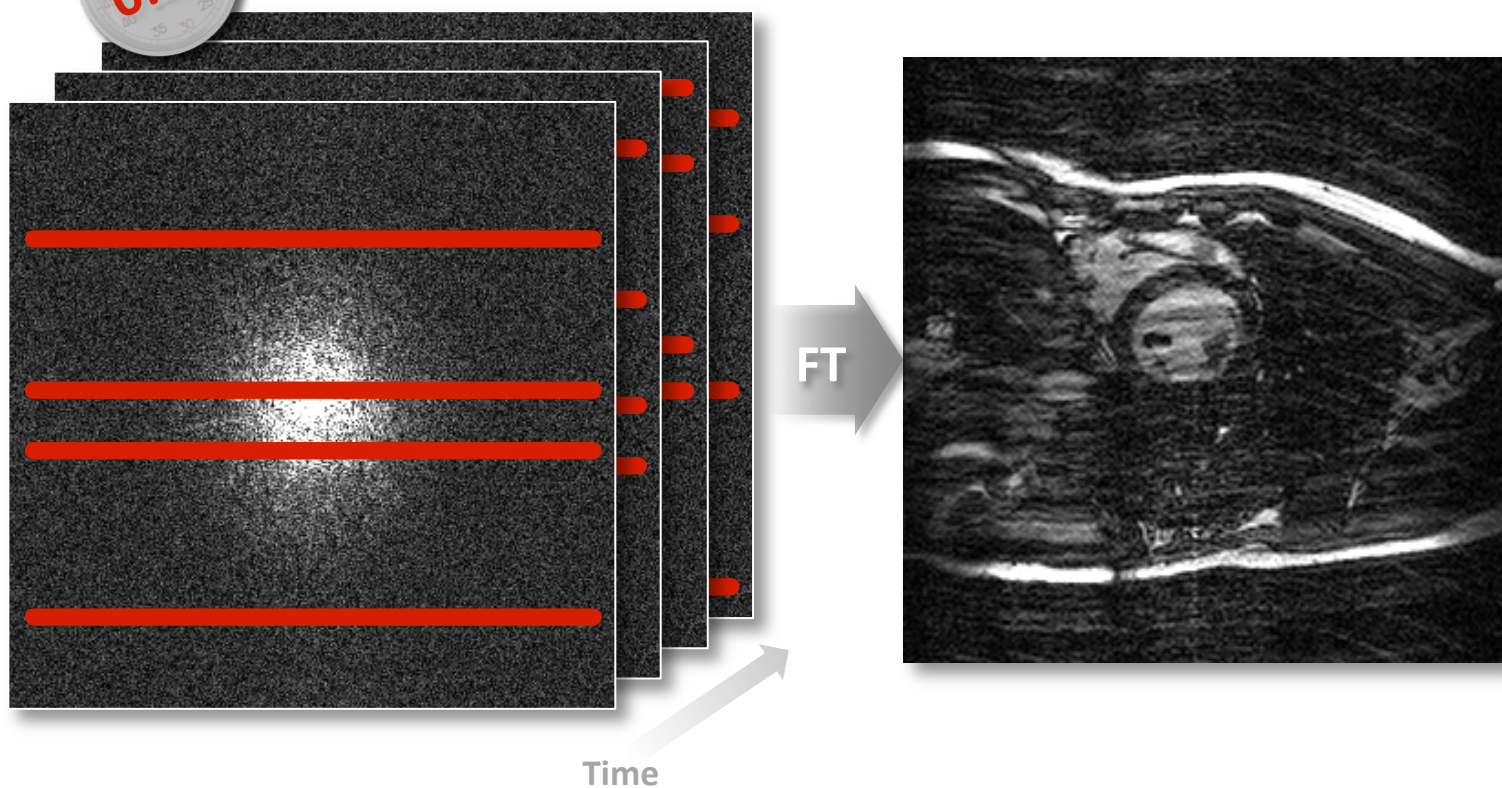
Limitations



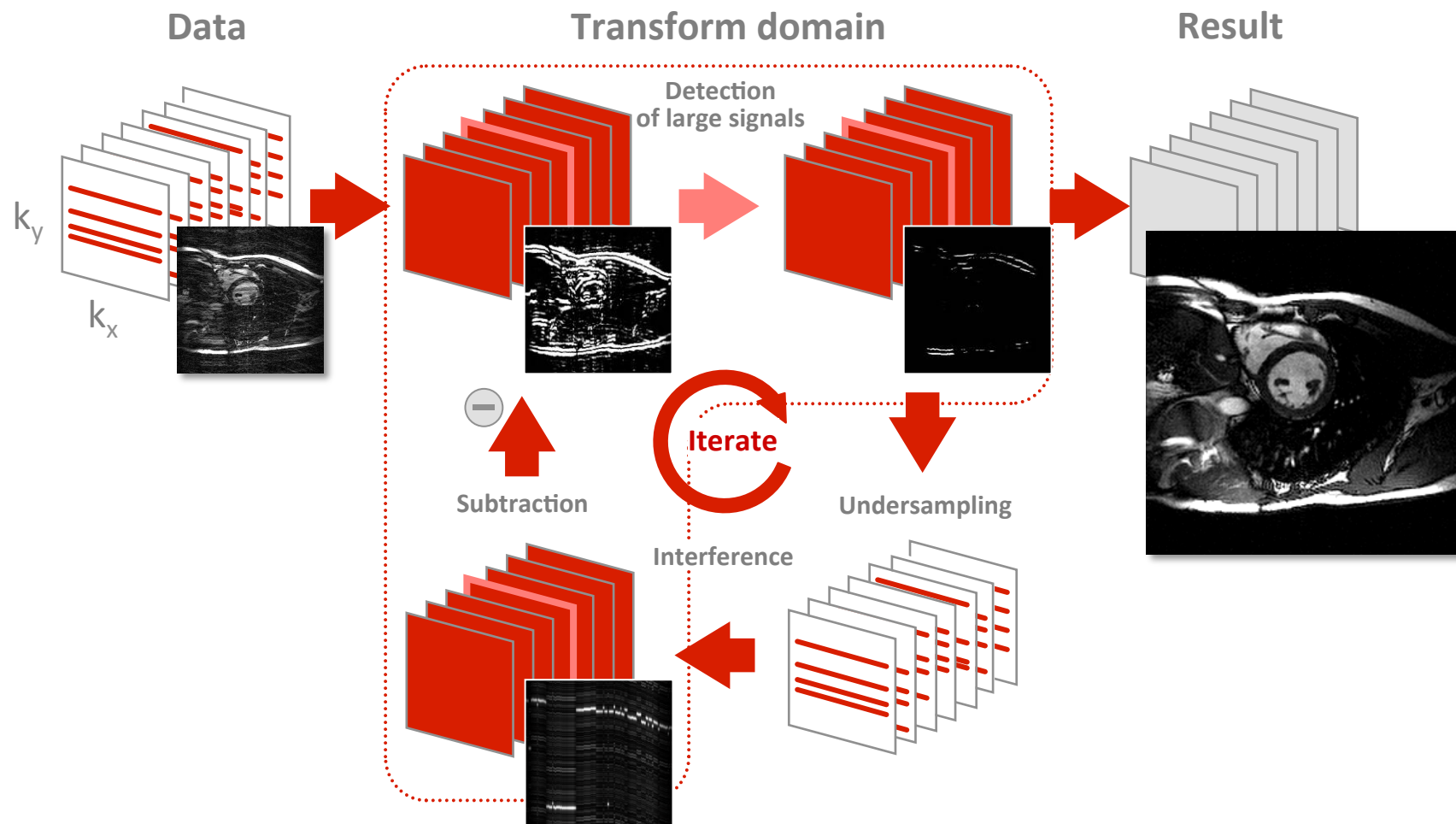
Random undersampling



Random spatiotemporal undersampling



Compressed Sensing

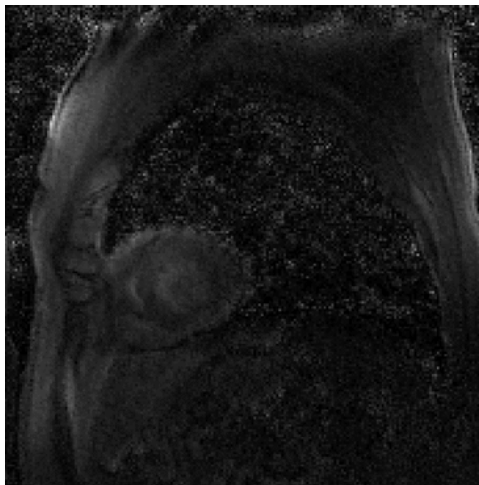


Perfusion imaging

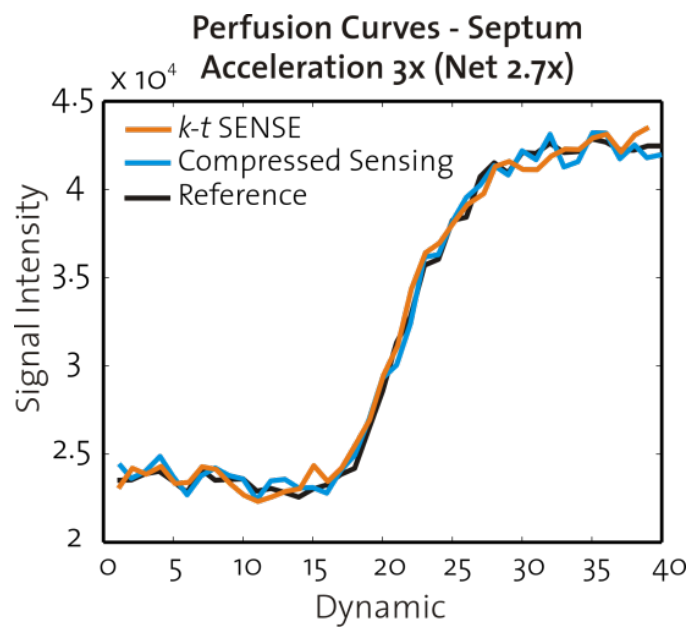
Reference



3x *k-t* CS



3x *k-t* SENSE

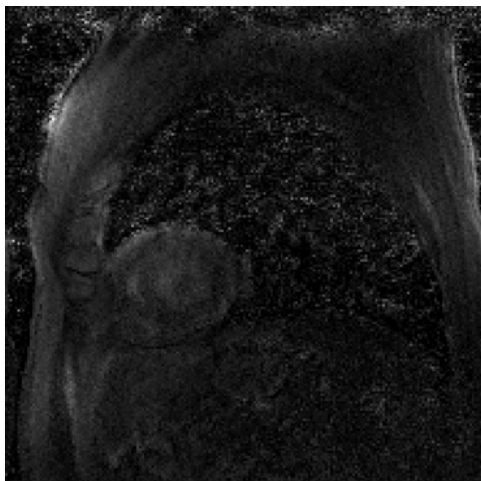


Perfusion imaging

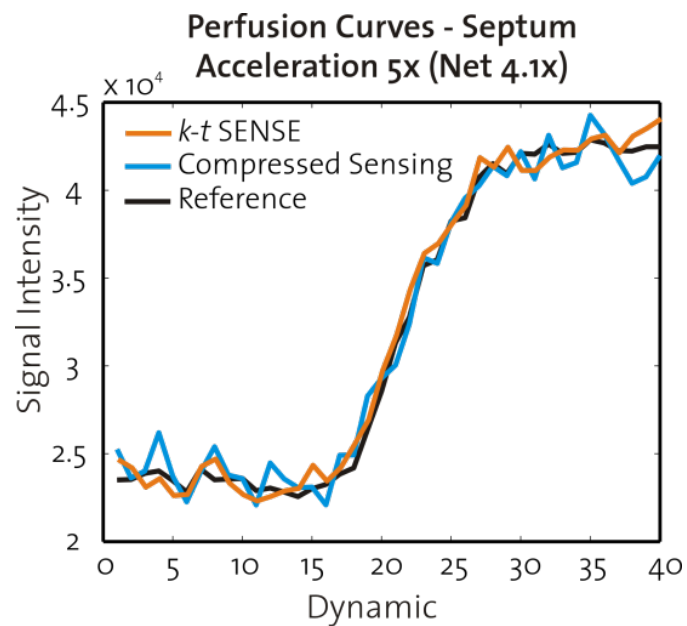
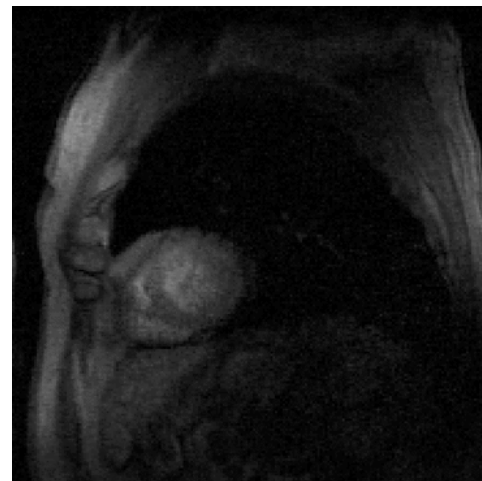
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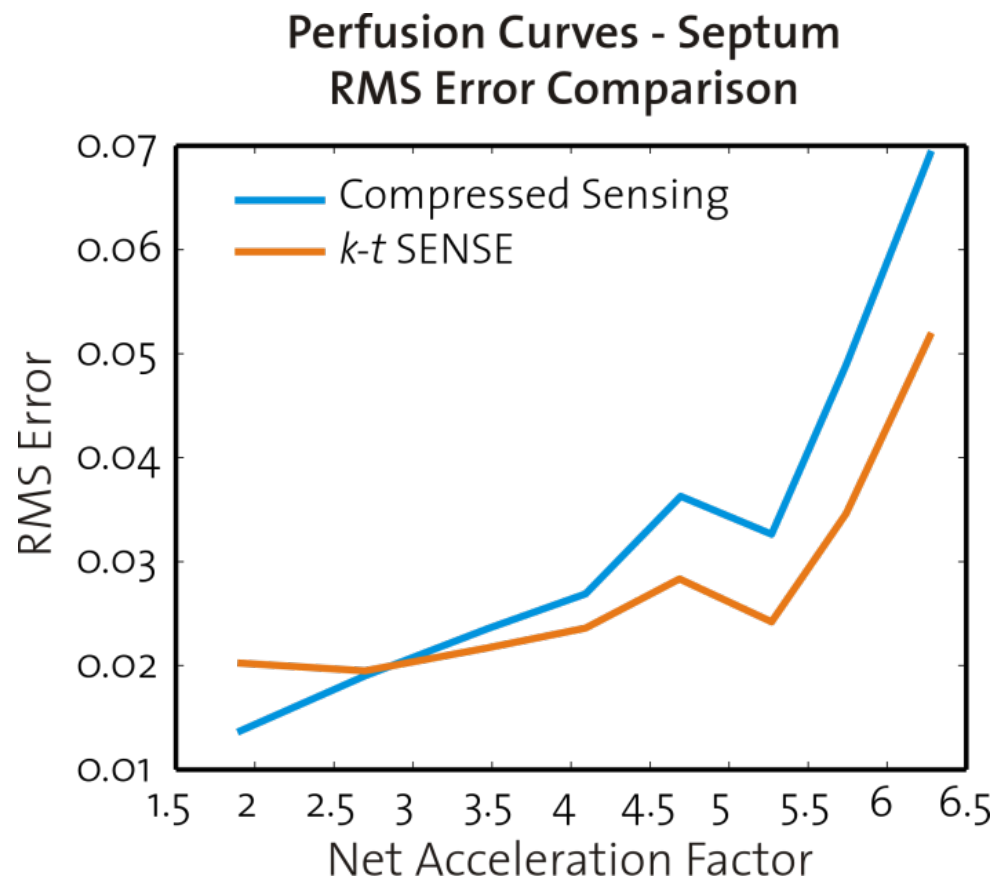
5x *k-t* CS



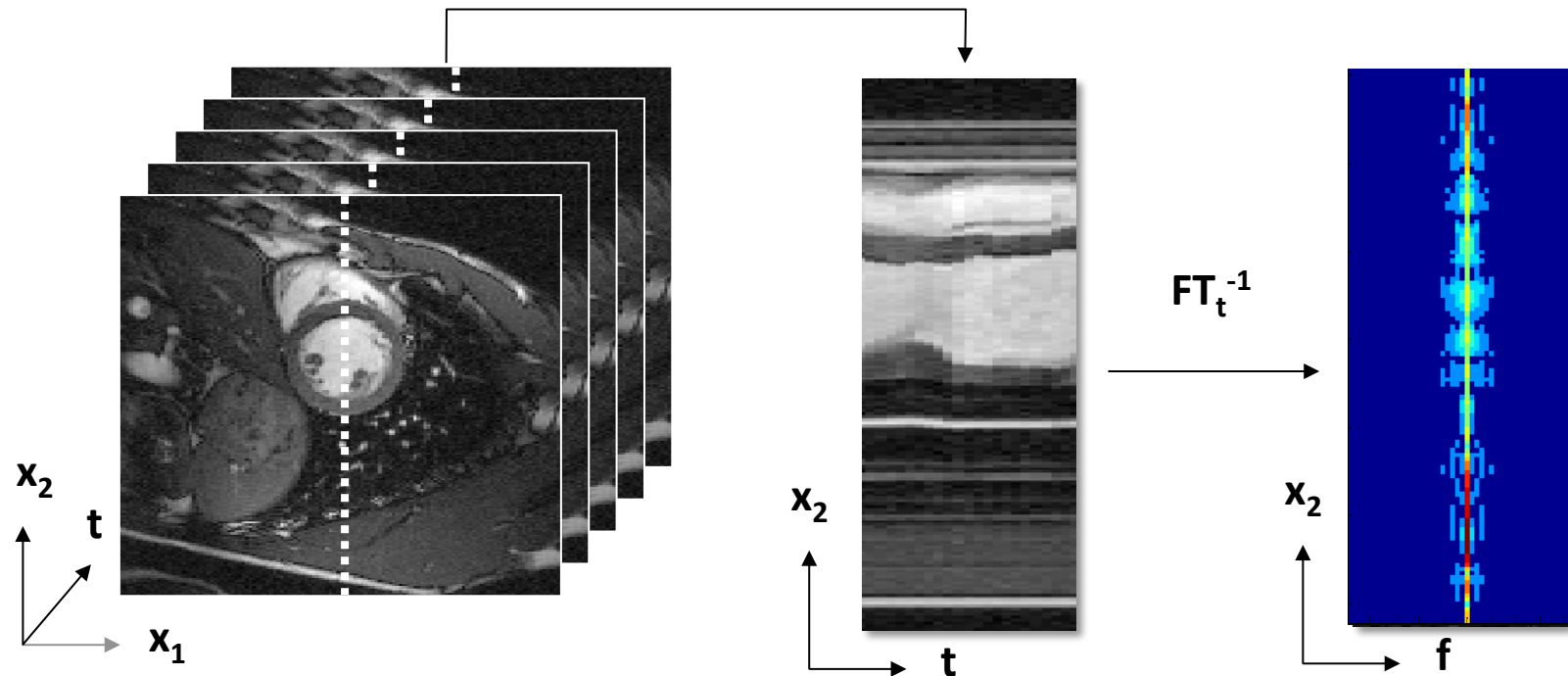
5x *k-t* SENSE



Random versus uniform undersampling



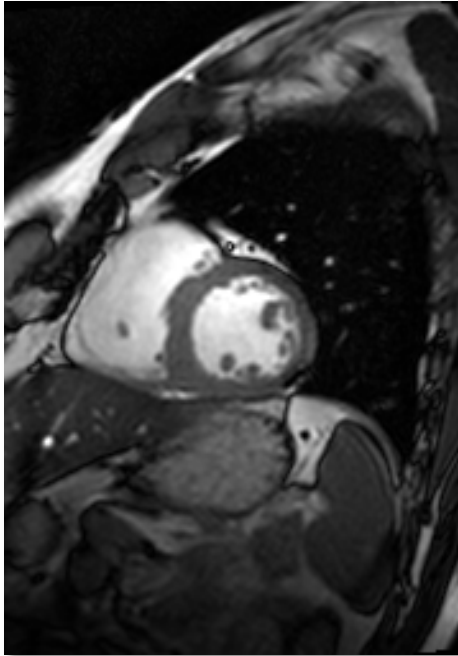
k-t group sparsity intensity (k-t GSI)



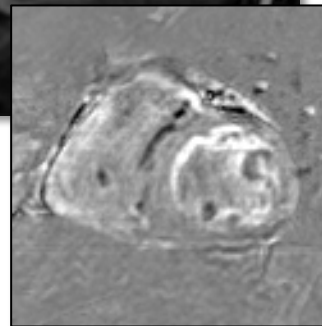
$$\min_m \|i^g\|_{1,2} = \|i_1^g\|_{1,2} + \|i_2^g\|_{1,2} + \dots + \|i_k^g\|_{1,2} \quad \text{s.t.} \quad \|E_u i - d\|_2 \leq \varepsilon$$

k-t GSI and k-t SPARSE

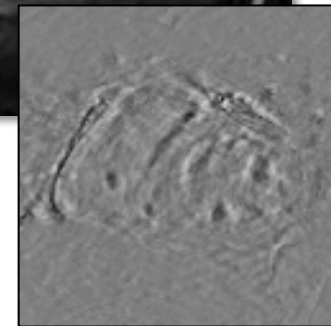
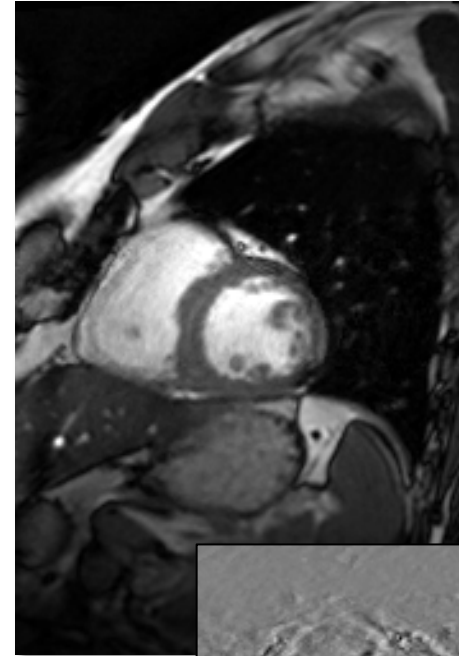
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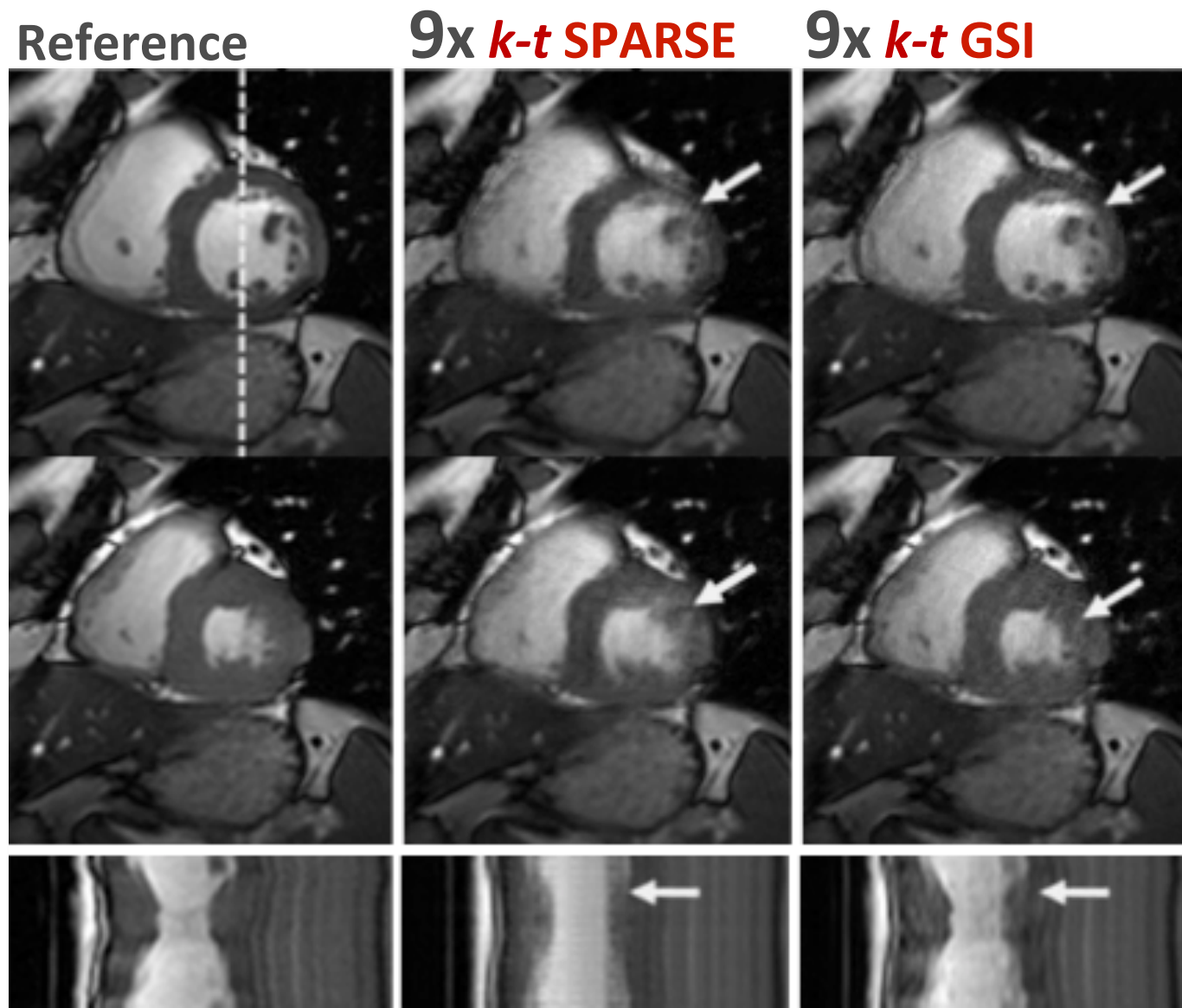
9x *k-t* SPARSE



9x *k-t* GSI



k-t GSI and k-t SPARSE



Summary

- Significant spatiotemporal correlation** in dynamic data
 - Compact data representation** in x-f and x-pc spaces
 - Uniform undersampling** of 2-8x with L2 reconstruction
 - Random undersampling** of 9x by exploiting group sparsity
 - Temporal filtering** results if undersampling exceeds limits
 - Combination with parallel imaging** to effort higher factors
- 