

Performance of l1-norm minimizing regularizers on phase/magnitude reconstruction in flow encoded MRI

2011 BASP Frontiers Wokshop

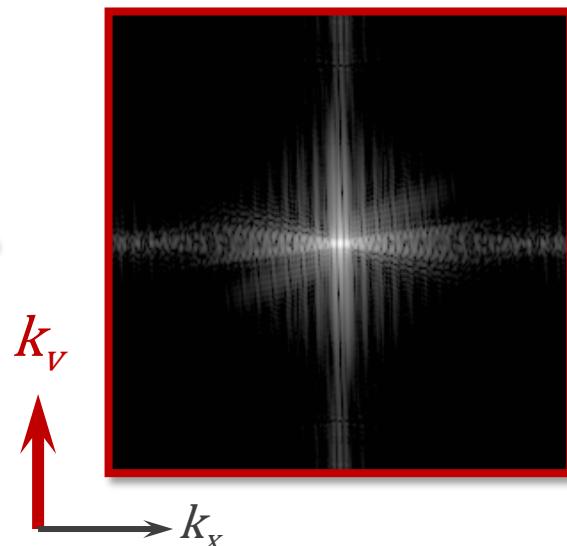
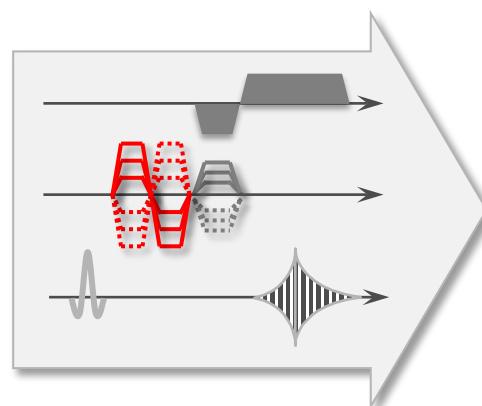
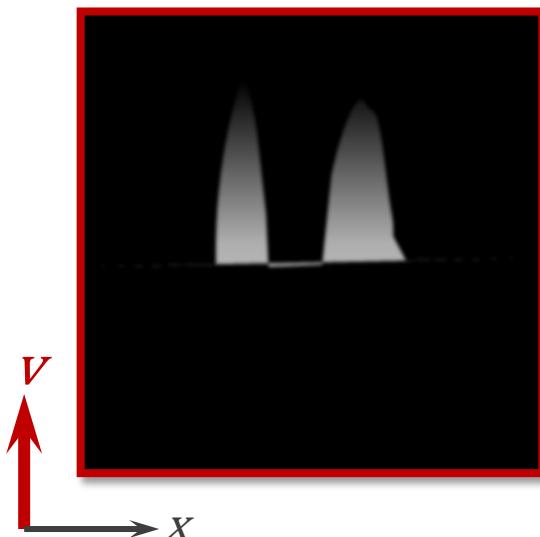
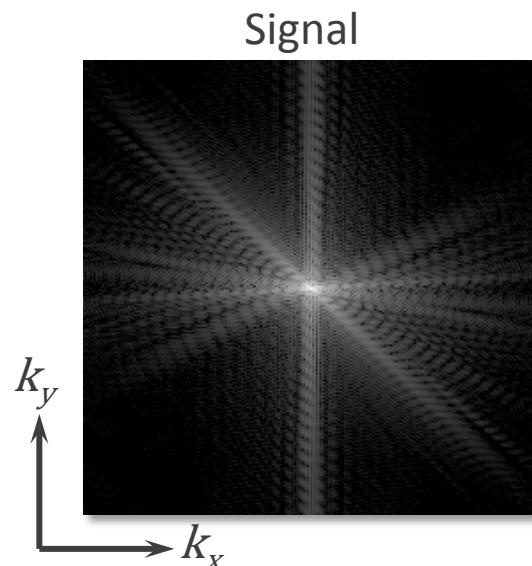
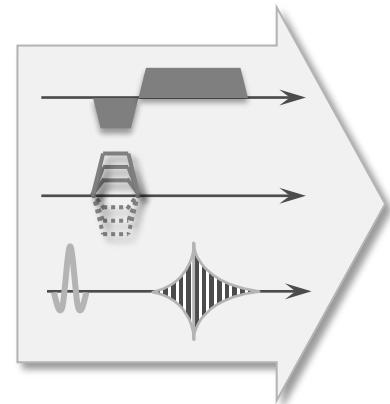
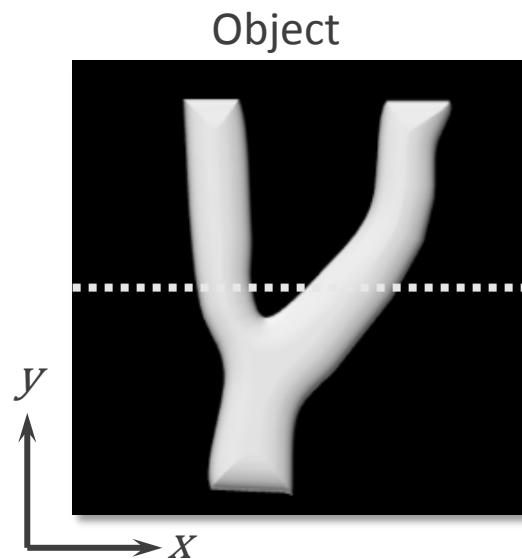
Claudio Santelli



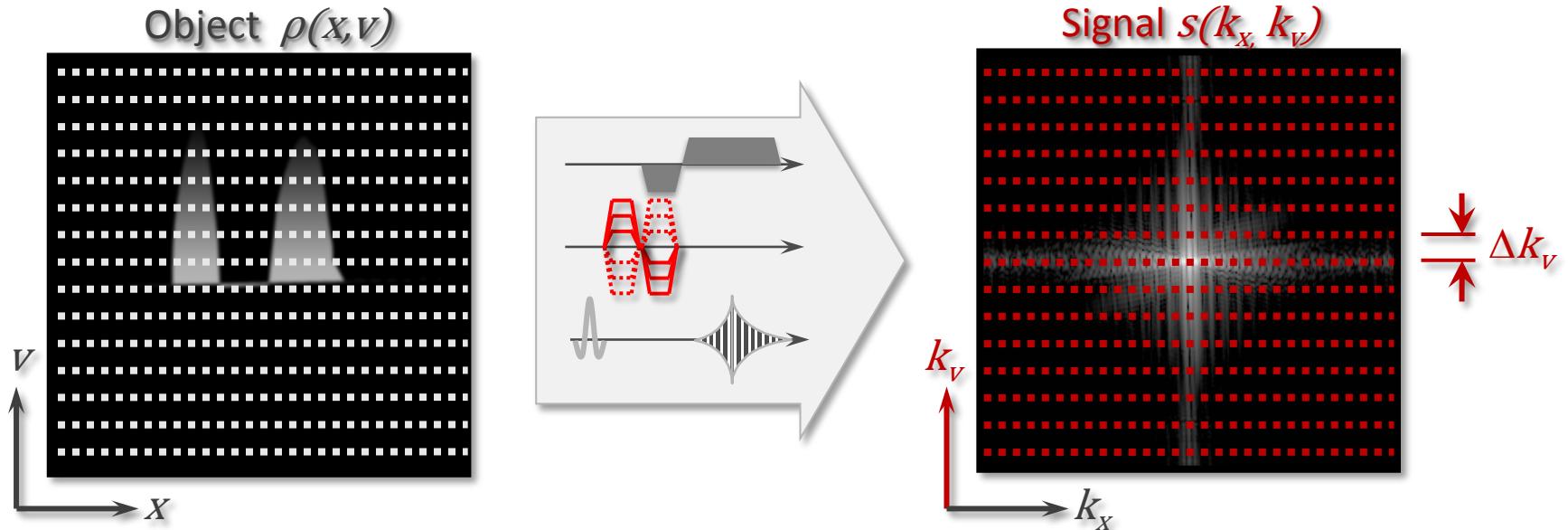
Flow encoding



Fourier velocity encoding (FVE)



Fourier velocity encoding (FVE)



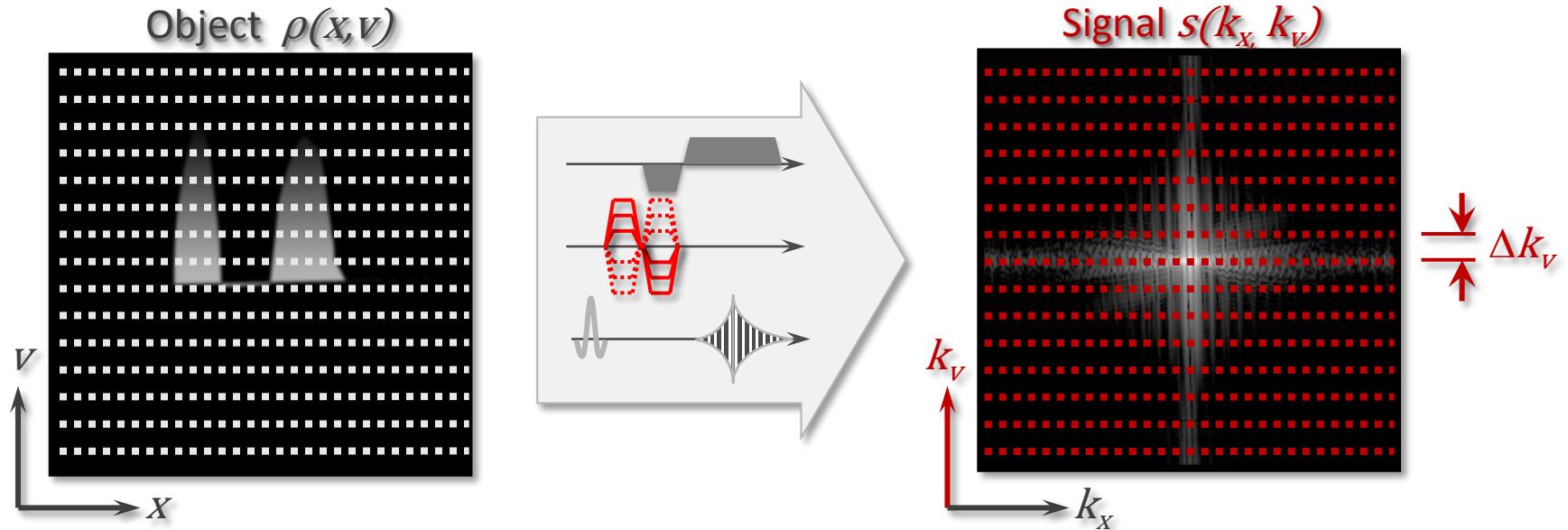
$$s(\mathbf{k}_x(t), \mathbf{k}_v(t)) = \iint \rho(\mathbf{x}, \mathbf{v}) e^{-j[\mathbf{k}_x(t) \cdot \mathbf{x} + \mathbf{k}_v(t) \cdot \mathbf{v}]} d\mathbf{x} d\mathbf{v}$$

$$\mathbf{k}_x(t) = \gamma \int_0^t \mathbf{G}(\tau) d\tau , \quad \mathbf{k}_v(t) = \gamma \int_0^t \tau \mathbf{G}(\tau) d\tau$$

FOV ... field-of-view

FOS ... field-of-speed

Fourier velocity encoding (FVE)



$$s(n, m) = \sum_{x_i} \sum_{v_j} \rho(x_i, v_i) e^{-jn \Delta k_x x_i} e^{-jm \Delta k_v v_j}$$

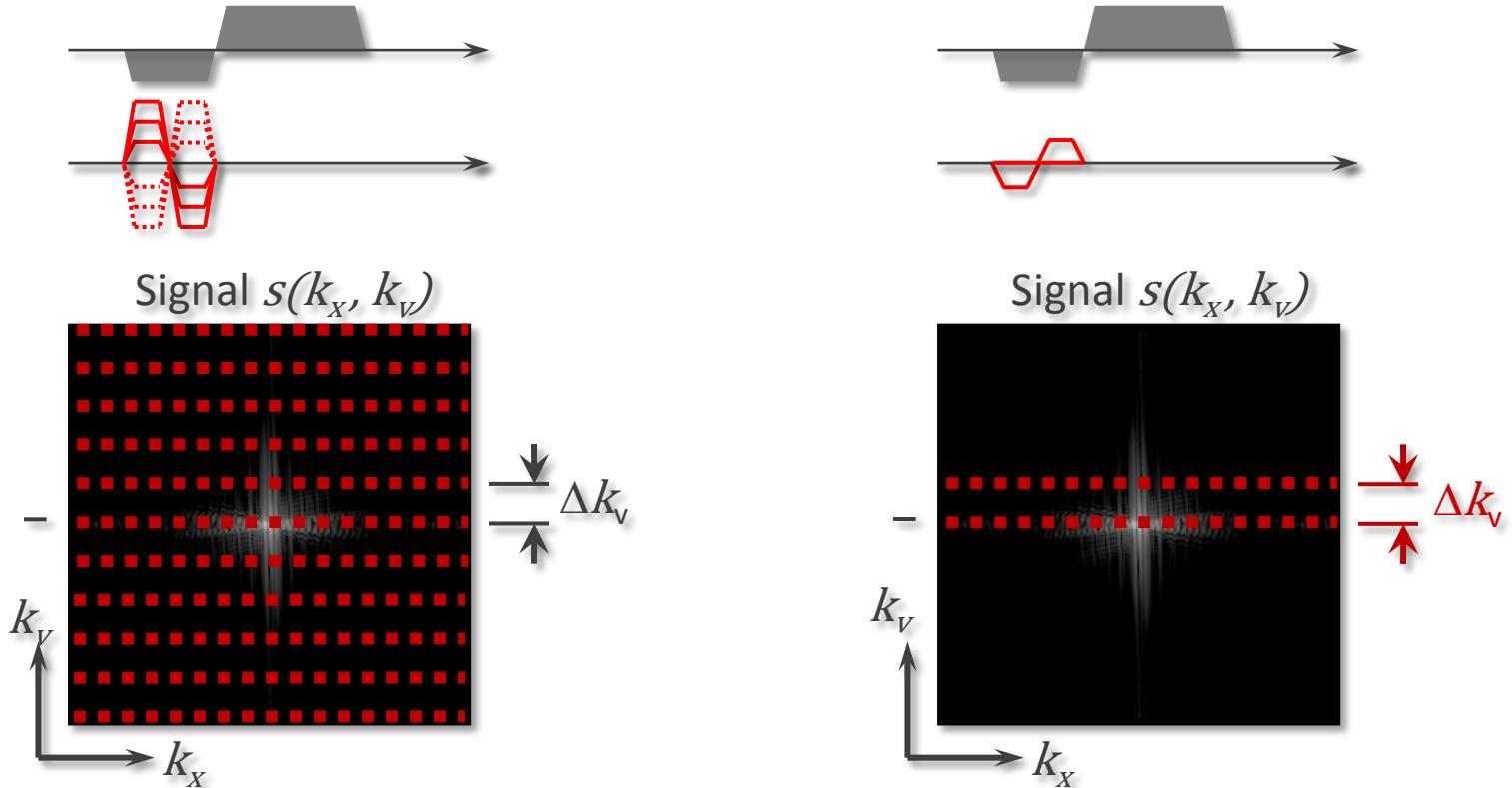
$$\Delta k_x = \frac{2\pi}{FOV} , \quad \Delta k_v = \frac{2\pi}{FOS} = \frac{\pi}{V_{enc}}$$

FOV ... field-of-view

FOS ... field-of-speed

$$\mathbf{s} = \mathbf{E}\boldsymbol{\rho}$$

FVE vs. phase contrast (PC)



$$\rho(x, v) = \sum_{k_x} \sum_{k_v} s(k_x, k_v) e^{j k_x x} e^{j k_v v}$$

$$\rho_A(x, y) = \sum_{k_x} s(k_x, k_v = 0) e^{j k_x x}$$

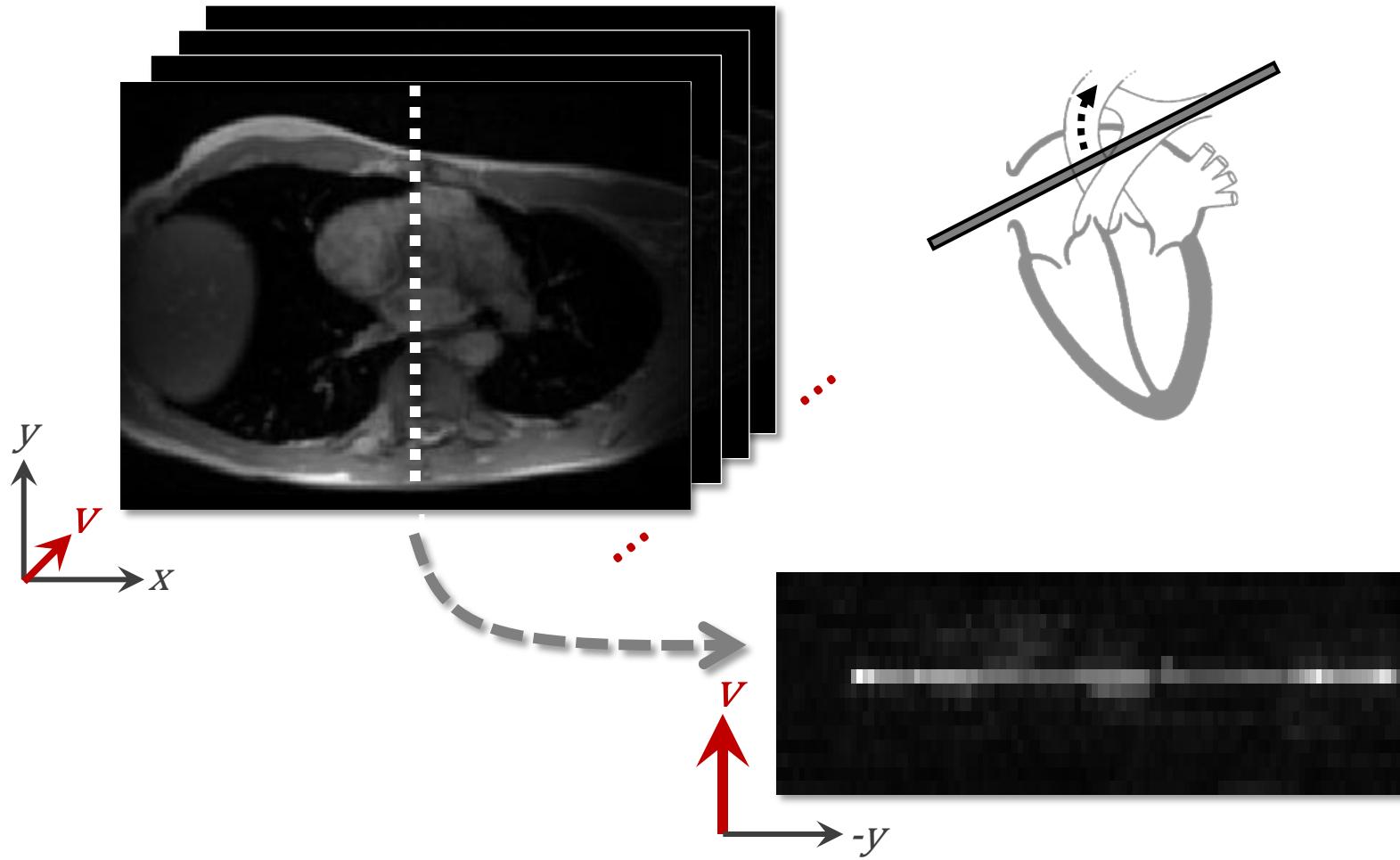
$$\rho_B(x, y) = \sum_{k_x} s(k_x, k_v = \Delta k_v) e^{j k_x x} e^{j \Delta k_v v}$$

Phase difference

$$\angle \rho(x, y) = \angle (\rho_A^* \cdot \rho_B) = \Delta k_v v$$

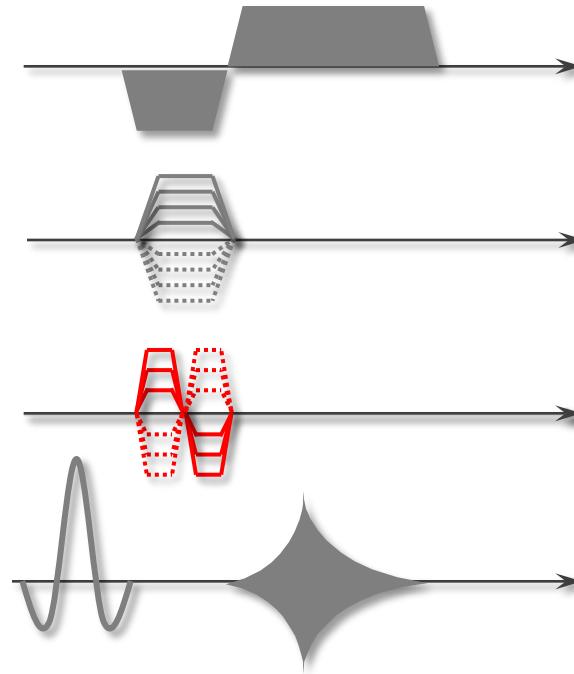
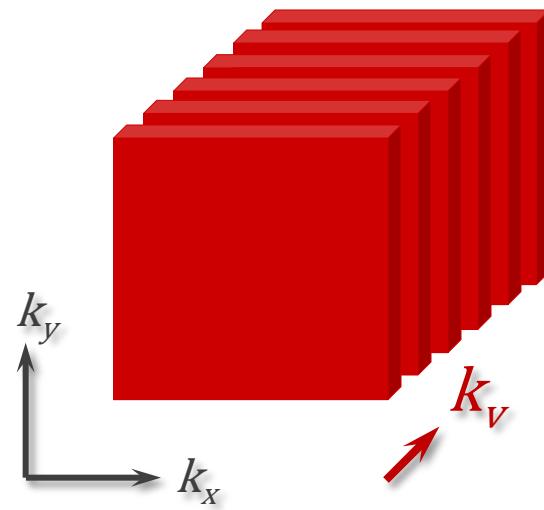
Fourier velocity encoding

- Velocity distributions in ascending aorta



Fourier velocity encoding

- Scan time



$$t_{scan} \propto N_y \mathbf{N_v}$$

Speed-up

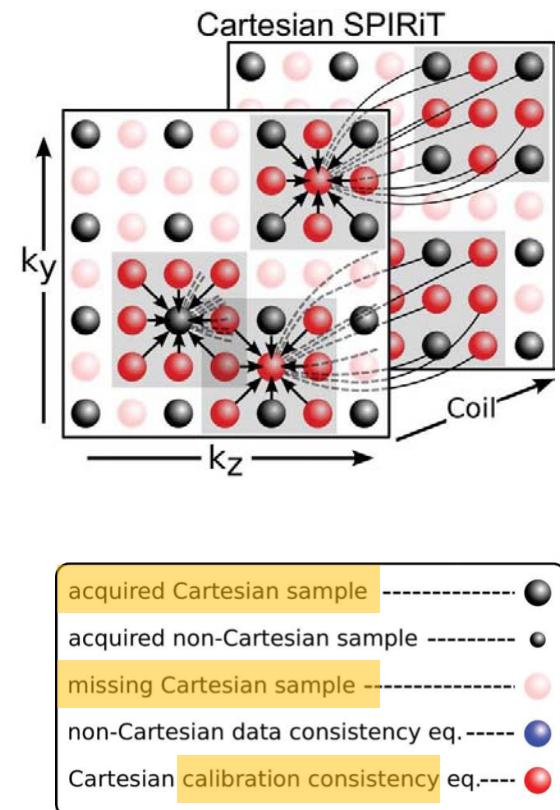


SPIRiT

- “Iterative self-consistent parallel imaging reconstruction”
- Full and shift invariant interpolation kernel, i.e. convolution in k-space
- $s_c = \mathbf{G}s_c$ Consistency equation for all k-space points
 - s_c : Cartesian k-space over all coils
 - \mathbf{G} : Interpolation matrix (Convolution operator)

Calibration consistency:

Minimize $\|(\mathbf{G} - \mathbf{I})s_c\|_2^2$

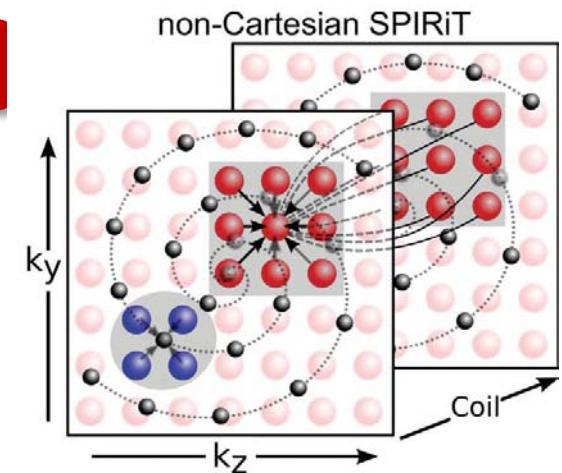


SPIRiT

- Data consistency:

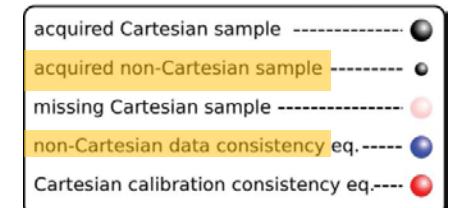
$$\text{Minimize } \|(\mathbf{G} - \mathbf{I})\mathbf{s}_c\|_2^2, \text{ s.t. } \|\mathbf{D}\mathbf{s}_c - \mathbf{s}\|_2^2 \leq \varepsilon$$

\mathbf{D} : Linear Operator which maps reconstructed
Cartesian k-space to the acquired data \mathbf{s}



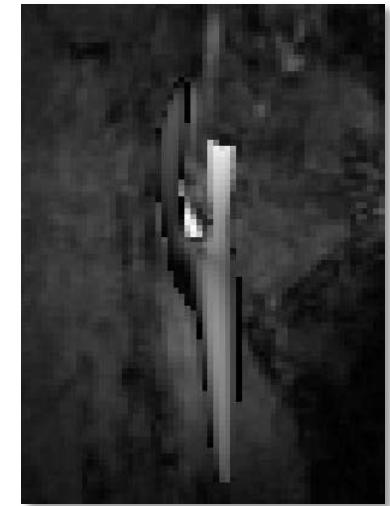
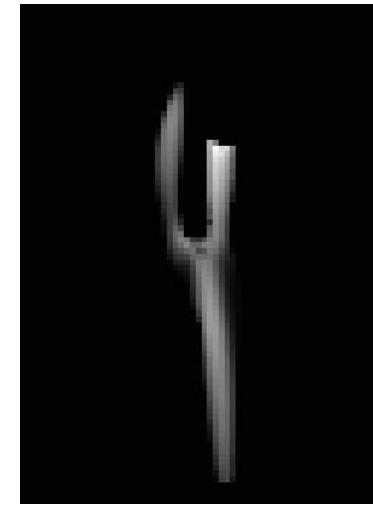
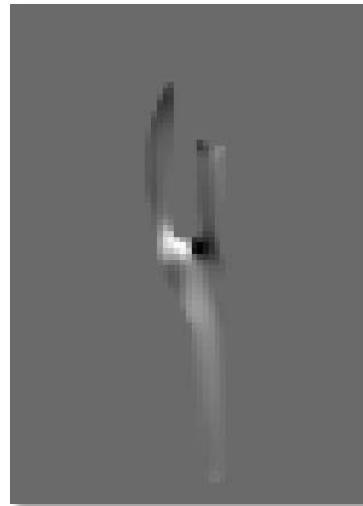
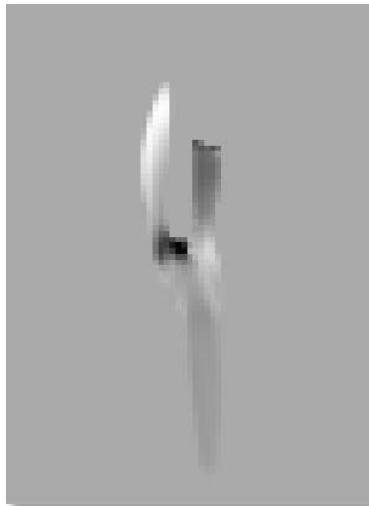
- Image priors

$$\operatorname{argmin}_{\mathbf{s}_c} \|\mathbf{D}\mathbf{s}_c - \mathbf{s}\|_2^2 + \lambda_1 \|(\mathbf{G} - \mathbf{I})\mathbf{s}_c\|_2^2 + \sum_{j \geq 2} \lambda_j R_j(\mathbf{s}_c)$$



SPIRiT – PC (Cartesian)

- CFD model (SNR = 30, 66 x 89 x 33 voxels)



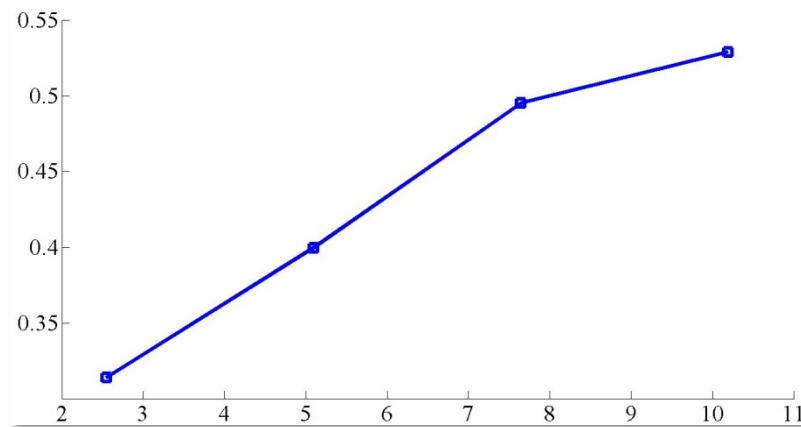
- Image priors (wavelet, total variation)

$$\underset{\mathbf{s}_c}{\operatorname{argmin}} \quad \|\mathbf{D}\mathbf{s}_c - \mathbf{s}\|_2^2 + \lambda_1 \|(\mathbf{G} - \mathbf{I})\mathbf{s}_c\|_2^2 + \lambda_2 \|\boldsymbol{\Psi}\mathcal{F}^{-1}\mathbf{s}_c\|_1 + \lambda_3 \|\nabla\mathcal{F}^{-1}\mathbf{s}_c\|_1$$

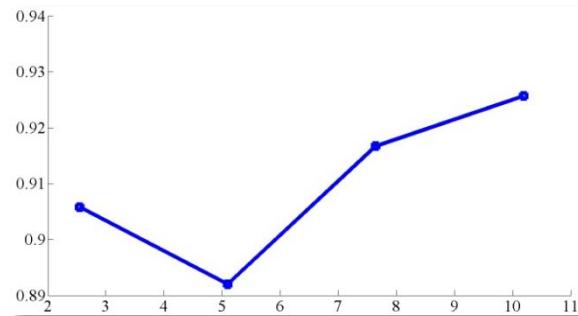
SPIRiT – PC (Cartesian)

- RMS error vs. undersampling factor

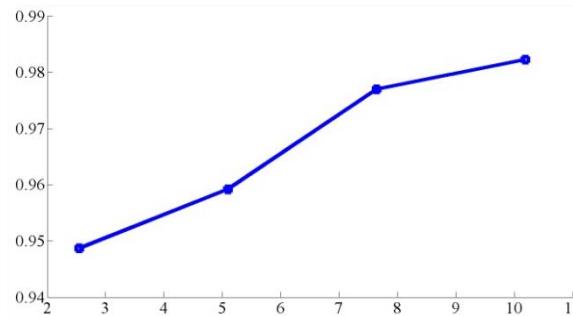
- Magnitude



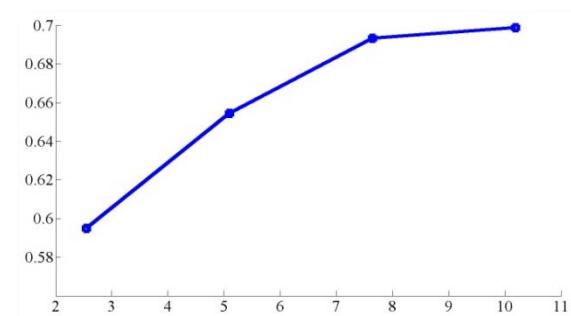
- Phase



Segm 1



Segm 2

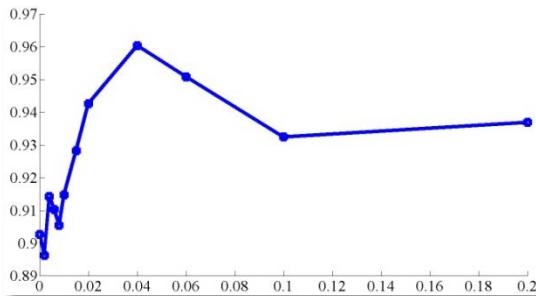


Segm 3

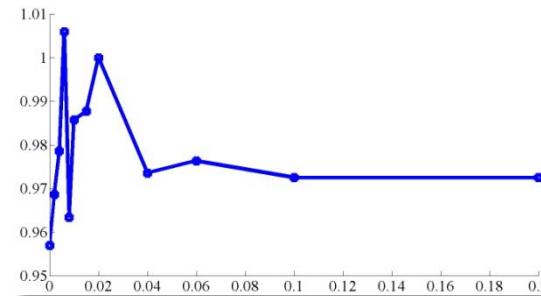
SPIRiT – PC (Cartesian)

- Phase error vs regularization weights (7.5x undersampling)

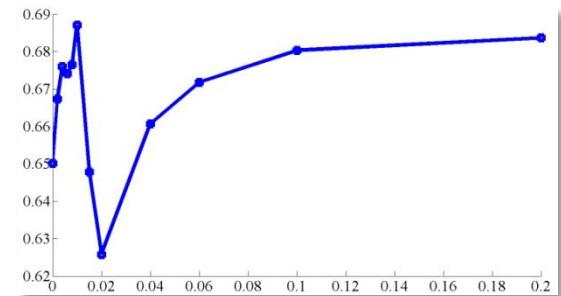
$$\operatorname{argmin}_{\mathbf{s}_c} \|\mathbf{D}\mathbf{s}_c - \mathbf{s}\|_2^2 + \lambda_1 \|(\mathbf{G} - \mathbf{I})\mathbf{s}_c\|_2^2 + \lambda_2 \|\nabla \mathcal{F}^{-1}\mathbf{s}_c\|_1$$



Segm 1

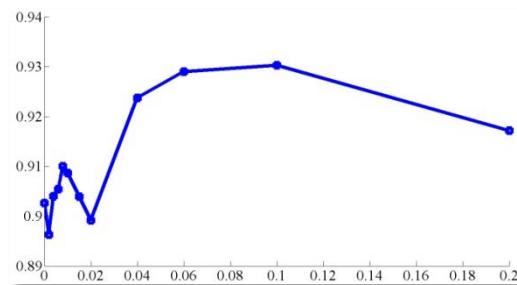


Segm 2

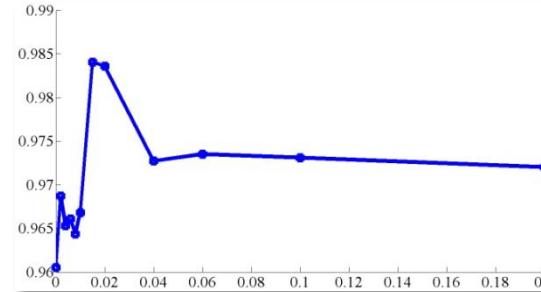


Segm 3

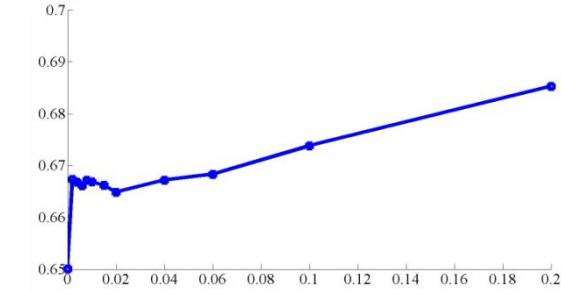
$$\operatorname{argmin}_{\mathbf{s}_c} \|\mathbf{D}\mathbf{s}_c - \mathbf{s}\|_2^2 + \lambda_1 \|(\mathbf{G} - \mathbf{I})\mathbf{s}_c\|_2^2 + \lambda_2 \|\Psi \mathcal{F}^{-1}\mathbf{s}_c\|_1$$



Segm 1



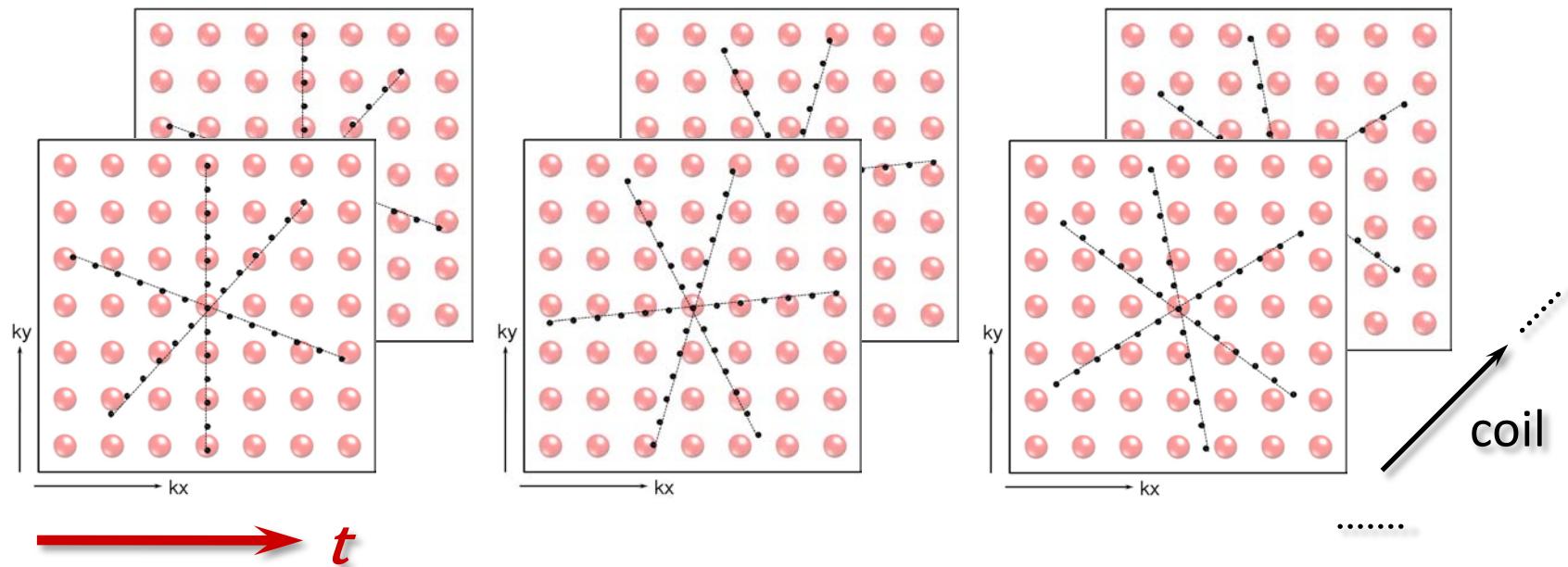
Segm 2



Segm 3

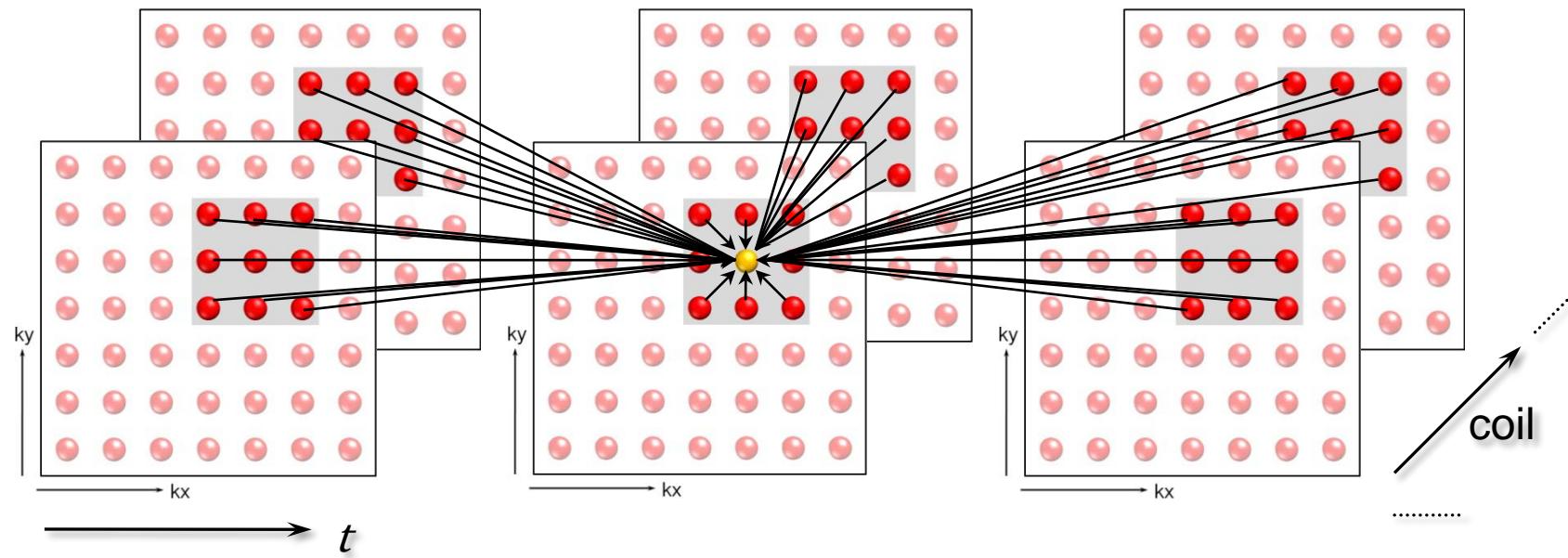
Dynamic imaging

- Dynamic imaging
- Arbitrary k-space trajectories (e.g. golden angle profile)



Dynamic imaging

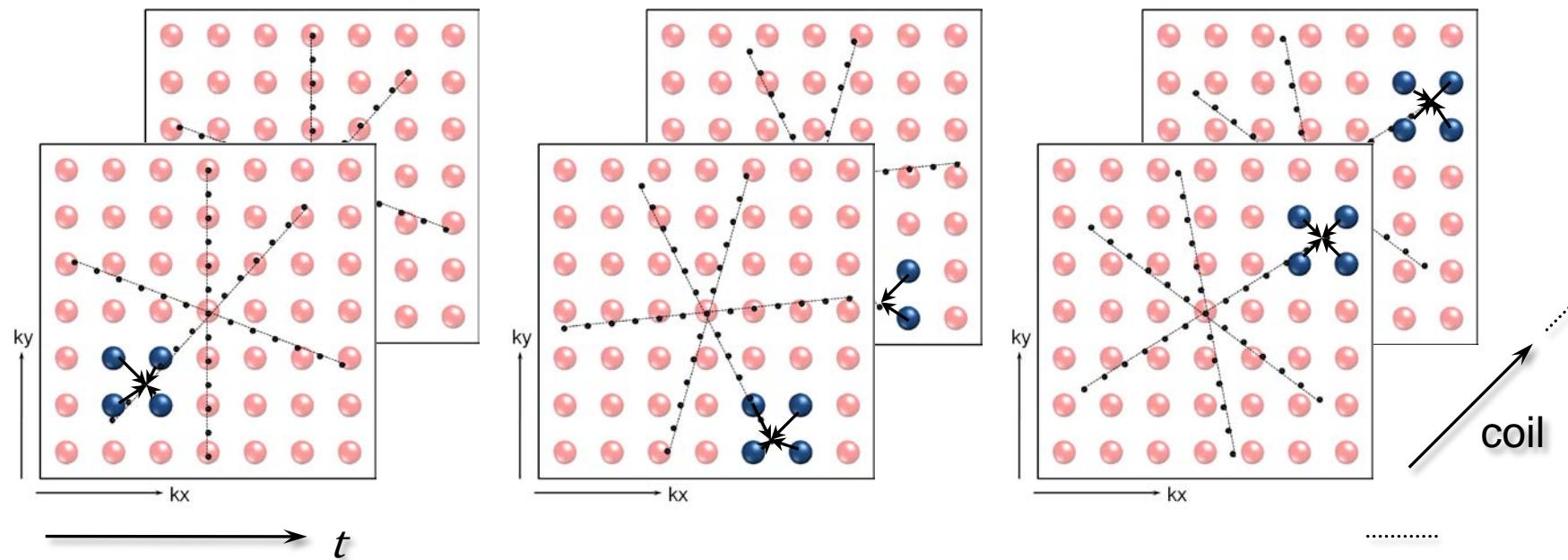
- Calibration consistency



$$\text{Minimize } \|\mathbf{(G - I)s}_c\|_2^2$$

Dynamic imaging

- Data consistency



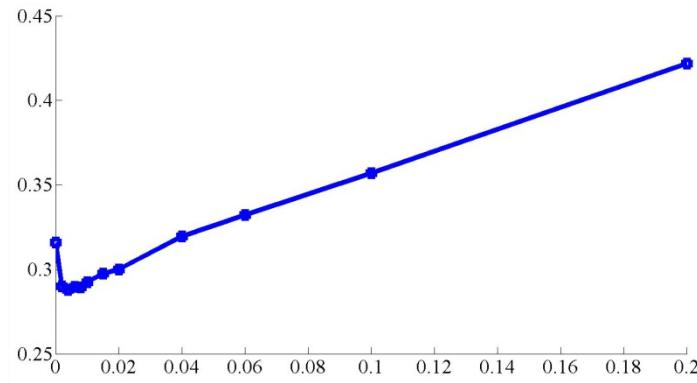
$$\text{Minimize } \|(\mathbf{G} - \mathbf{I})\mathbf{s}_c\|_2^2 , \text{ s.t. } \|\mathbf{D}\mathbf{s}_c - \mathbf{s}\|_2^2 \leq \varepsilon$$

Temporal Fourier transform – PC (Cartesian)

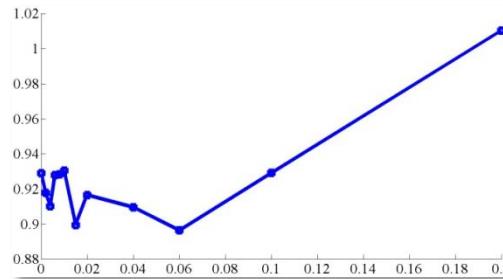
- RMS error vs. regularization weight

$$\underset{\mathbf{s}_c}{\operatorname{argmin}} \quad \|\mathbf{D}\mathbf{s}_c - \mathbf{s}\|_2^2 + \lambda_1 \|(\mathbf{G} - \mathbf{I})\mathbf{s}_c\|_2^2 + \lambda_2 \|\mathcal{F}_t \mathcal{F}^{-1} \mathbf{s}_c\|_1$$

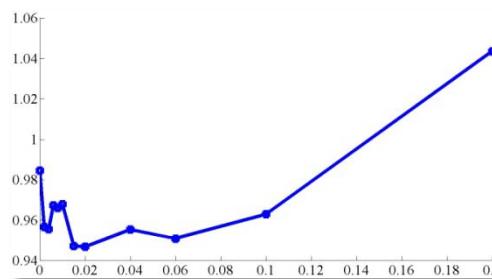
- Magnitude



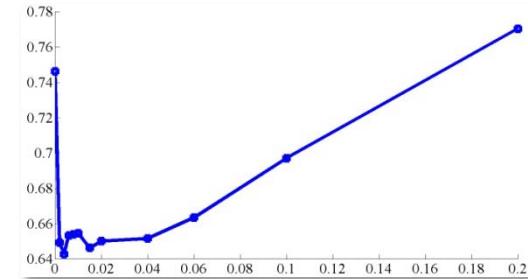
- Phase



Segm 1



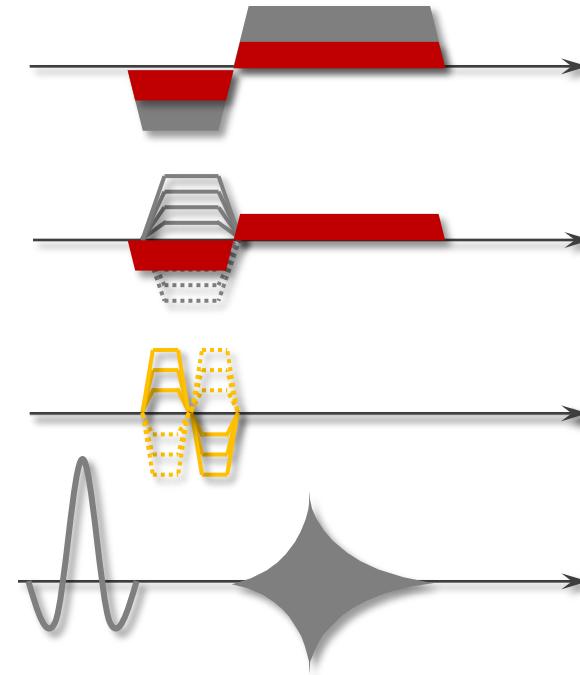
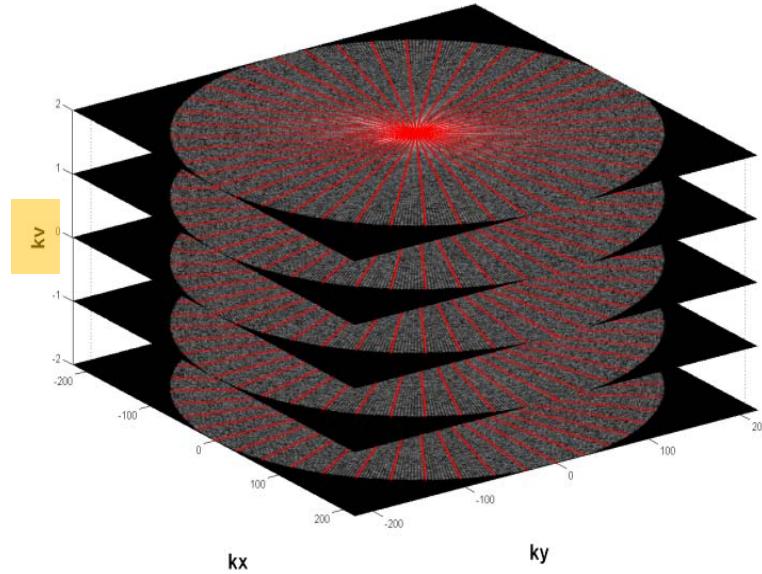
Segm 2



Segm 3

Radial FVE (rFVE)

- Data acquisition

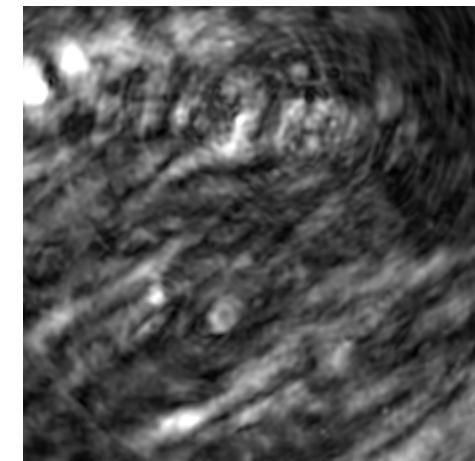
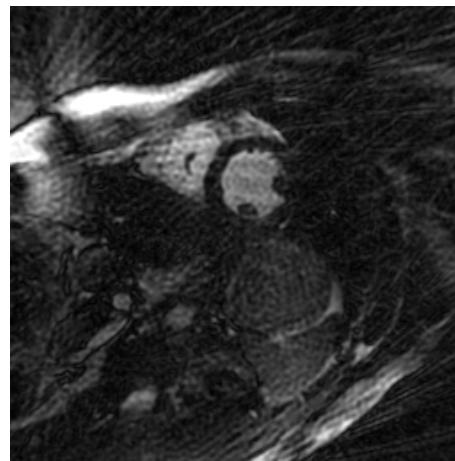
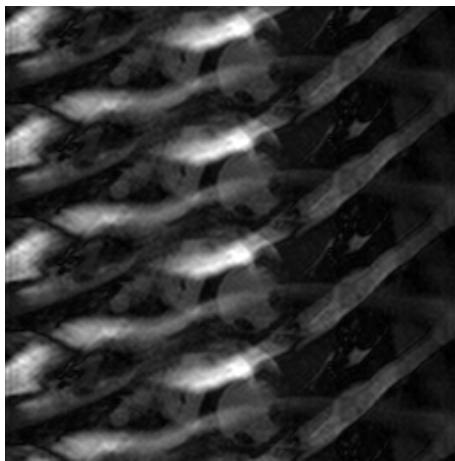
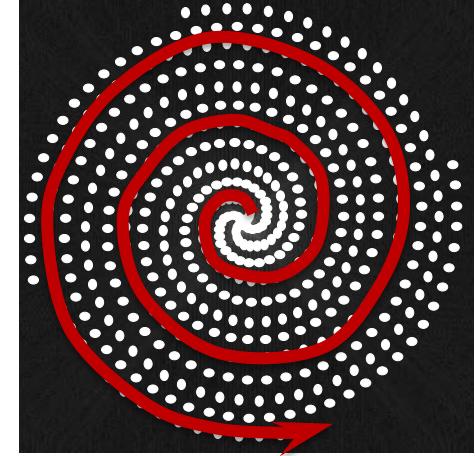
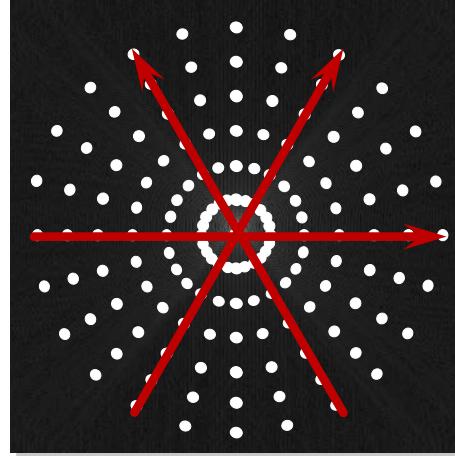
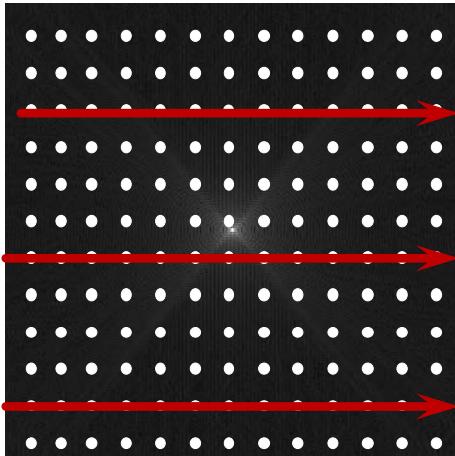


- Why radial?

- Shorter minimum TE
- Oversampled center of k-space (Low res., training/acquisition stage, contrast)
- Motion (Low sensitivity, streak artifacts, motion tracking)

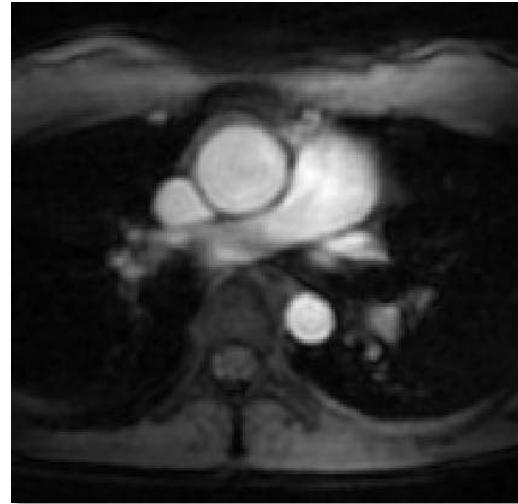
Radial FVE (rFVE)

- Undersampling artifacts (4x, Cartesian, radial, spiral)

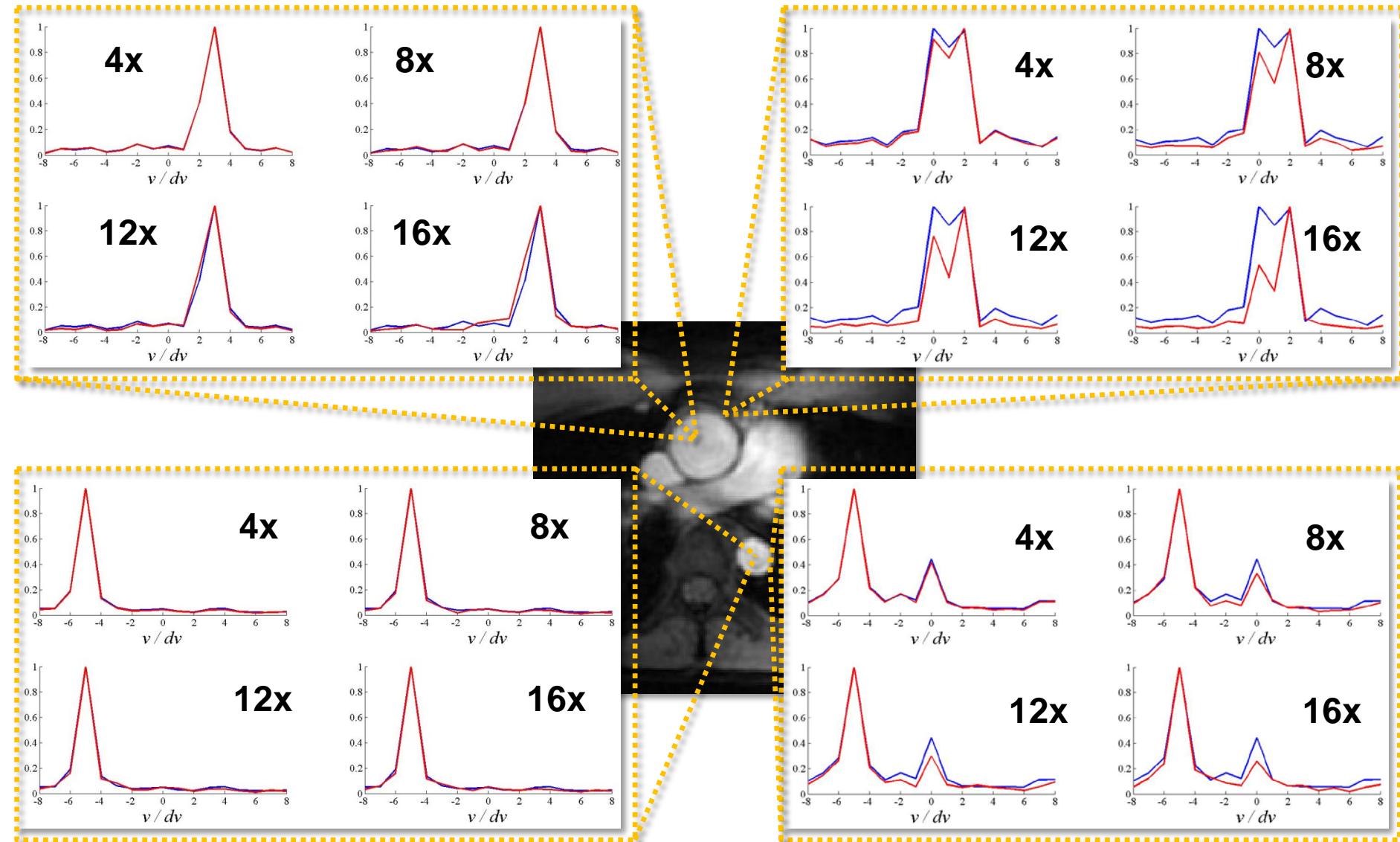


In-vivo - k-t rFVE

- 2D radial (FFE)
- FOV = 250 x 250 mm, voxel size = 2 x 2 x 10 mm
- 24 heart phases, 35.5 ms, 157.1 % (**100% radial Nyquist**)
- **16 velocity encodes + 1 ref**
- Kernel size = **$7 \times 7 \times 3$ ($k_x - k_y - t$)**, **$30 \times 30 \times 24$** calibration area



In-vivo - k-t rFVE



Discussion



Discussion

- Wavelet- or TV-l1 minimization seem not to be suitable for frame-by-frame reconstruction in dynamic PC-MRI
- Exploiting temporal correlations with temporal FT as a sparsifier can improve phase reconstruction accuracy
- rFVE has the potential to significantly accelerate FVE, and therefore, to assign complex flow patterns
 - Inclusion of **priors** (e.g. sparsity)

$$\underset{\mathbf{s}_c}{\operatorname{argmin}} \quad \|\mathbf{D}\mathbf{s}_c - \mathbf{s}\|_2^2 + \lambda_1 \|(\mathbf{G} - \mathbf{I})\mathbf{s}_c\|_2^2 + \sum_{j \geq 2} \lambda_j R_j(\mathbf{s}_c)$$

- **Self-gating and motion correction**

$$\underset{\mathbf{s}_c}{\operatorname{argmin}} \quad \|\mathbf{DTs}_c - \mathbf{y}\|_2^2 + \lambda \|(\mathbf{G} - \mathbf{I})\mathbf{Ts}_c\|_2^2$$