

Image Registration for Dynamic Contrast Enhanced Magnetic Resonance Image Sequences

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Registration of DCE-MRI Sequences

Example: [Video]



Objective: Remove the motion in a DCE-MRI sequence so that individual tissue points can be investigated.

Registration of DCE-MRI Sequences

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Plan A: Register all images to a fixed reference.

Registration of DCE-MRI Sequences

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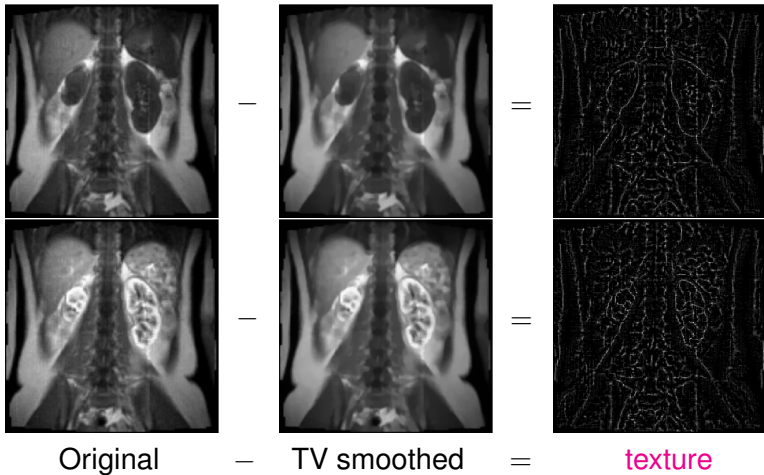


Challenges for an image similarity measure:

- ▶ Higher contrast creates **new structures** (edge based?)
- ▶ **Intensities change** within the sequence (intensity based?)
- ▶ Gradual **intensity variations** within single images (segmentation based?)

Image Similarity Measures for DCE-MRI

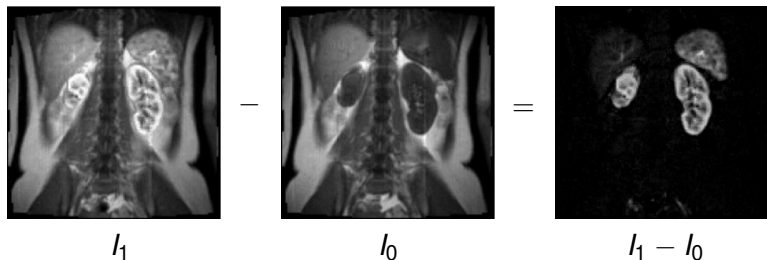
Higher contrast creates **new structures**:



Edge based similarity measure?

Image Similarity Measures for DCE-MRI

Intensities change within the sequence:



Here the patient at I_0 is a trivial displacement of the patient at I_1 .

Intensity based similarity measure?

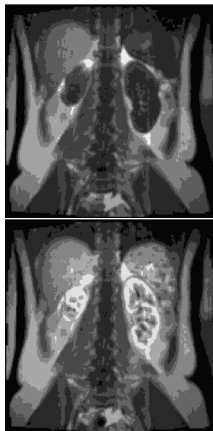
Image Similarity Measures for DCE-MRI

Gradual **intensity variations** within single images:

original
images



piecewise
constant
segmentations



Segmentation based similarity measure?

Explicit Similarity Measures

For **Sum of Squared Differences**,

$$S(l_0, l_1, u) = \int_{\Omega} |l_0 \circ (\text{Id} + u) - l_1|^2$$



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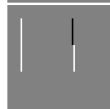


Force field in the optimality system:

vertical component:



horizontal component:



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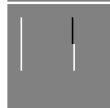


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stronger still
for *symmetric*
sum of squared
differences

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For an **edge based** measure, e.g.,

$$S(I_0, I_1, u) = \int_{\Omega} |n_0 \circ (\text{Id} + u) \times n_1|^2, \quad n_k = \nabla I_k / |\nabla I_k|$$



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Further, $I_0 \circ (\text{Id} + u)$
must start very
close to I_1 for
correct convergence

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Proposed Rescaling Measure:

$$S(l_0, l_1, u) = \int_{\Omega} |l_0 \circ (\text{Id} + u) - R[l_1]|^2$$

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$$S(I_0, I_1, u) = \int_{\Omega} |I_0 \circ (\text{Id} + u) - R[I_1]|^2$$

where R is a **local rescaling of intensities**:

$$R[I_1] = \sum_{\omega \subset \mathcal{S}_d(I_1)} p_{\omega} \chi_{\omega}, \quad p_{\omega} = \arg \min_{p \in \mathcal{P}^d} \int_{\omega} |I_0 \circ (\text{Id} + u) - p|^2$$

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and $\mathcal{S}_d(I_1)$ is a **dth degree segmentation** of I_1 : $\text{conn}(\{\omega_m\})$

$$\sum_{m=1}^M \int_{\omega_m} |q_m - I_1|^2 = \min \left\{ \{q_m\} \subset \mathcal{P}^d, \omega_m \cap \omega_n = \emptyset, \Omega = \bigcup_{m=1}^M \omega_m \right\}$$

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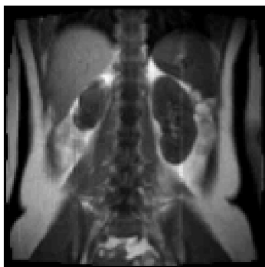
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an approximation to a **higher order Mumford Shah** functional:

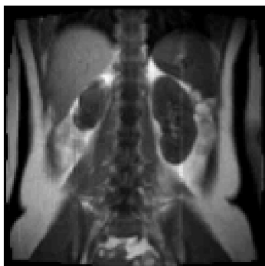
$$\min_{I, \Gamma} \int_{\Omega} |I - I_1|^2 + \alpha \int_{\Omega \setminus \Gamma} |\nabla^{d+1} I|^2 + \beta |\Gamma|$$

Segmentation of Degree d

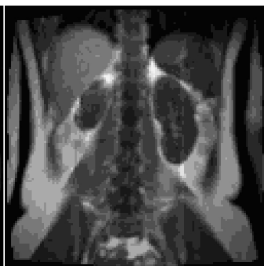


original

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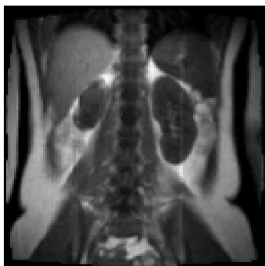


original

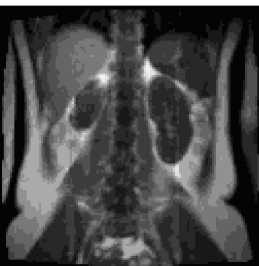


piecewise constant

Segmentation of Degree d



original

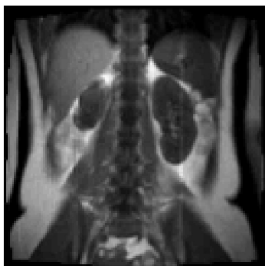


piecewise constant



piecewise linear

Segmentation of Degree d



original



piecewise constant



piecewise linear



piecewise quadratic

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[Video]

Registration Regularization

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Registration Result

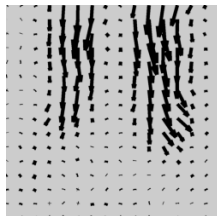
Optimality system solved by Newton's method with line search.



I_0



$I_0 \circ (\text{Id} + u)$



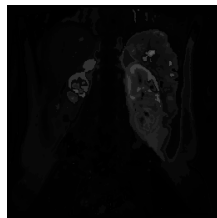
u



I_1



$R[I_1]$



$|R[I_1] - I_1|$

Result superior to those with TV regularization: [\[Video\]](#).

Kidneys are particularly motionless.

Higher Order Models: Total Generalized Variation

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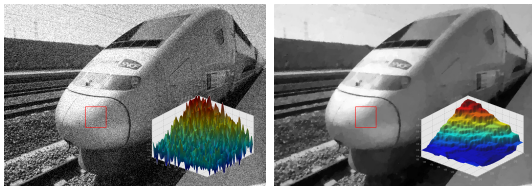
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Noisy and TV-reconstructed images:



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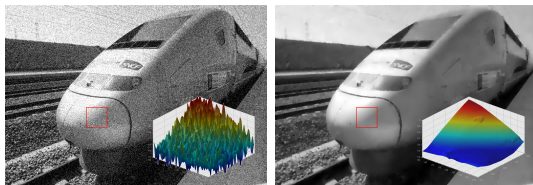
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Noisy and TGV_{α}^2 -reconstructed images: [\[Bredies, Kunisch, Pock\]](#)



Higher Order Segmentation

Forthcoming results from [Kraft]: Higher order Mumford Shah,

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Phase function (Ambrosio-Tortorelli) approximation:

$$\min_{I, \phi} = \int_{\Omega} \left\{ |I - \tilde{I}|^2 + \alpha |\nabla^{d+1} I|^2 \phi^2 + \epsilon |\nabla \phi|^2 + \epsilon^{-1} |1 - \phi|^2 \right\}$$

but ϕ is drawn to zero by varying strengths of discrete $|\nabla^{d+1} I|^2$.

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$$\min_{I, \Gamma} = \int_{\Omega} |I - \tilde{I}|^2 + \alpha \int_{\Omega \setminus \Gamma} |\nabla^{d+1} I|^2 + \beta |\Gamma|$$

Phase function (Ambrosio-Tortorelli) approximation:

$$\min_{I, \phi} = \int_{\Omega} \left\{ |I - \tilde{I}|^2 + \alpha |\nabla^{d+1} I|^2 \phi^2 + \epsilon |\nabla \phi|^2 + \epsilon^{-1} |1 - \phi|^2 \right\}$$

but ϕ is drawn to zero by varying strengths of discrete $|\nabla^{d+1} I|^2$.

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$$\begin{aligned} \min_{c_m, \chi_m} = & \int_{\Omega} \left\{ \left| \sum_{m=1}^M c_m \chi_m - \tilde{I} \right|^2 + \alpha \sum_{m=1}^M |\nabla^{d+1} c_m|^2 |\chi_m|^2 \right. \\ & \left. + \sum_{m=1}^M \left[\epsilon |\nabla \chi_m|^2 + \epsilon^{-1} |\chi_m (1 - \chi_m)|^2 \right] + \epsilon^{-1} \left| 1 - \sum_{m=1}^M \chi_m \right|^2 \right\} \end{aligned}$$

Higher Order Registration

Forthcoming results from [Fürtinger]: Counterpart formulation,

$$\min_{c_0^m, \chi_0^m, c_1^m, \chi_1^m, u} = \int_{\Omega} \left\{ \left| \sum_{m=1}^M c_0^m \chi_0^m - \tilde{l}_0 \right|^2 + \alpha \sum_{m=1}^M |\nabla^{d+1} c_0^m|^2 |\chi_0^m|^2 \right. \\ \left. + \sum_{m=1}^M \left[\epsilon |\nabla \chi_0^m|^2 + \epsilon^{-1} |\chi_0^m (1 - \chi_0^m)|^2 \right] + \epsilon^{-1} \left| 1 - \sum_{m=1}^M \chi_0^m \right|^2 \right\}$$

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Spatially Dependent Regularization

For image restoration, modify ROF model from global to local,

$$\min_{u \in BV(\Omega)} \int_{\Omega} |Du| \quad \text{subject to} \quad [w \star (I - \tilde{I})^2] \leq \sigma^2$$

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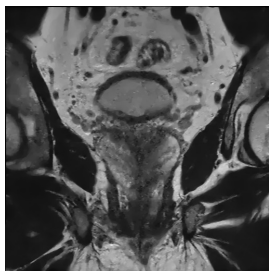
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noisy



restored



regularization

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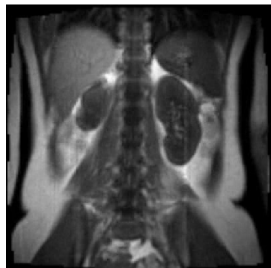
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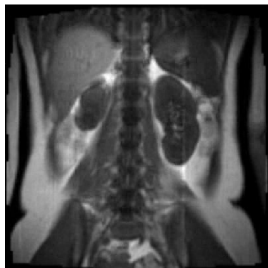
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regularization

Ergodic Sequences

Temporal averages are (locally) equal to spatial averages:



ave $t = 30 : 40$



ave $x \in 5 \times 5$

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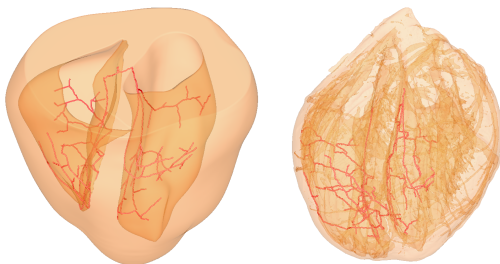
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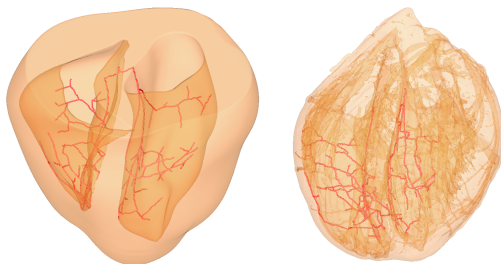
Registration of Edge Sets

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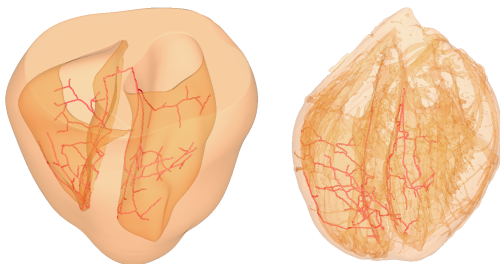


Performed using **2D slices**,

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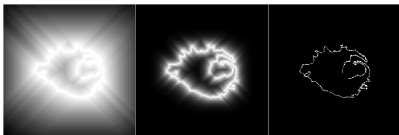
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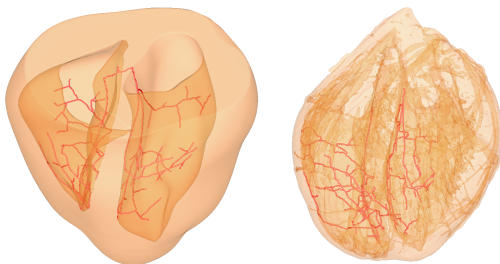
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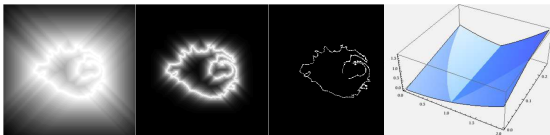


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Reducing $\epsilon \rightarrow \epsilon_0 > 0$
 $\epsilon_0 = 0 \Rightarrow \text{argmin} = 0!$



Thank you for your attention!