



# Motion management and adaptive control for treating abdominal tumors

**Baudouin DENIS de SENNEVILLE**

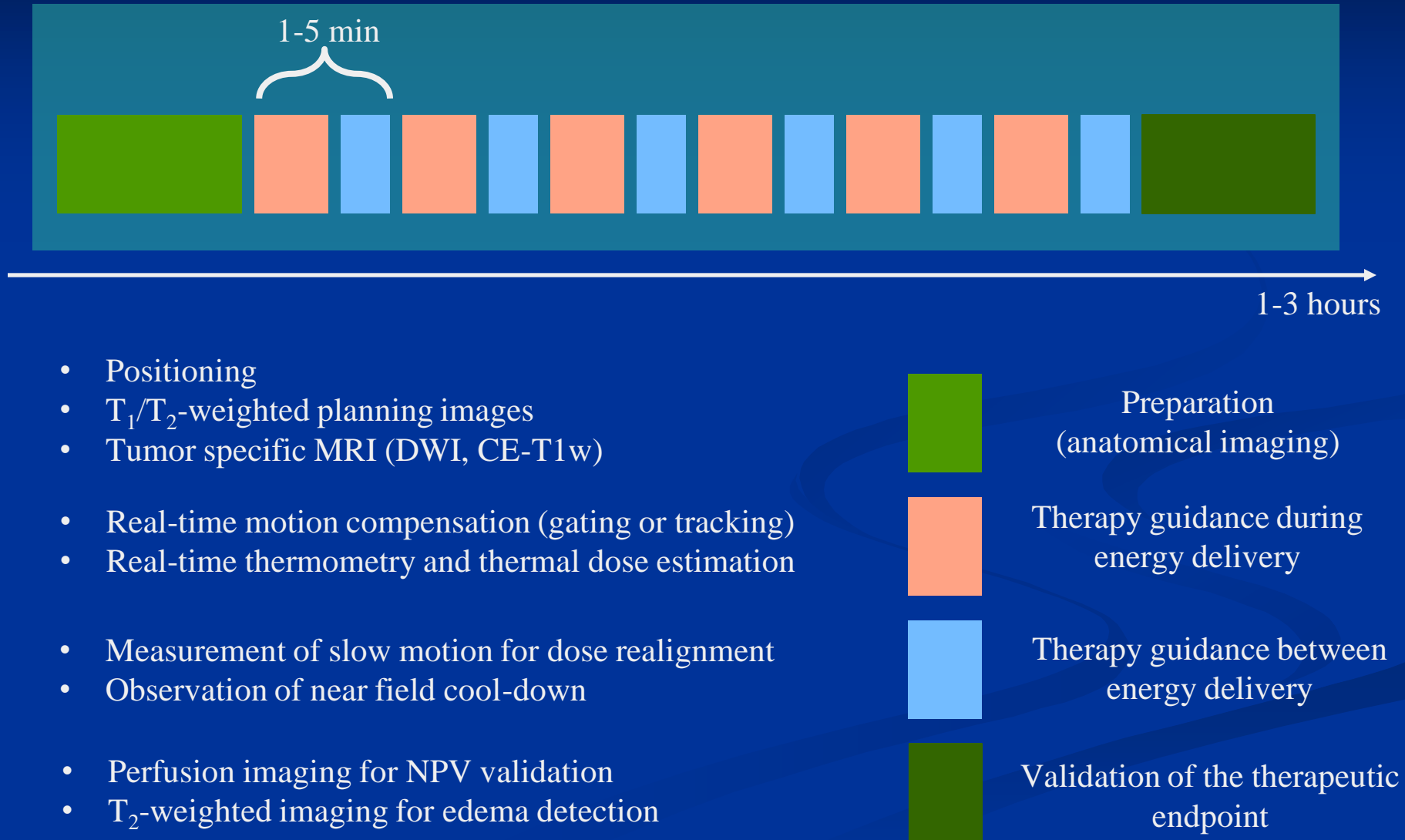
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Mathematical Institute of Bordeaux, CNRS UMR5251/University of Bordeaux, France  
Imaging Division, University Medical Center Utrecht, The Netherlands

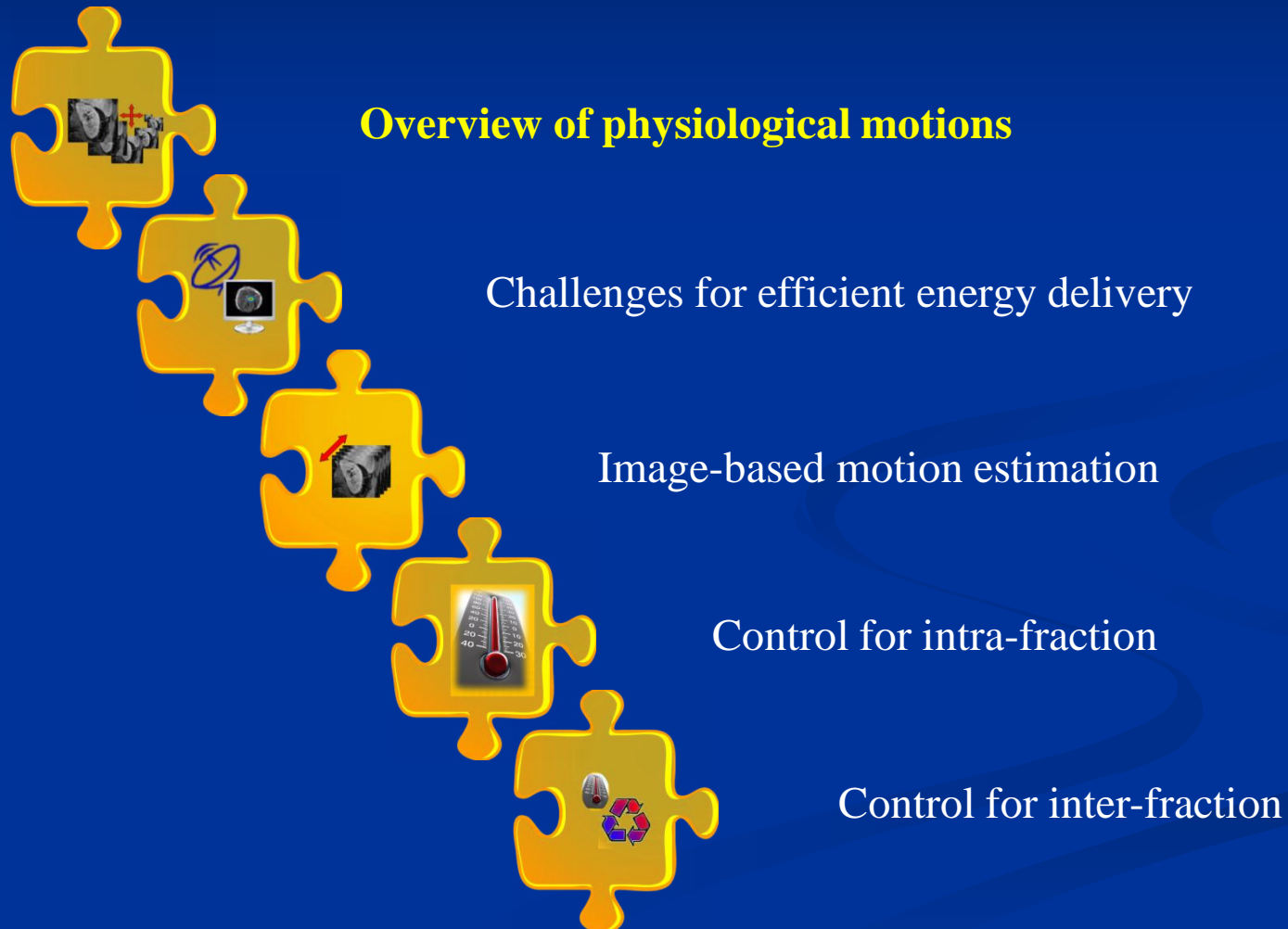
[b.desenneville@gmail.com](mailto:b.desenneville@gmail.com)

<http://www.math.u-bordeaux.fr/~bdenisde>

# Workflow of a MR-HIFU intervention



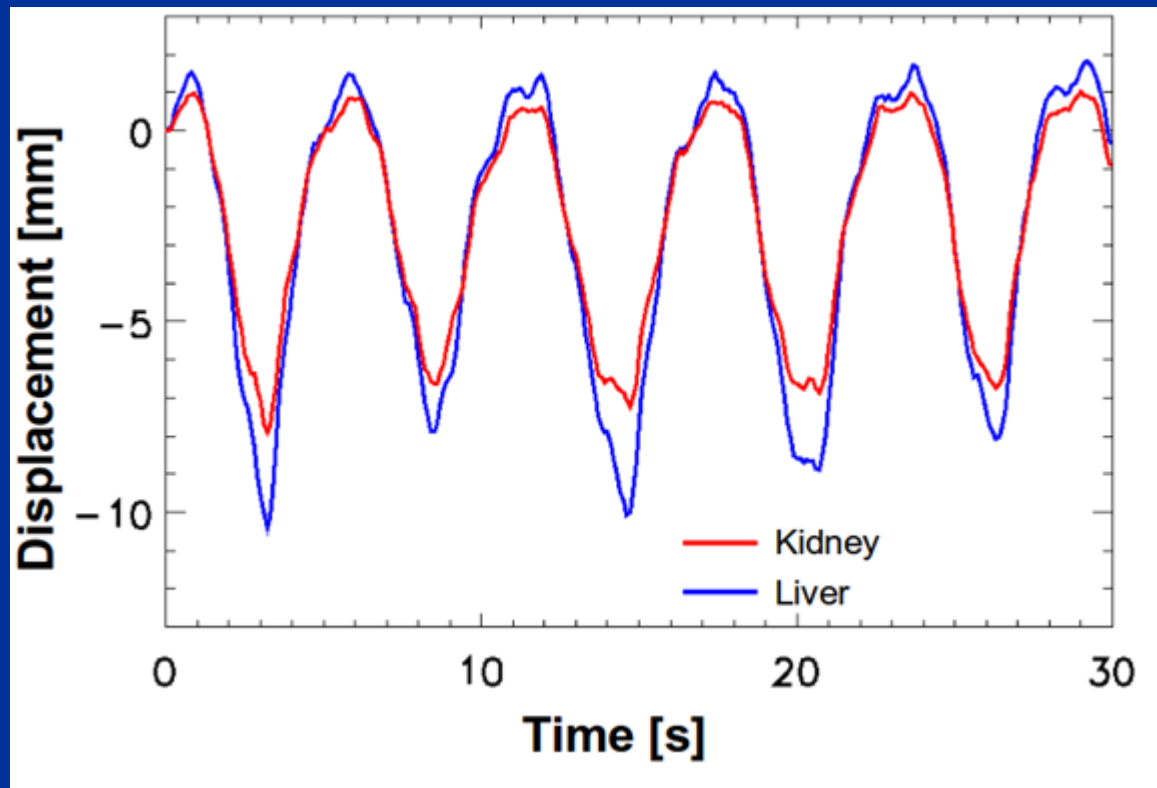
# MR-guided HIFU in abdominal organs





# Physiological motion

## Respiratory motion



# MR-guided HIFU in abdominal organs



Overview of physiological motions



**Challenges for efficient energy delivery**



Image-based motion estimation



Control for intra-fraction



Control for inter-fraction



# Challenges for efficient energy delivery in the liver

## Technical solutions

- Induced apneas [*Gedroyc et al. 2006, Kopelman et al. 2006, [...]*]
- Gating strategies [*Vigen 2003, Wu et al. 2005, Cornelis et al. 2011, [...]*]
- Real-time beam steering strategies [*de Zwart et al. 2001, de Senneville et al. 2007, Auboiroux et al. 2012 [...]*]

## Complete Respiratory Depression



# MR-guided HIFU in abdominal organs



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# Image-based motion estimation

- **General** formulation

$$\hat{T} = \underset{T}{\operatorname{argmin}} \underbrace{S}_{\text{Metric}}(I_{ref}, \underbrace{T}_{\text{Transformation}}(I_{cur}))$$

- **Elastic** organ deformation (voxelwise estimation)
  - Low **computation time**
  - Low number of **control parameters**

- **Inverse problem** solved throw a **variational** approach

$$E(\vec{V}) = \int_{\Omega} \underbrace{D(\vec{V})}_{\text{Data fidelity}} + \underbrace{\alpha}_{\text{weigh}} \underbrace{R(\vec{V})}_{\text{Regularization}} d\vec{r}$$

- *Optical Flow algorithms*
- *[Horn&Schunck, 1981]*



# Image-based motion estimation

## Optical flow algorithm

$$E(\vec{V}) = \int_{\Omega} \left| \vec{\nabla} I \cdot \vec{V} + I_t I \right| + \alpha \left\| J(\vec{V}) - 1 \right\|_2^2 d\vec{r}$$

- **Data fidelity term :**

$$I_t + \vec{V} \cdot \vec{\nabla} I = 0$$

$$D(\vec{V}) = \left| I_t + \vec{V} \cdot \vec{\nabla} I \right|$$

- *Transport equation*
- *[Horn&Schunck, 1981]*
- *[Zachiu et al. PMB 2016]*

- **Regularization term :**

$$\vec{\nabla} \cdot \vec{V} = 0$$

- *Incompressibility of the tissue*

$$R(\vec{V}) = \left\| J(\vec{V}) - 1 \right\|_2^2$$

- *Continuum mechanics :  
Incompressible material subjected to an external force*
- *[Zachiu et al. PMB 2018]*







# Image-based motion estimation

## Variational approaches

$$E(\vec{V}) = \int_{\Omega} \overset{\text{Data fidelity}}{D(\vec{V})} + \underbrace{\alpha}_{\text{weight}} \overset{\text{Regularization}}{R(\vec{V})} d\vec{r}$$

- *Multi-resolution* approach
- *Iterative refinement* approach
- *Deep-learning* approach
- *Mono-/Multi- modal* registrations

	 <b>Matlab CPU</b>	 <b>C++ CPU</b>	 <b>C++ 8 CPUs</b>	 <b>CUDA GPU</b>
<b>Mono-modal 128×84</b>	<b>&gt; 10 s</b>	<b>300 ms</b>	<b>50 ms</b>	<b>&lt; 10 ms</b>
<b>Multi-modal 256×256×64</b>	<b>&gt; 1h</b>	<b>600 s</b>	<b>100 s</b>	<b>~ 20 s</b>

# MR-guided HIFU in abdominal organs



Overview of physiological motions



Challenges for efficient energy delivery



Image-based motion estimation

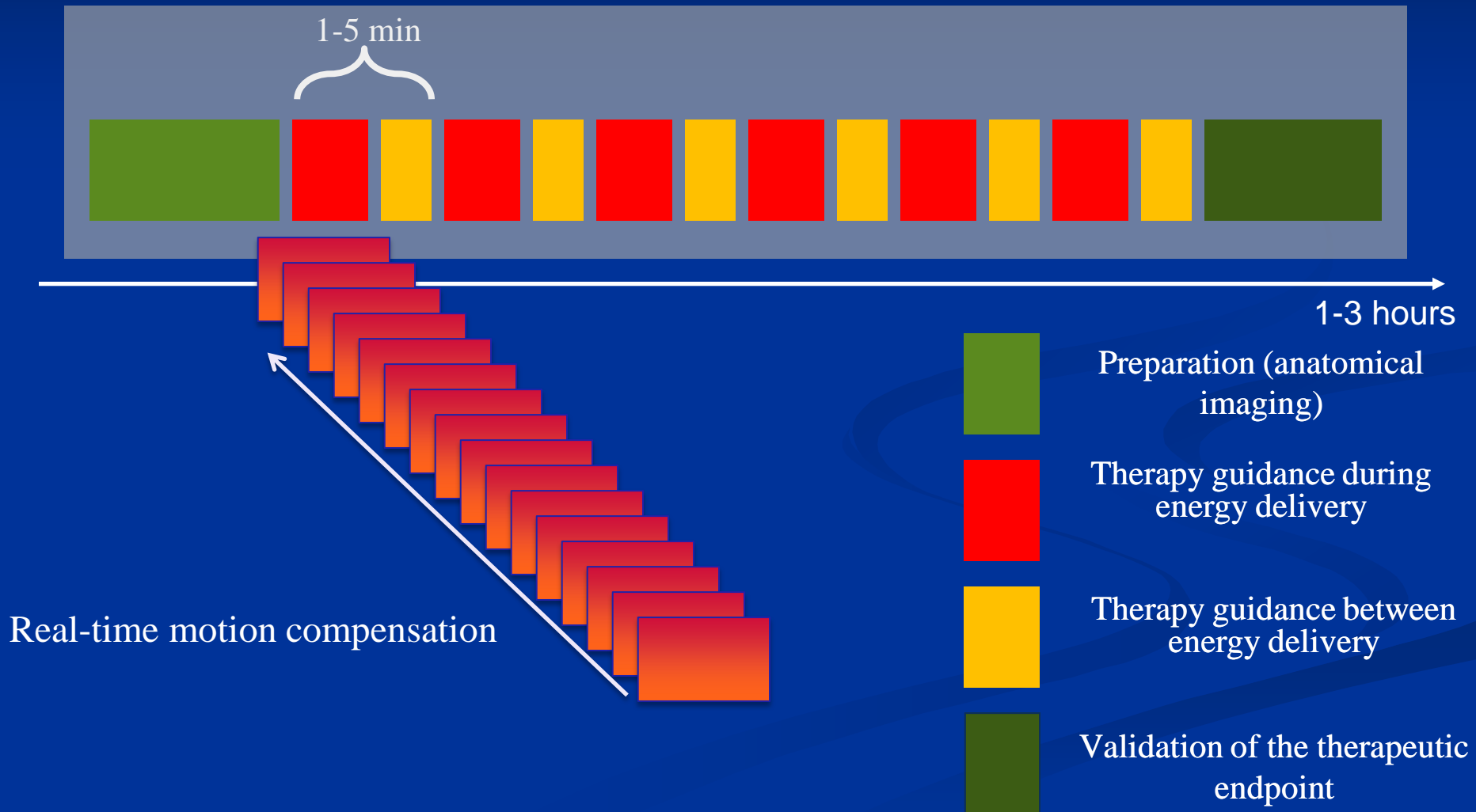


**Control for intra-fraction**



Control for inter-fraction

# Fast 2D motion correction





# Thermometry in the presence of physiological motion

**Before** hyperthermia

**During** hyperthermia

Breathing cycle



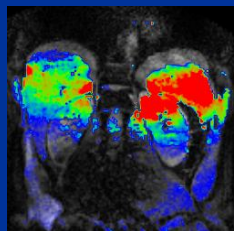
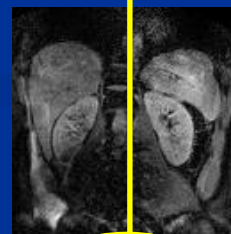
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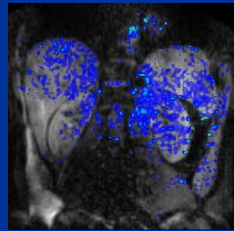
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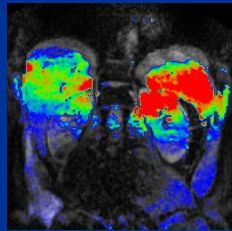
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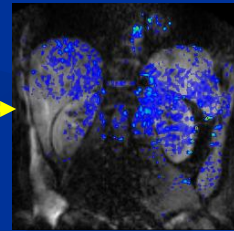
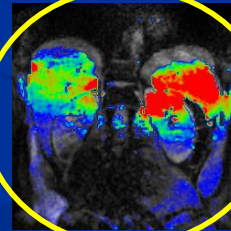
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**Learning** temperature perturbation  
with motion

# MR-guided HIFU in abdominal organs



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Image-based motion estimation



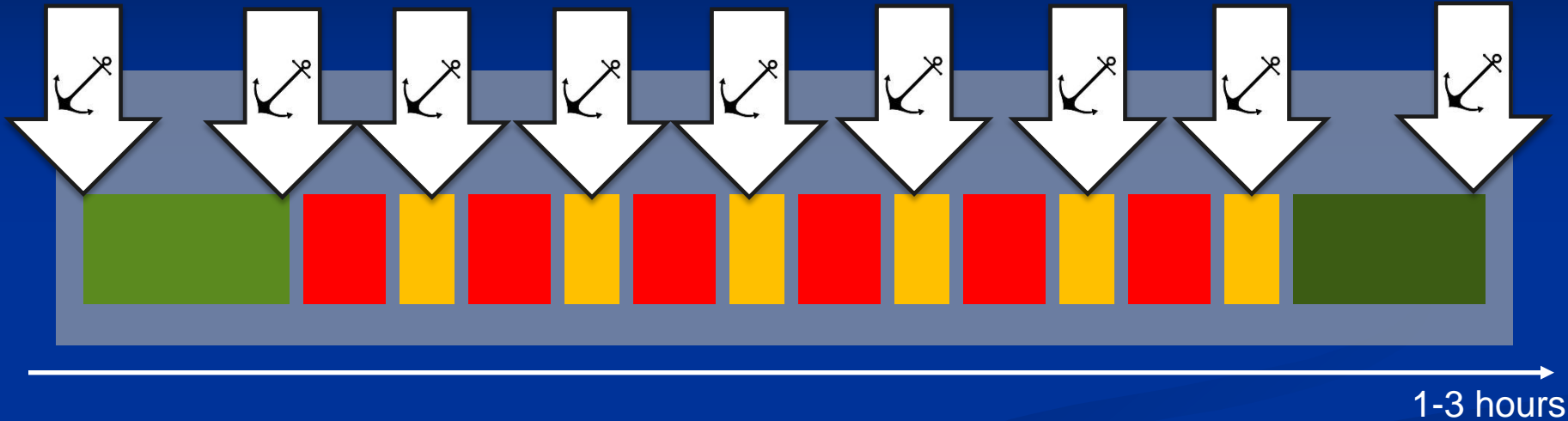
Control for intra-fraction







**Control for inter-fraction**



# « Slow » 3D Motion correction for abdominal HIFU

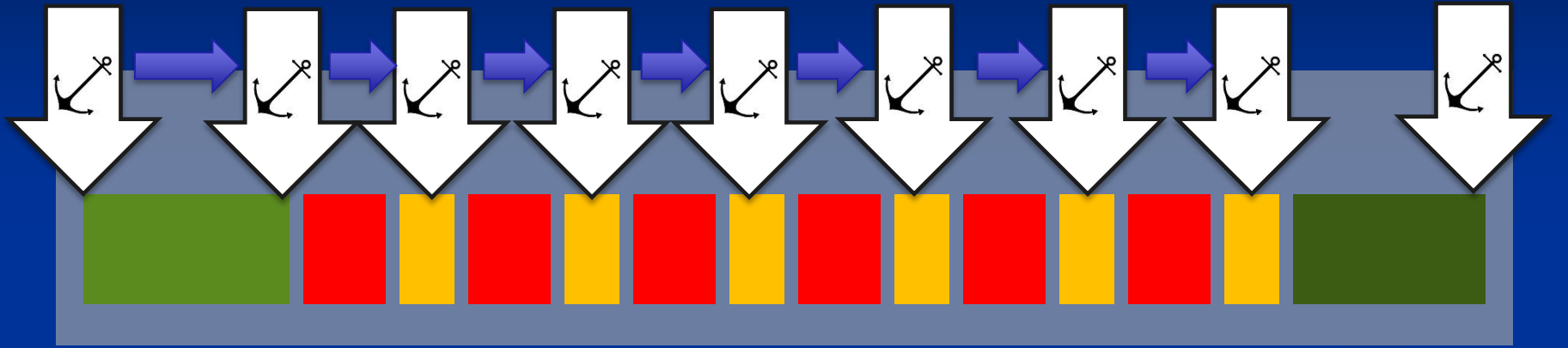


3D anchor images are periodically obtained and compared to a reference.

-  Preparation (anatomical imaging)
-  Therapy guidance during energy delivery
-  Therapy guidance between energy delivery
-  Validation of the therapeutic endpoint







# « Slow » 3D Motion correction for abdominal HIFU



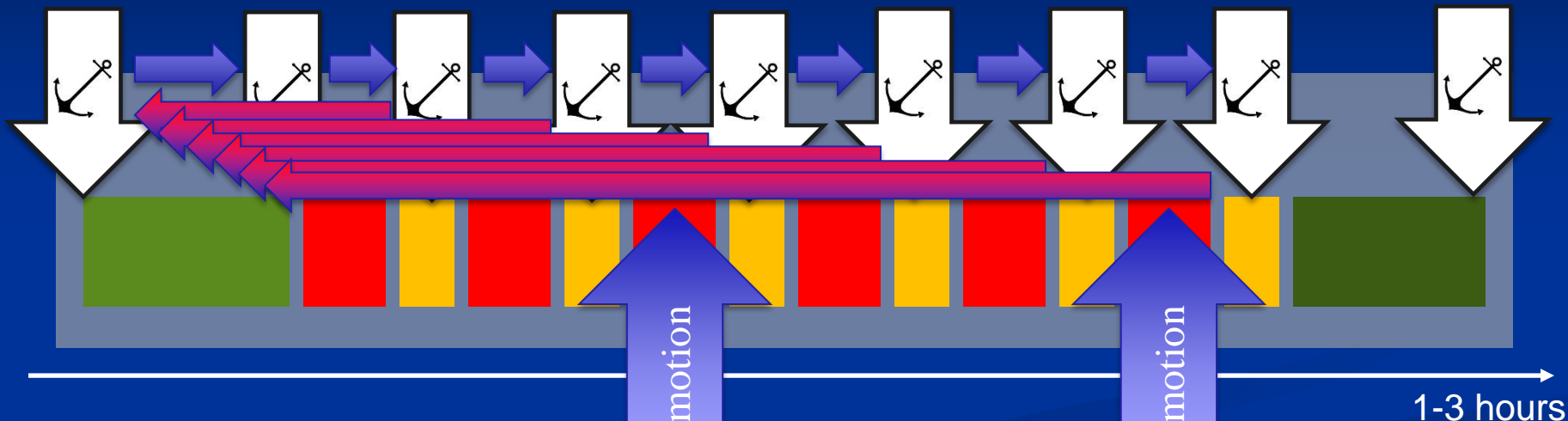
1-3 hours

Propagate the initial treatment plan down the flow of the motion  
(for the RT-community: “virtual couch shift”)

-  Preparation (anatomical imaging)
-  Therapy guidance during energy delivery
-  Therapy guidance between energy delivery
-  Validation of the therapeutic endpoint



# Experimental validation: In-Vivo Ablation of a Porcine Liver



Propagate the initial treatment plan down the flow of the motion  
(for the RT-community: “virtual couch shift”)

Project and accumulate the currently delivered thermal dose on the  
initial treatment plan “upstream” the flow of the motion  
(for the RT-community: “Continuous dose accumulation”)

- 1-3 hours
- Preparation (anatomical imaging)
- Therapy guidance during energy delivery
- Therapy guidance between energy delivery
- Validation of the therapeutic endpoint

# Concluding remarks

- **Gating** strategies can be easily implemented in MR-guided HIFU
- MRI allows detailed intra- and inter- procedure **motion tracking** of the order of 1 mm
- A framework has been developed for **3D** correction of (slow) **peristaltic** motion and **2D** correction for **respiratory** motion
- MR motion tracking can be combined with **MR thermometry**
- MR **PRFS thermometry** can be adapted for moving organs
- Further technical progress is needed to reach ablation rates that allow **widespread clinical applications**