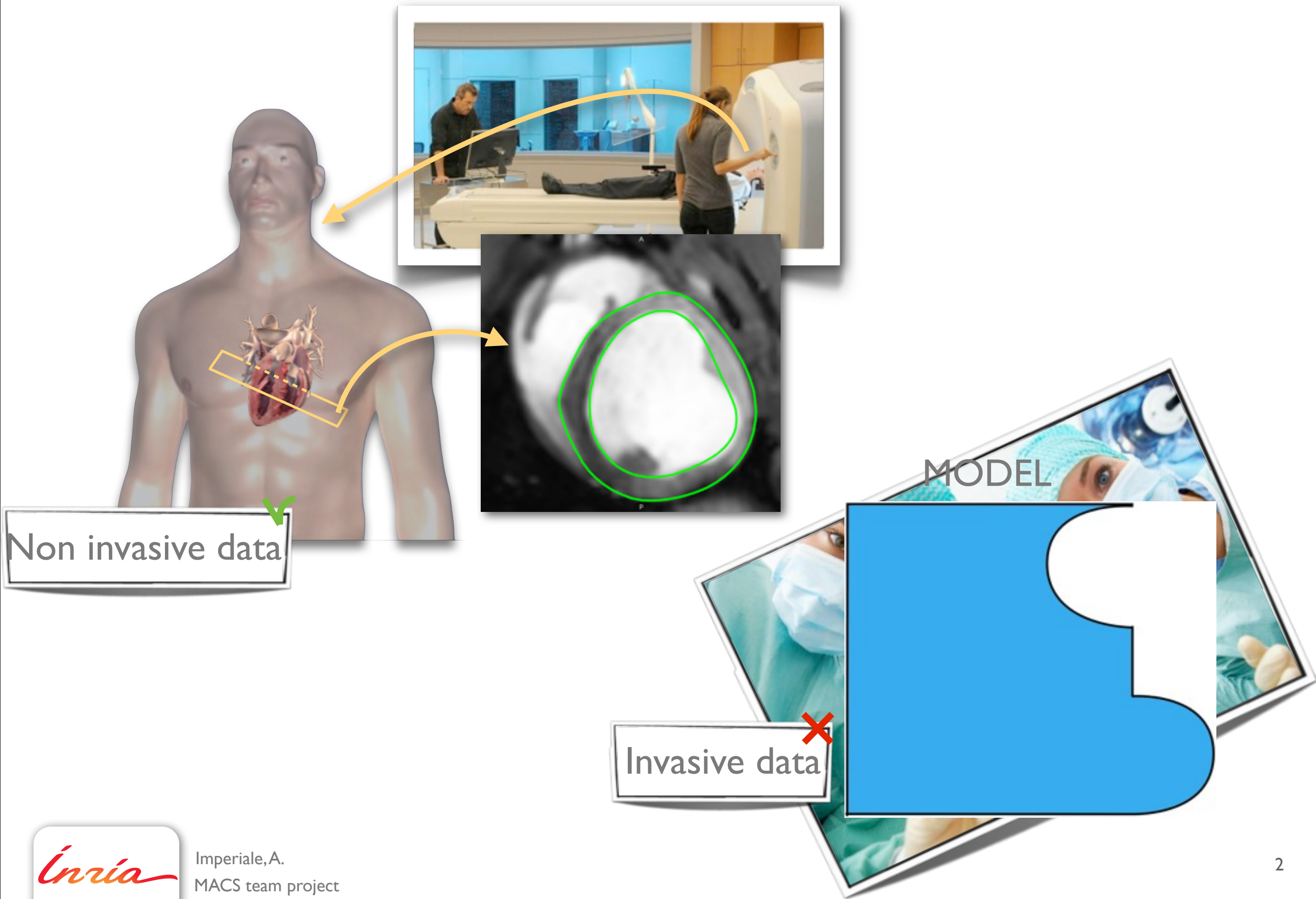


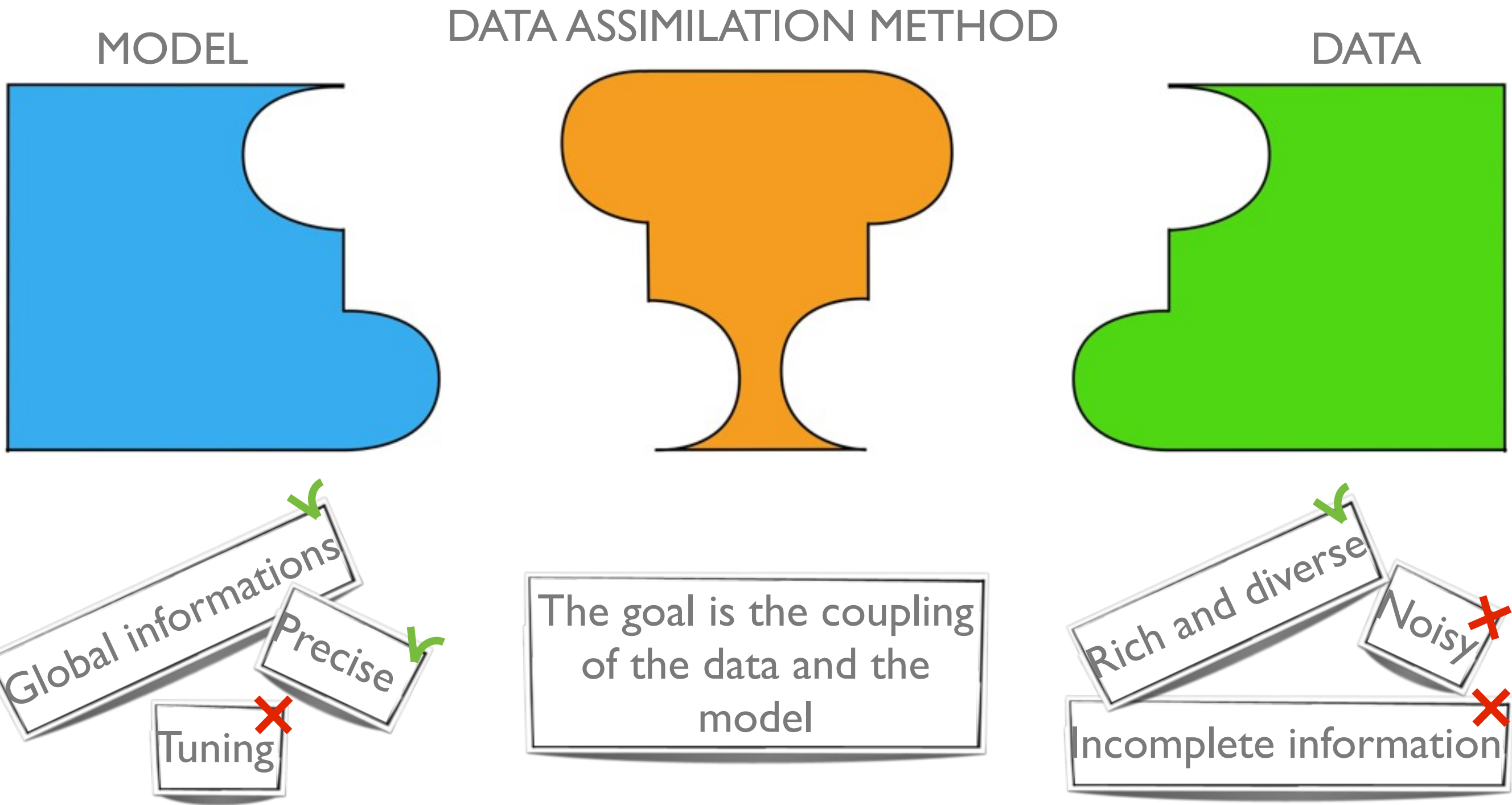
# ***Concepts of data assimilation in cardiac modeling***

***MACS team project, INRIA Saclay.***

# Motivation of data assimilation for a cardiac model



# “Have you said data assimilation ?”



# Outline of the presentation

## MODEL



- A little bit of physiology
- Overview of a cardiac biomechanical model
- Time and spacial discretization

## DATA



- Magnetic Resonance Images
- Cine and Tagged-MRI processing
- Visible displacements

## DATA ASSIMILATION

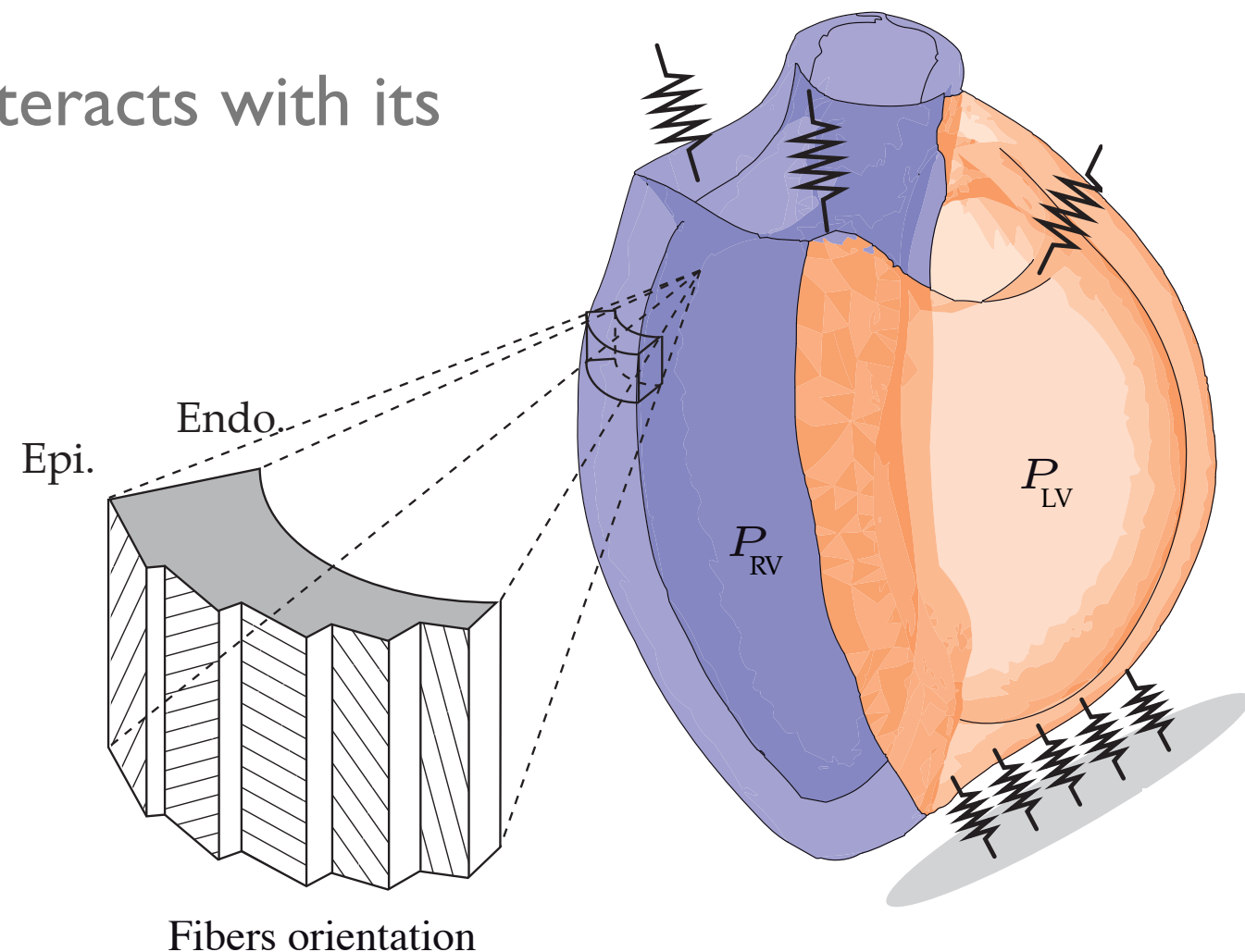


- Principle of data assimilation
- Assimilating data from images
- Application on a real case

# A little bit of physiology

- Responsible of the blood circulation. Two pumps :
  - Right ventricle : drives blood to the lungs.
  - Left ventricle : drives blood to the rest of the body.
- The heart material is composed with fibers sensitive to electrical activation.
- As the heart is deforming it interacts with its surrounding.

How is the heart  
deforming in  
time ?





# A cardiac biomechanical model

Non linear elastic model

$$\ddot{y} - \text{div}(\sigma(y)) = f \quad \& \quad \text{Boundary conditions}$$

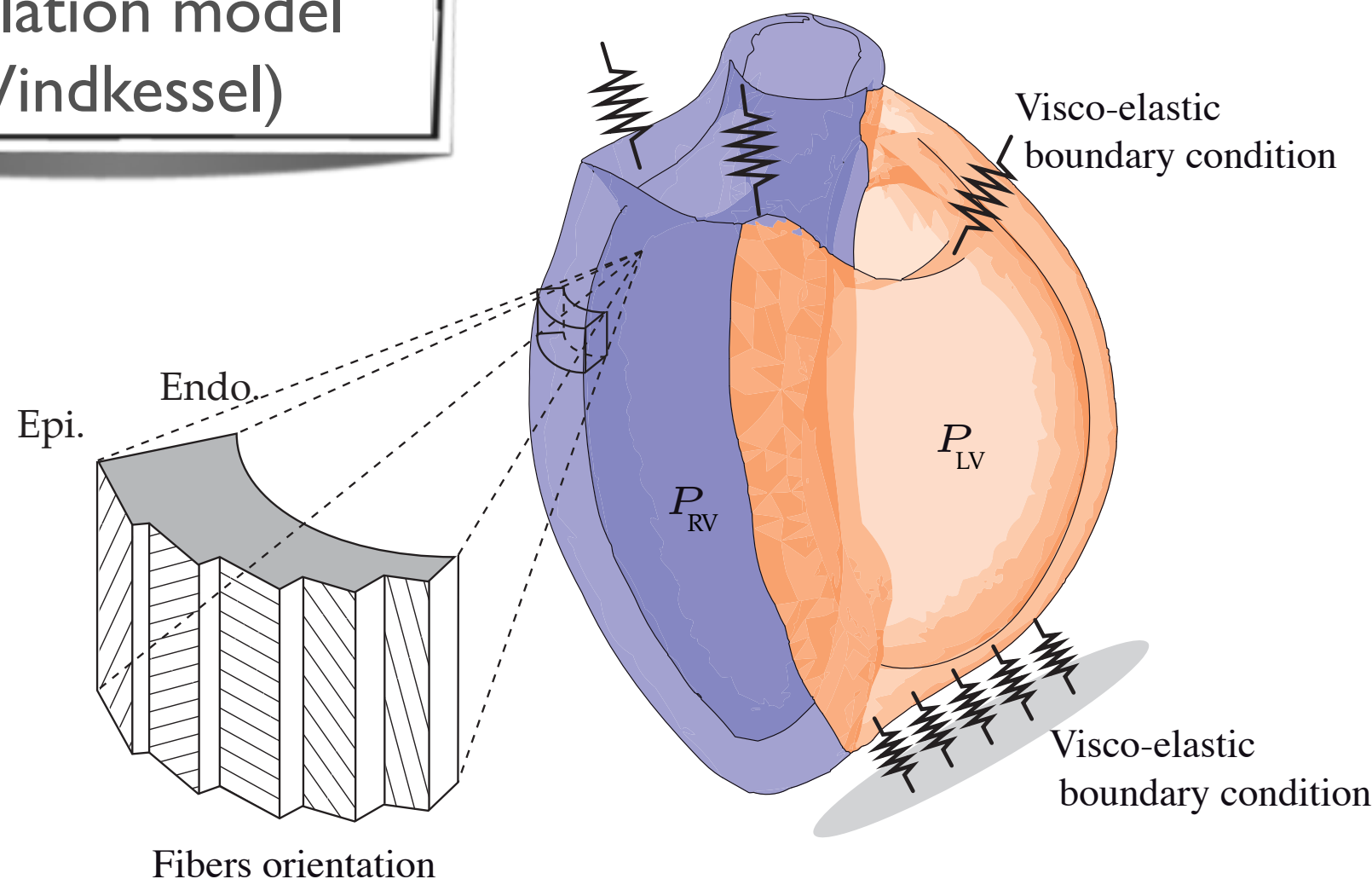
Constitutive law :

Passive anisotropic behavior

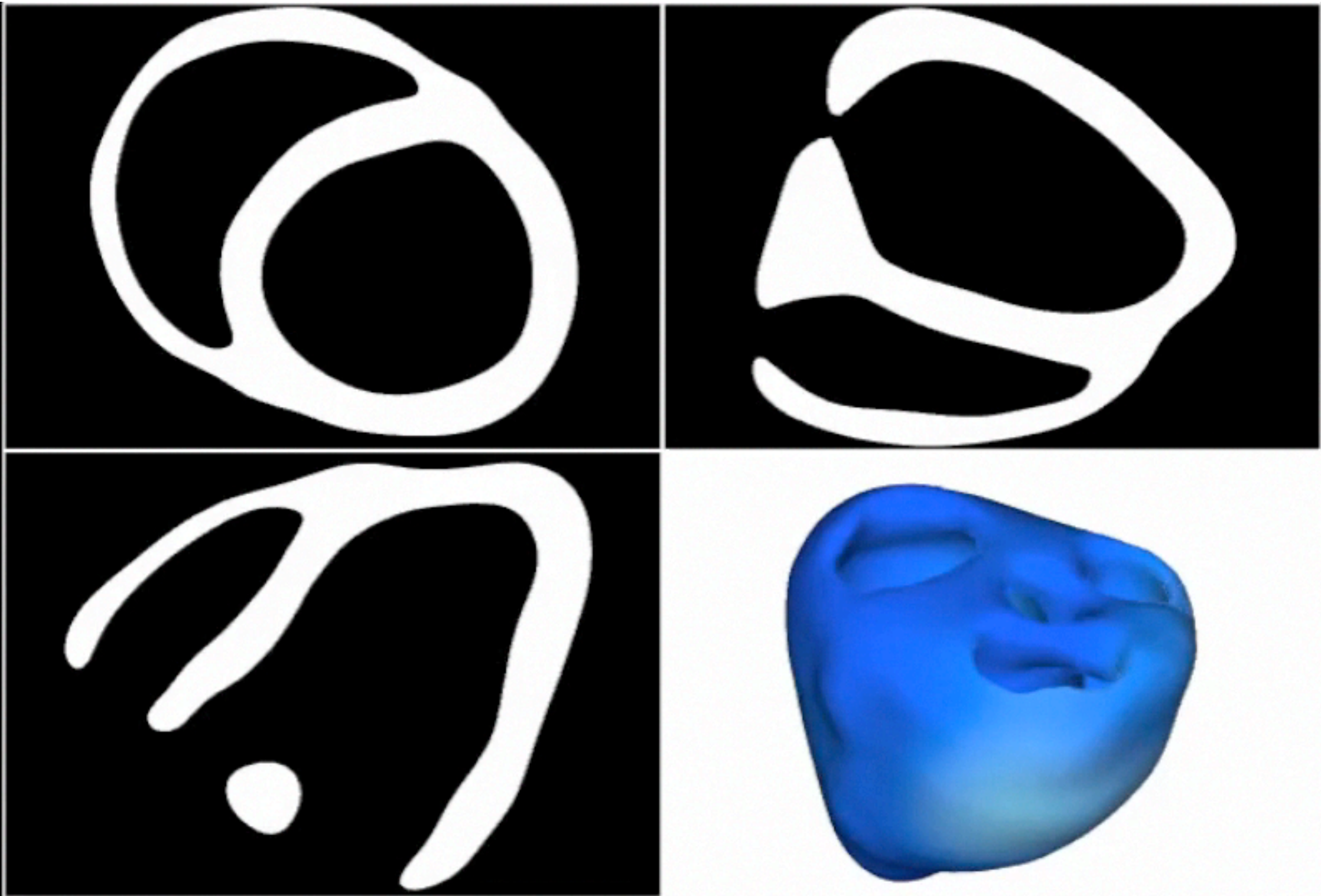
Pressure managed with simplified valve and circulation model (Windkessel)

&

Active anisotropic behavior



# Time and spacial discretization



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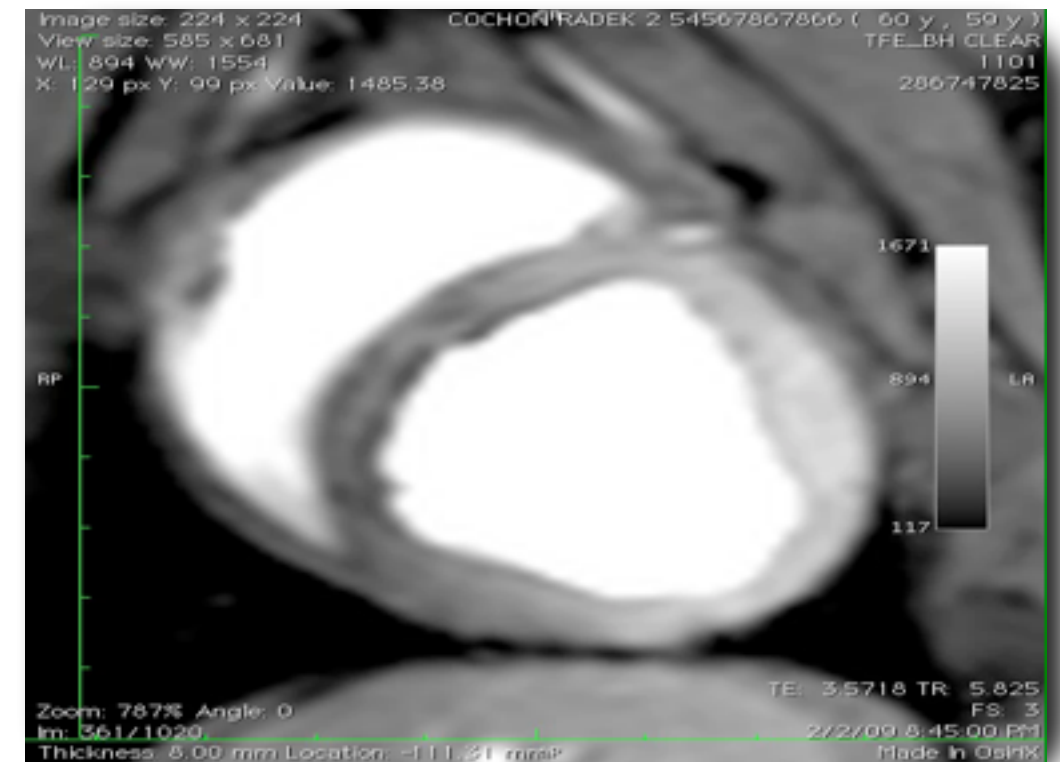


# Magnetic resonance images

## cine - MRI

Magnetization alignment of the heart tissue.

Expected visibility : Internal and external surface of the heart



## tagged - MRI

Positioning a regular grid onto the heart wall.

Expected Visibility : Intramural deformation

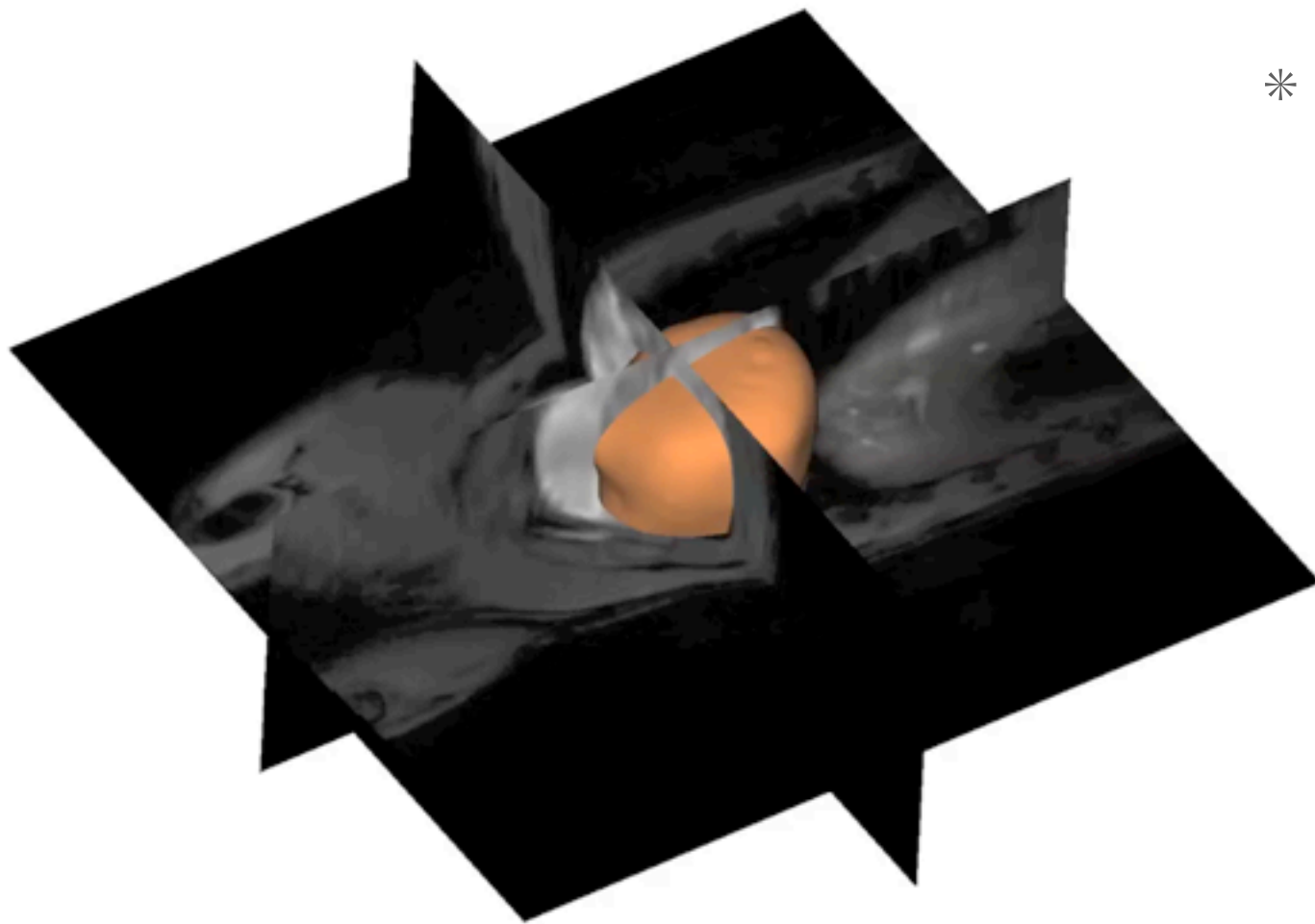


# Cine-MRI processing



3D + time

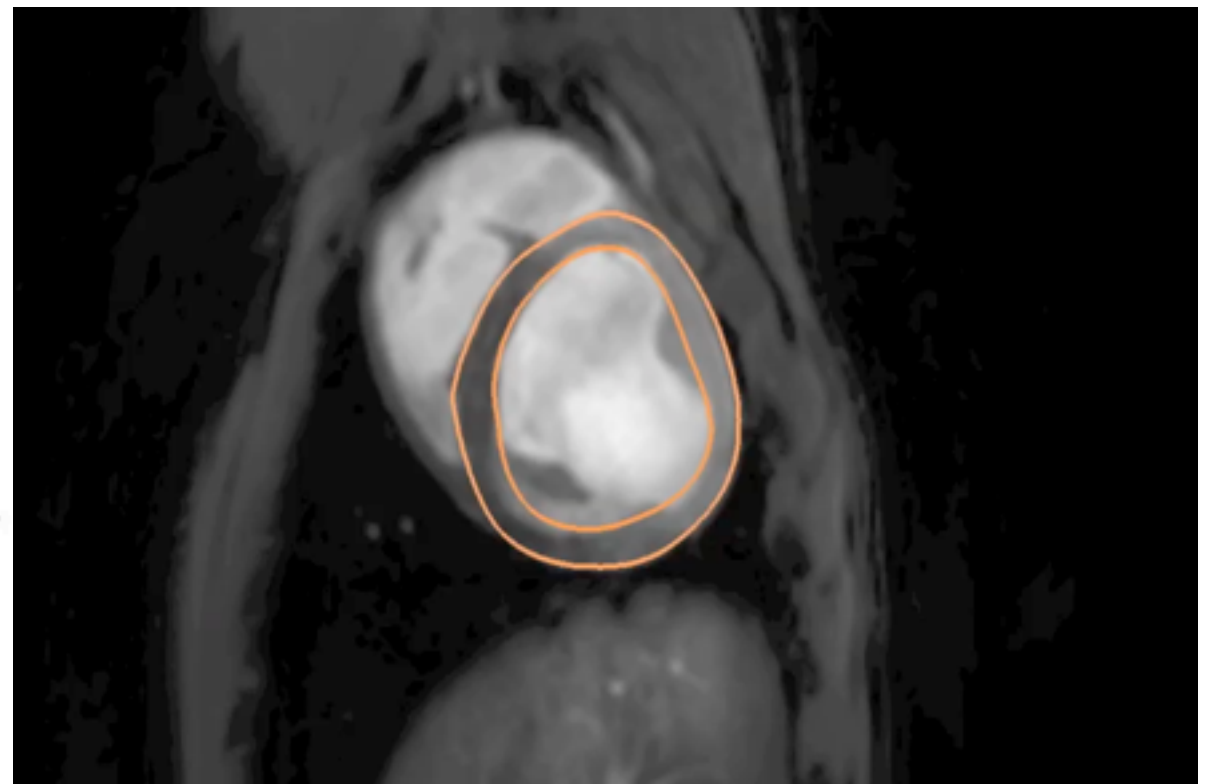
internal and external surfaces



2D + time

internal and external contours

\*

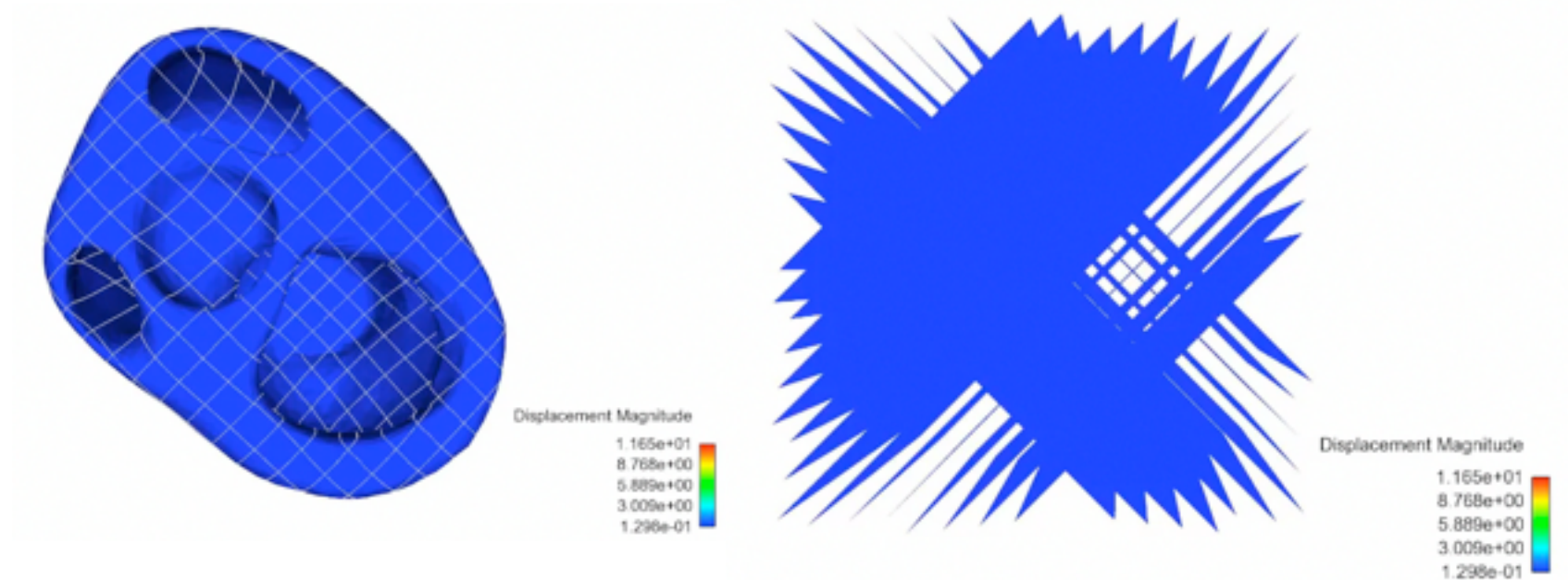


*\*Manual segmentation of cine-MRI by Radomir Chabiniok, King's College London, UK.*

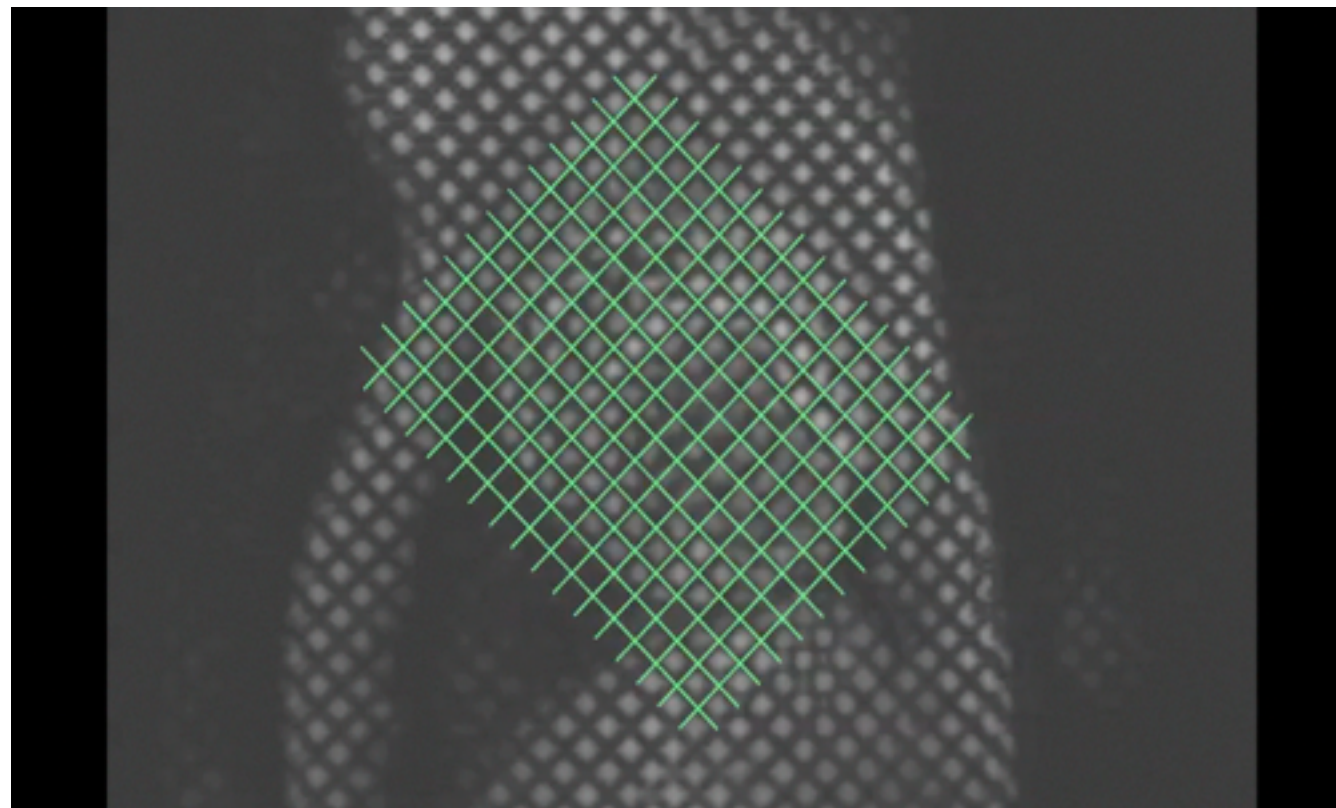


# Tagged-MRI processing

3D + time  
deforming tag planes



2D + time  
deforming tag grid



\*In collaboration with Patrick Claryss, INSA Lyon.





# Tagged-MRI processing : visible displacement



- Tracking deforming patterns
- Provides only visible displacements on the image plane.
- Displacement in the transversal direction are invisible

*\*In collaboration with Patrick Claryss, INSA Lyon.*

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## DATA ASSIMILATION

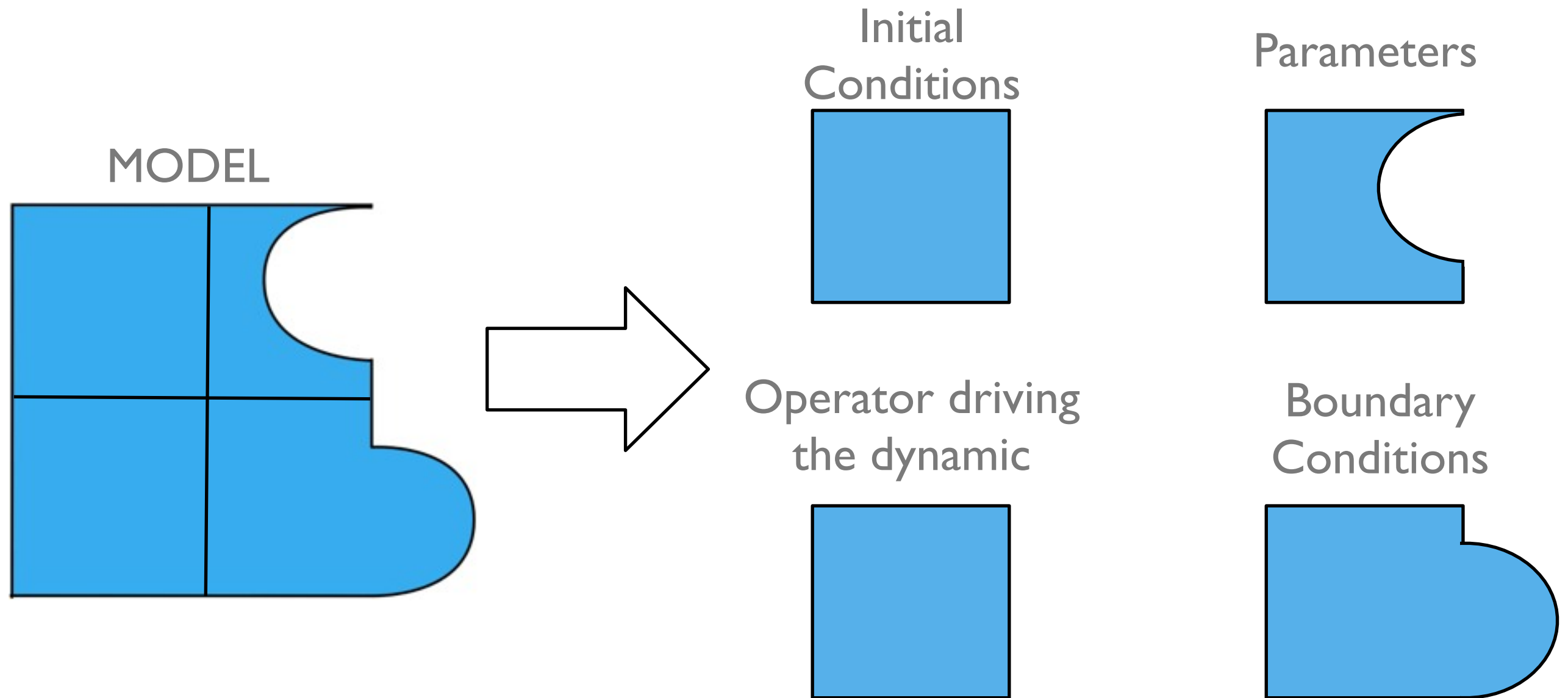


- Principle of data assimilation
- Assimilating data from images
- Application on a real case



# Principle of Data Assimilation

“Where are the errors coming from ?”




Can we use available informations to reduce these errors ?

# Principle of Data Assimilation

$$\mathbf{x} = \begin{pmatrix} y \\ \dot{y} \end{pmatrix}$$

MODEL




$$\dot{\mathbf{x}}(t) = \mathbf{A}(\mathbf{x}(t))$$

$$\mathbf{x}(0) = \mathbf{x}_0 + \zeta_{\mathbf{x}}$$

How to correct  
this error ?

DATA



$$z(t) = \mathbf{H}\mathbf{x}(t)$$

Observation  
operator

$$\hat{\mathbf{x}}(t)$$

Estimation of  
the real system

?

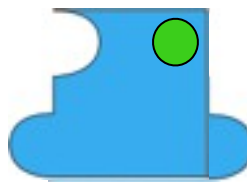
1st step

Use the observation operator to  
reach the data space.

2nd step

Filter into the dynamic the differences between the  
data and the observed estimation.

OBSERVED  
ESTIMATION





$$\mathbf{H}\hat{\mathbf{x}}(t)$$

ESTIMATION = MODEL + FILTER

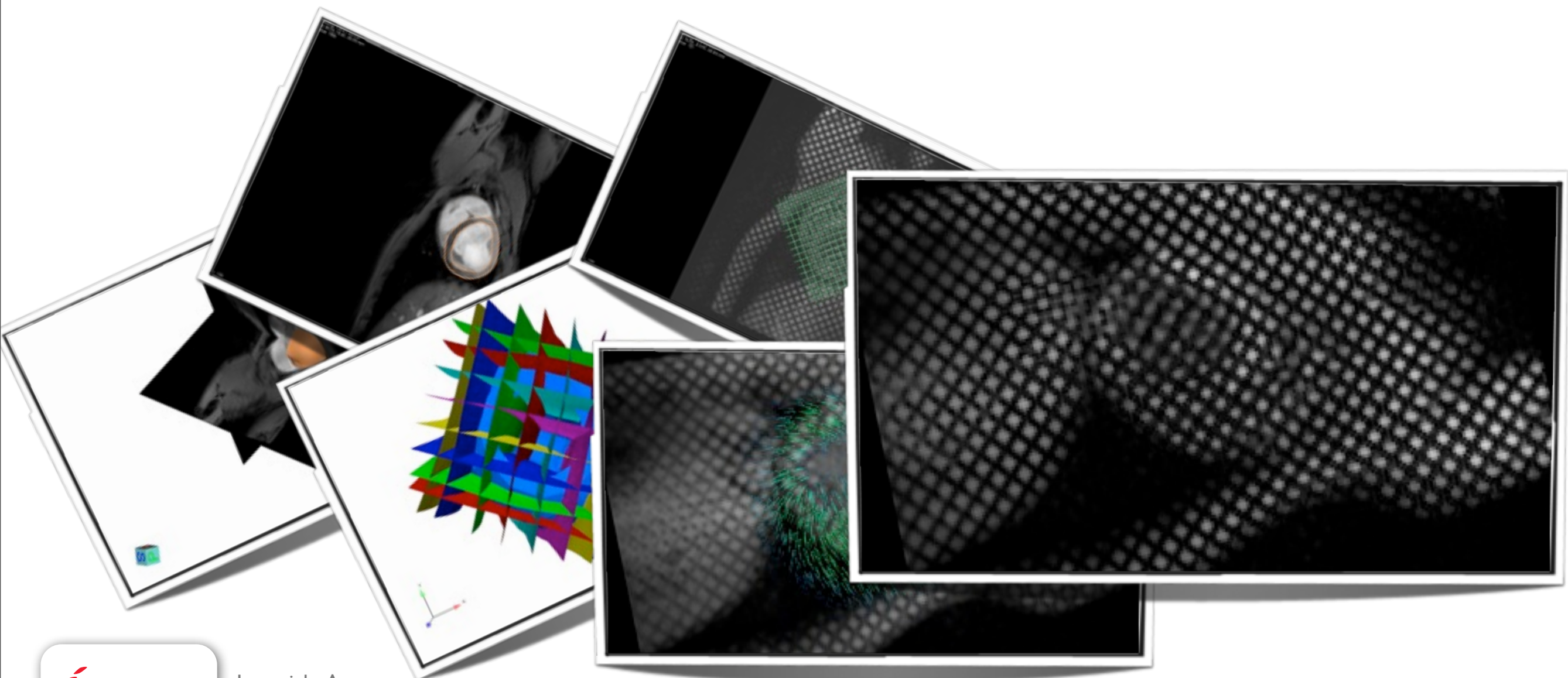
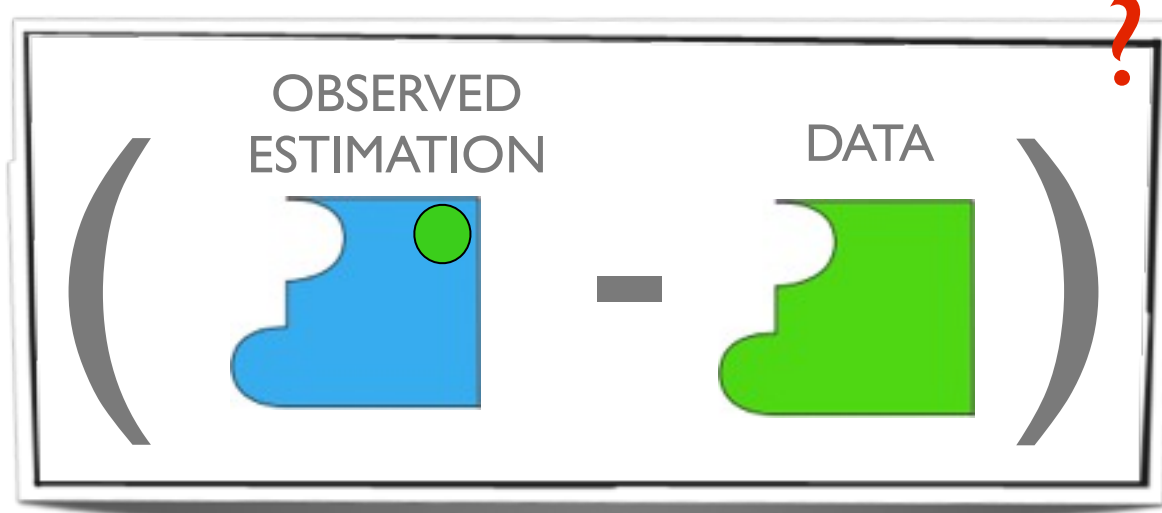
$$\dot{\hat{\mathbf{x}}}(t) = \mathbf{A}(\hat{\mathbf{x}}(t)) + \mathbf{K} \left( \mathbf{H}\hat{\mathbf{x}}(t) - z(t) \right)$$

OBSERVED ESTIMATION DATA

Filtering approach is not the only way ...

# Assimilating data from images



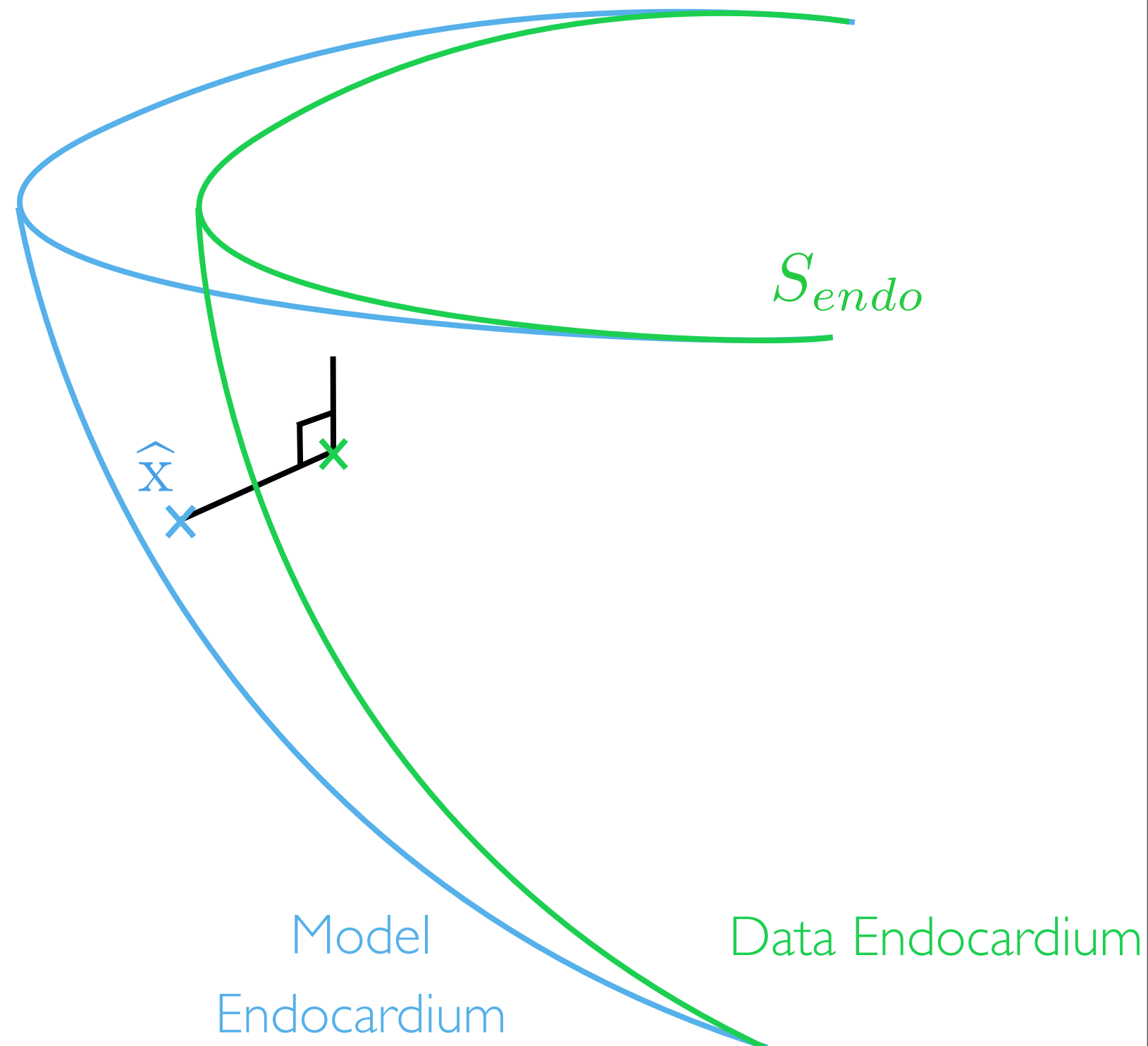
# Assimilating data from images

The simple case of cine-MRI segmentation :

$$\text{dist}(\hat{x}, S_{endo})$$

Signed distance  
between the model and  
the data

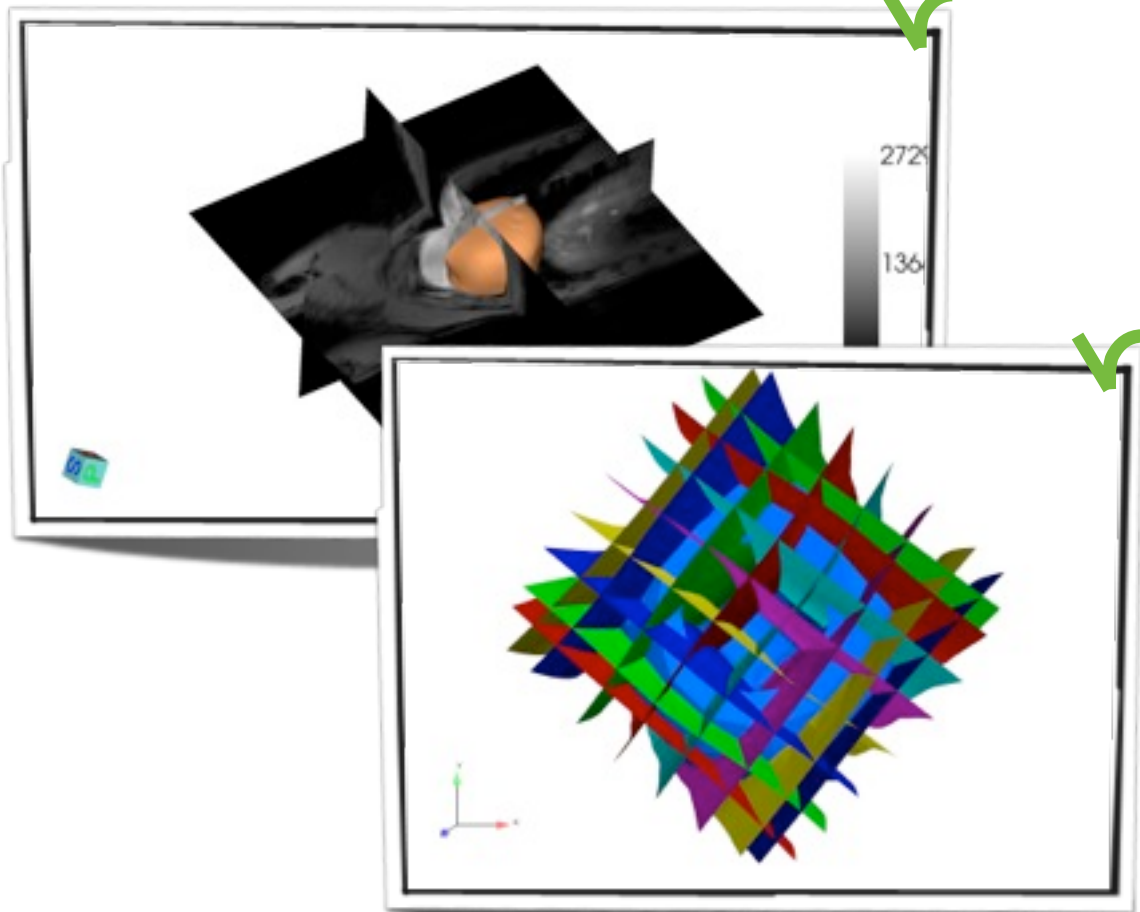
Non linear observation  
operators



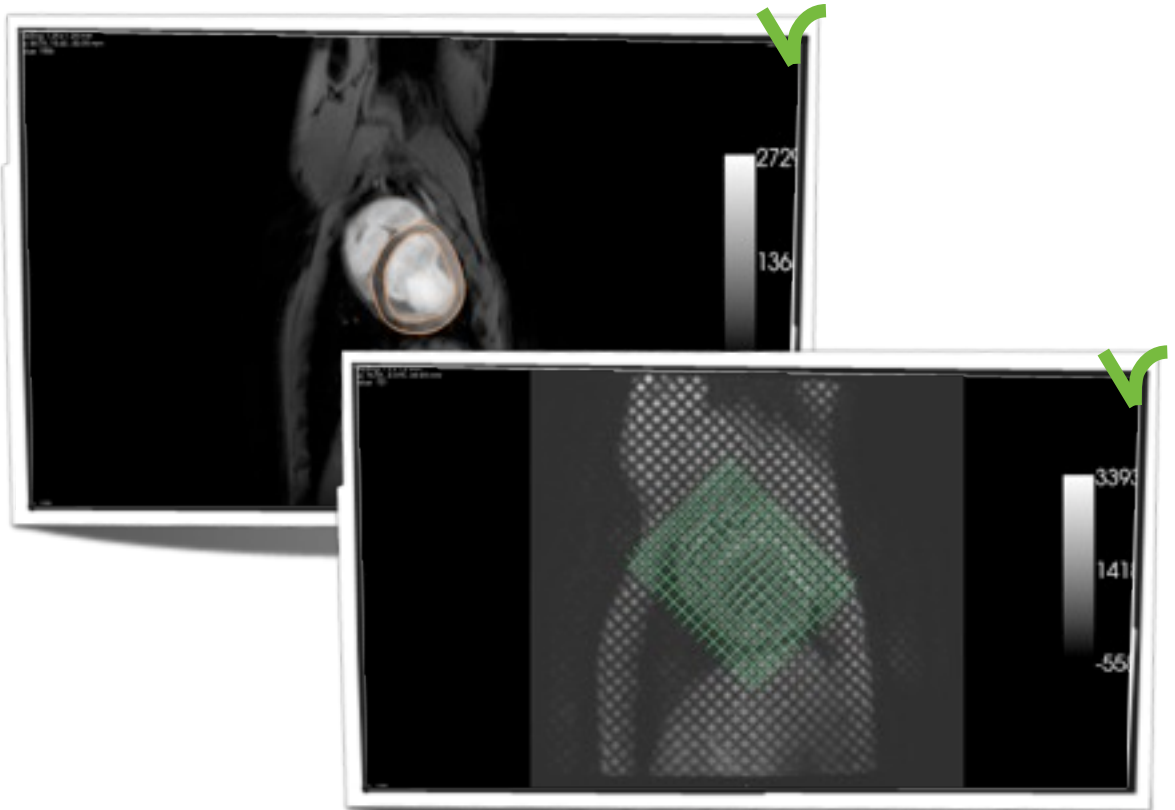


# Assimilating data from images

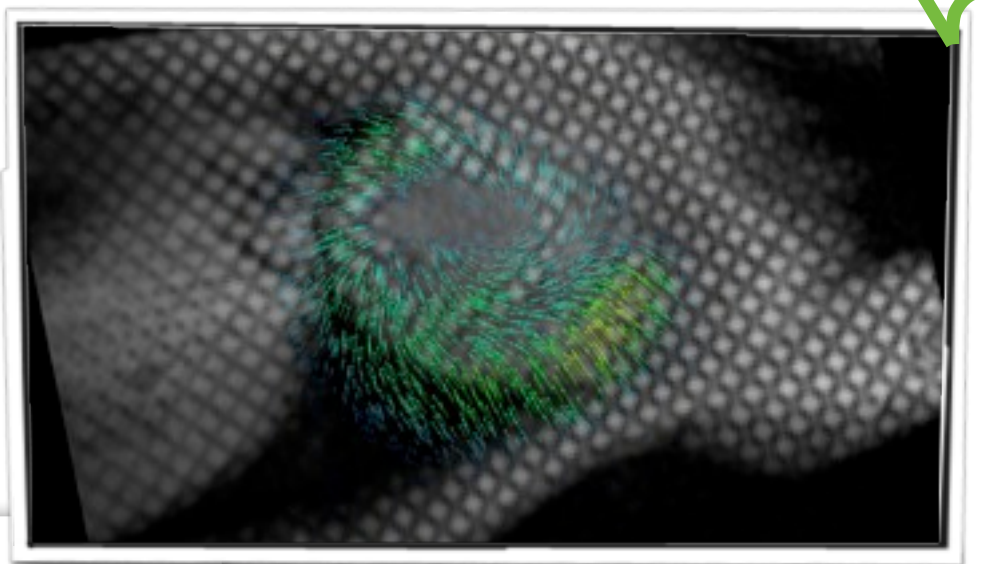
Signed distance



Signed distance with spatial interpolation



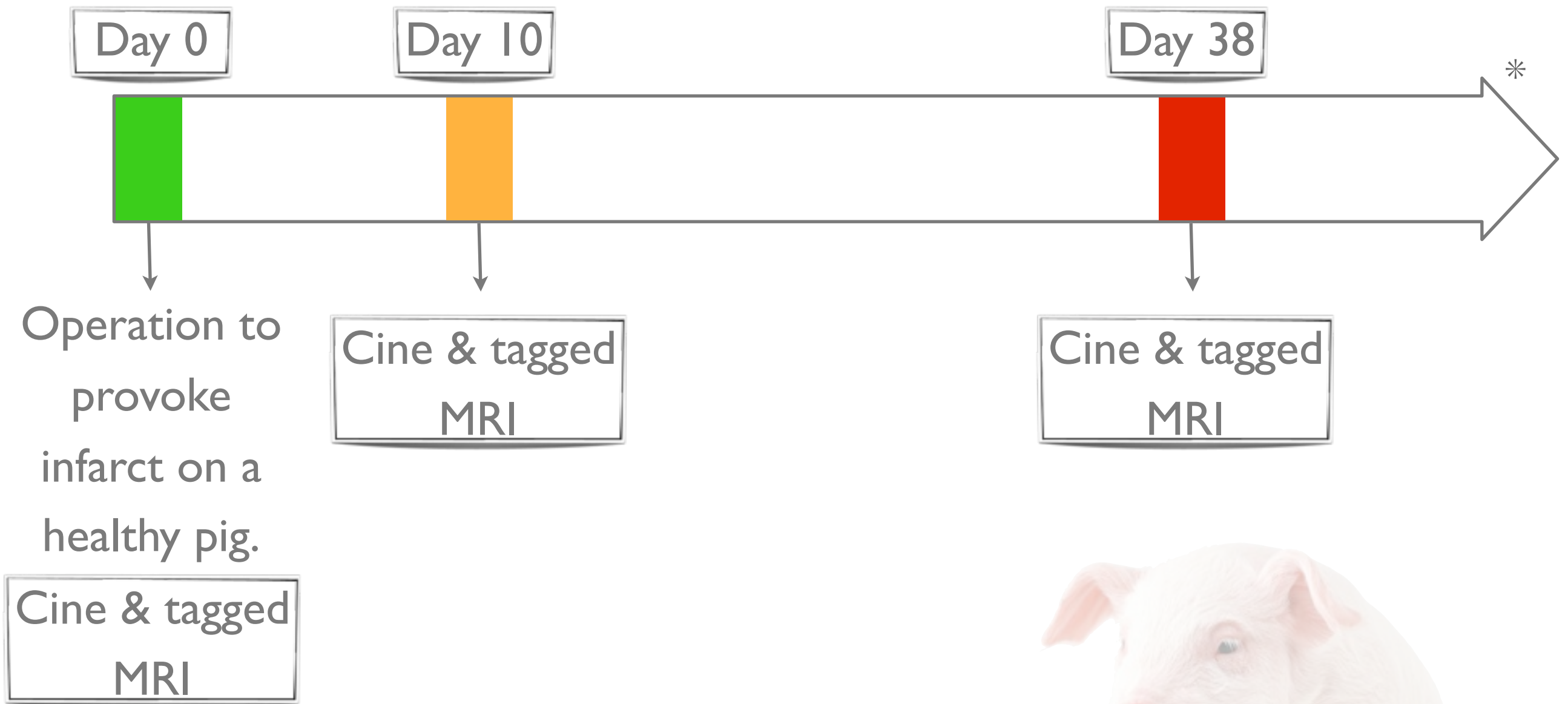
Modeling visible displacements





# A Real case : experience settings

Experience timeline :

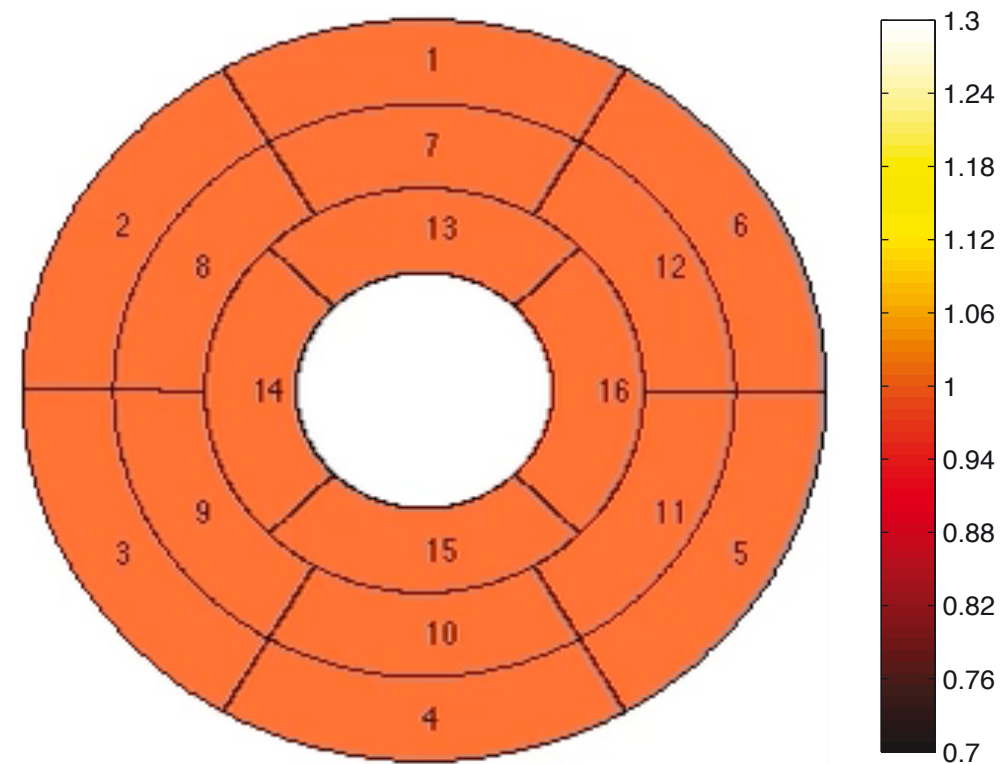
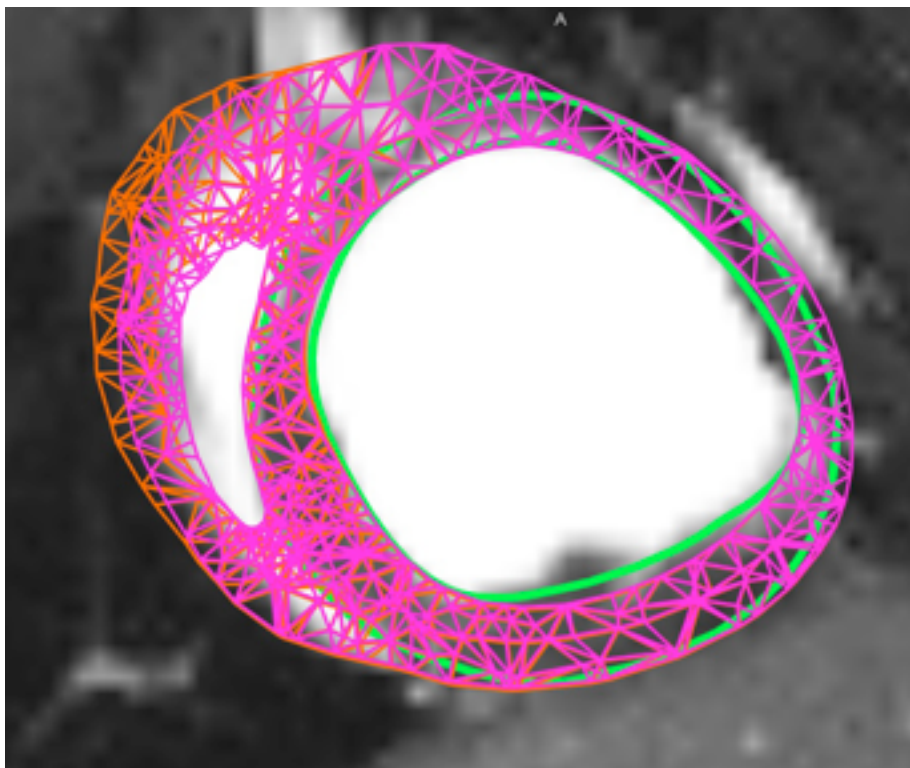
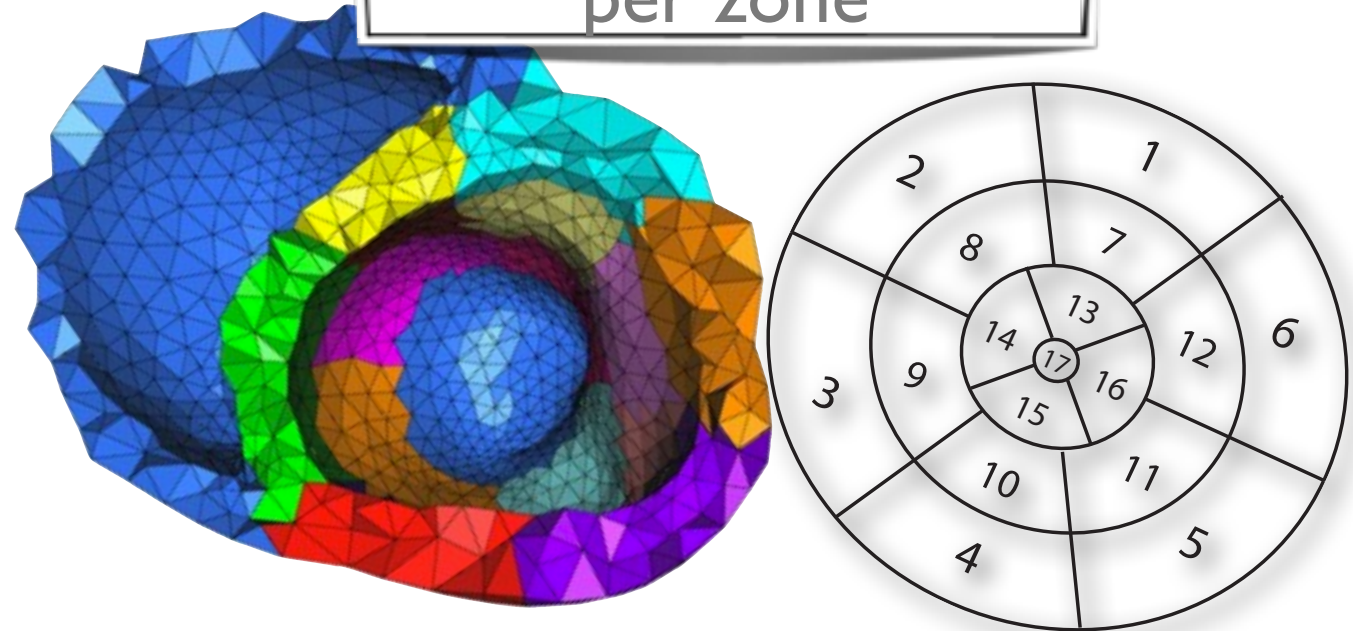


*\*Experience setting by Radomir Chabiniok, King's College London, UK in collaboration with the hospital Henri-Mondor, Créteil.*

# A Real case : numerical results

- Estimation of a contractility parameter in the active law
- The Data are the internal and external surfaces at Day 38.

Parameter is constant  
per zone



# Conclusion

- Provide a methodology for parameter estimation.
- Different types of data are considered.
- Robustness tested on a real case.
- Up coming work :

Assimilating raw images

Convergence analysis of the estimator

Improvement in modeling

