### 4D signal processing for lung cancer treatment by radiotherapy

Nicolas Gallego-Ortiz

### Respiratory motion

- Breath in, breath out
- What moves and how does it moves?





Kyriakou et al. Phys. Med. Biol. **56** (2011) 2999–3013



### Treatment: patient setup critical







4D CT: four dimensional computed tomography

### Agenda

- Why respiratory motion in RT?
- Imaging technologies
- Tumor and lungs motion modeling
	- Finite Element Method
	- Deformable Image Registration

## Why study respiratory motion?

- Correct for image artefacts
- To reduce tumor position **uncertainty margins**



**GTV** Gross Tumor Volume **CTV** Clinical Target Volume **ITV** Internal Target Volume **PTV** Planning Target Volume **OAR** Organ At Risk

E.B. Podgorsak RADIATION ONCOLOGY PHYSICS. Vienna : International Atomic Energy Agency, 2005. Ch 7 p 220

## Variations

### • **Intra-fraction variations**

- Short time changes in a regular respiratory pattern
- Coaching to keep respiration regular

### • **Interfraction variations**

- Day to day changes. Organ fillings, weight loss / gain etc. Changes in respiratory pattern from one day to the other
- Constraints to have real-time tumor tracking – Need to model

## Agenda

- Why respiratory motion in RT?
- **Imaging technologies**
- Tumor and lungs motion modeling
	- Finite Element Method
	- Deformable Image Registration

## Imaging technologies

- Computed tomography
- **4D Computed Tomography**
- 4D Magnetic Resonance
- … multimodality solutions with flouroscopy, Cone-beam CT.
- **Optical surface scanning technologies**

### Concept of 4D CT

• (3D + t) Computed Tomography



Low et al.: Breathing motion measurements **Medical Physics, Vol. 30, No. 6, June 2003**

### A sagittal slice of a lung tumor in motion

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## Optical surface technologies



## Agenda

- Why respiratory motion in RT?
- Imaging technologies
- **Tumor and lungs motion modeling**
	- Finite Element Method
	- Deformable Image Registration

## Lung and tumor motion modeling

• Biophysical models

- Deformable registration (data-driven)
	- Internal deformations related to surface
	- Surface deformation -> motion field

## Agenda

- Why respiratory motion in RT?
- Imaging technologies
- Tumor and lungs motion modeling

### – **Finite Element Method**

– Deformable Image Registration

## What is Finite Element Method?

- Numerical method
- Solve boundary problems
- Partial differential equations
- On a discrete finite element set (discretized solution domain )

### Dynamic model: simplified geometry

Kyriakou et al. Phys. Med. Biol. **56** (2011) 2999–3013



• Finite element model: Rectangle driven by a piston. Sinusoidal displacement profile

### Dynamic model:

Kyriakou et al. Phys. Med. Biol. **56** (2011) 2999–3013

#### lung material: hookean linear and viscoelastic

Table 1. Median values for the key parameters of the model.



ADINA (commercial software for Automatic Dynamic Incremental Nonlinear Analysis).

### Dynamic model: Assessment: 4DCT data

Kyriakou et al. Phys. Med. Biol. **56** (2011) 2999–3013



### Dynamic model: Parameters tunning



## Agenda

- Why respiratory motion in RT?
- Imaging technologies
- Tumor and lungs motion modeling – Finite Element Method
	- **Deformable Image Registration**

### What is image registration?

http://www.itk.org/ItkSoftwareGuide.pdf



### How does it work? Image Registration

http://www.itk.org/ItkSoftwareGuide.pdf



### What is de difference between?

http://www.itk.org/ItkSoftwareGuide.pdf

**Rigid Registration Deformable Registration?**





### Surface Volume Model Deformable registration

Fayad et al. Med. Phys. 39 (6), June 2012



### Principal Component Analysis Approximated relation surface volume

$$
d_{j} = [u_{1,1,j}, u_{1,2,j}, u_{1,3,j}, ..., u_{M,3,j}, s_{1,j}, s_{2,j}, ..., s_{N,j}]^{T}
$$
, (4)  
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D = [\tilde{d}_{1}, ..., \tilde{d}_{j}, ..., \tilde{d}_{J}],
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\bar{d} = \frac{1}{J} \sum_{j=1}^{J} d_{j}.
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D = \lambda D
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 $\ddot{\tilde{d}} = [\tilde{u}, \tilde{s}]^T$ 

 $(8)$ 

 $(9)$ 

Fayad et al. Med. Phys. 39 (6), June 2012

$$
\tilde{u}(t) = E_u E_s^{-1} \tilde{s}(t) = B \tilde{s}(t),\tag{11}
$$

 $E = [e_1, ..., e_k]$ <br>  $W = [..., w_k, ...]$ 

 $d(t) \approx \bar{d} + \sum_{k=1}^{K} w_k(t) e_k.$ 

 $\tilde{d} \approx EW$ ,

 $\widetilde{u} \approx E_u W$  $\tilde{s} \approx E_{\rm s} W$ <sup>'</sup>

### Principal Component Analysis Approximated relation surface volume

d<sub>j</sub> = [u<sub>1,1,j</sub>, u<sub>1,2,j</sub>, u<sub>1,3,j</sub>,..., u<sub>M,3,j</sub>, s<sub>k</sub>, s<sub>2,j</sub>,..., s<sub>N,j</sub>]<sup>T</sup>, (4) u<sub>m,i,j</sub> *ith* displacement of  
\n
$$
D = [\tilde{d}1, ..., \tilde{d}j, ..., \tilde{d}J],
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\n(5) Displacement nth  
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\tilde{d}
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 $\tilde{u}(t) = E_u E_s^{-1} \tilde{s}(t) = B \tilde{s}(t),$  $(11)$ 

#### Reconstructed volumes Med. Phys. Med. Phys. 39 (6), June 2012







Generated CT: SurfaceMapModel



**Generated CT: AmpModel** 



Generated CT: AmpPhModel



**Generated CT: PhModel** 



Pixel

### Surface + registration Measuring respiratory patterns



## Surface + registration

### Measuring respiratory patterns



### Surface + registration Measuring respiratory patterns



### Surface + registration Measuring respiratory patterns



#### Patient surface motion J Schaerer et al. Phys. Med. Biol. **57** (2012) 357–373



## Summary

- Modeling + Simulation
	- Very simplified model approaches real tumor trajectories
- Deformable registration (motion field)
	- Typical breathing cycle motion (4D CT data)
	- Given a surfaces -> reconstruct volumen
	- Sensing + registering surfaces -> extract external motion field

### **Discussion**

- Still no evidence that monitoring skin thoracoabdominal area can improve tumor position estimation
	- Initial condition for tumor trajectories
	- Breaking correlation -> changes in respiratory pattern
	- Same tumor size and position DONT => same trajectory
- Will a model based on planning data hold for delivery fractions? (Time invariance)

# Thank you! What is your opinion?

### Lung cancer

- Mortality rates
- Categories
- Treatment options
- Radiotherapy
	- Margins

### death rates US data 1930-2007

#### **CA: A Cancer Journal for Clinicians**

Volume 62, Issue 1, pages 10-29, 4 JAN 2012



**Year of Death** 

 $\partial$  death rates

US data 1930-2007

#### **CA: A Cancer Journal for Clinicians**

Volume 62, Issue 1, pages 10-29, 4 JAN 2012



**Year of Death** 

### Two categories

- Small cell lung cancer (SCLC)  $15 - 20$  % of lung cancers Agressive
- Non-small cell lung cancer (NSCLC)

### Patient setup

- Registration
- Surface Matching

## Margins



GTV Gross tumor Volume CTV Clinical Target Volume ITV Internal Target Volume PTV Planning Target Volume OAR Organ At Risk

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