Combination of Markerless Surrogates for Motion Estimation in Radiation Therapy CARS 2016

**T. Geimer**, M. Unberath, O. Taubmann, C. Bert, A. Maier June 24, 2016 Pattern Recognition Lab (CS 5) FAU Erlangen-Nürnberg

Universitätsklinikum Erlangen









# **Motivation**

External Beam Radiation Therapy

- Planning CT  $\rightarrow$  Optimized dose distribution
- Respiratory motion  $\rightarrow$  Target volume displacement
- ! Potential survival of malignant cells





© BrainLAB AG



# **Motion Estimation**

### **Clinical State of the Art**

- Marker Tracking
  - × Control points based
  - $\times$  Only sparse deformation





# **Motion Estimation**

### **Clinical State of the Art**

- Marker Tracking
  - × Control points based
  - $\times$  Only sparse deformation

#### **Novel Approaches**

- Range Imaging
  - ✓ Dense surface information
  - ✓ Non-intrusive
- Fluoroscopy
  - ✓ Image-based signal
- Internal motion models
  - Dense internal deformation





# Outline

- Motion Estimation
  - Motion Models
  - Correlation & Prediction
  - Multi Surrogate Approach
- Results & Discussion
- Outlook



# **Motion Models**

#### **Internal Motion**

- Non-rigid 3D/3D registration w.r.t. reference phase t<sub>ref</sub>
- Cropping to internal ROI

#### **Surface Motion**

- Surface extracted from 3-D data set
- Deformation fields **d**<sub>i</sub> interpolated at mesh-vertices





## **Framework Overview**





## **Framework Overview**





## **Framework Overview**





# Fluoroscopy Surrogate Model

#### Preprocessing

- DRRs from volumetric CT data
- According to Vero kV beam geometry

#### **Breathing Signal Extraction**

- · Linearized images
- · PCA of the change in intensity
- First few principle component correlate highly with breathing phase





# Multi-surrogate Approach

#### Motivation

- · Combined information from range imaging and fluoroscopy
- · Improve the estimation or compensate for one surrogate failing

#### Combination

- Concatenated feature vector  $\boldsymbol{\sigma}_{{}_{\mathsf{CB}_{i}}} = \begin{bmatrix} \boldsymbol{\sigma}_{{}_{\mathsf{R}_{i}}} \\ \boldsymbol{\sigma}_{{}_{\mathsf{FL}_{i}}} \end{bmatrix} \in \mathbb{R}^{(f_{\mathsf{R}_{i}} + f_{\mathsf{FL}}) \times 1}$
- Regression matrix  $\mathbf{R}_{CB} \in \mathbb{R}^{I \times (f_{RI} + f_{FL})}$

Example:  $\mathbf{R}_{CB} \in \mathbb{R}^{3 \times (2+2)}$ 





## **Evaluation**

### Data

- Nine 4-D CT patient data sets
- Ten volumes each (voxel size:  $0.97 \times 0.97 \times 2.5 \text{ mm}^3$ )
- DRRs: detector with  $1024 \times 768$  pixels and 0.39 mm pixel spacing

### Experiments

- Correlation study on feature weights
- Leave-one-out-assessment of the estimation error w.r.t. ground-truth



## **Results: Correlation**





## **Results: Estimation**





# Discussion

### Single Surrogate

- Fluoro outperforms surface surrogate: 0.67  $\pm$  0.33 mm for I = 2
- No improvement with higher internal model dimension

## **Combination Approach**

- No consistent improvement
- $f_{\text{RI}}, f_{\text{FL}} = 1$ : 1-D respiratory phase  $\rightarrow$  Linearly dependent
- Best overall for l = 2 from (2+2) or higher combined features:

#### $0.62\pm0.28~\text{mm}$

#### ightarrow Surrogate combination useful under certain circumstances



# Outlook

### Surrogate Extraction

- DR eliminates non-redundant information
- · Sophisticated ways to extract mutually exclusive information

### Data Acquisition & Training

- Training currently only done on 6 of 9 phases (necessity of leave-one-out approach)
  - $\times$  ASMs do not describe an entire cycle
  - ✓ Training on entire cycle
  - ✓ Prediction on another of the same patient



# Thank you for your attention.

### **Questions?**

