

# Preliminary study investigating brain shift compensation using 3D CBCT cerebral vascular images

Siming Bayer<sup>1</sup>, Roman Schaffert<sup>1</sup>, Nishant Ravikumar<sup>1</sup>, Andreas Maier<sup>1</sup>, Xiaoguang Tong<sup>3</sup>, Hu Wang<sup>3</sup>, Martin Ostermeier<sup>2</sup>, Rebecca Fahrig<sup>2</sup>

<sup>1</sup> Pattern Recognition Lab, Department of Computer Science, Friedrich-Alexander-Universität Erlangen-Nürnberg, Erlangen, Germany

<sup>2</sup> Siemens Healthcare GmbH, Forchheim, Germany

<sup>3</sup>Tianjin Huanhu Hospital, Tianjin, China

### Introduction

- **Brain shift** phenomenon in a neurosurgical procedure:
  - Time dependent elastic deformation (see Fig. 1).

- Image guided navigation systems:
  - Assume rigid behavior of the head and its content,
  - Are not able to recover elastic deformation.
  - => Brain shift affects the accuracy of the neurosurgery.

### Materials and Methods

- The overall brain shift compensation pipeline includes the following steps:
  - Feature extraction,
  - Point matching (coherent point drift vs. robust point matching),
  - Displacement field interpolation (thin plate spline vs. B-spline).
- Feature extraction:
  - Use Frangi's vesselness filter [1] to enhance the vasculature.
  - Extract centerline by using an octree data structure [2].
  - Detect bifurcation points with a 3x3x3 window around each point of the centerline.
- Point matching:
  - Coherent point drift (CPD) [3]:
    - Target point set represents the data points,
    - Source point set represents the GMM centroids,



**Figure 1:** Cone beam CT images acquired during a tumor resection surgery. (a) Preoperative image. (b) After dura opening, a small deformation is visible. (c) After resection, the surrounding tissue undergoes a larger deformation.

	Before	CPD- B-Spline	CPD- TPS	RPM- B-Spline	RPM- TPS
Full brain	0.92	0.94	0.94	0.95	0.93
ROI	0.89	0.93	0.92	0.94	0.90
Full brain (+ Outlier)	0.92	0.85	0.66	0.85	0.65

- Maximize the posterior probability.
- Robust point matching (RPM) [4]:
  - Soft correspondence for fuzzy assignment,
  - Least square optimization of an energy function.
- Displacement field interpolation
  - Thin plate spline (TPS) [5]: globally controlled, belongs to the family of Radial Base Functions.
  - B-spline [6]: locally controlled, only use the information in the neighborhood.

# **Results and Discussion**

- Experiment:
  - With synthetic digital phantom data [7].
  - Use relative overlap metric for the evaluation.
  - Calculated both for the entire brain parenchyma and ROI.
  - Additionally, add 20% outliers to evaluate the robustness.
- Results (see Tab. 1 and Fig. 2)
  - Outliers affect the accuracy greatly.
  - B-spline interpolation outperforms TPS.

ROI		0 00	$\cap \circ \gamma$	0 0 2	$\cap \circ \gamma$
(+ Outlier)	0.89	0.00	0.02	0.92	0.82

**Table. 1**.: Relative overlap rate of the entire brain parenchyma and region of interest (ROI) after using different point matching and interpolation techniques.



**Figure 2**.: An example axial slice with displacement field overlay. (a) and (b) are the source and target images, respectively. (c) – (f) are registration result of different methods.

 Result of CPD depends on the choice of parameter controlling outlier proportion.

#### Conclusions

- A feature-based registration pipeline for brain shift compensation with 3D CBCT cerebral vasculature is proposed.
- We compared different point matching methods and interpolation techniques on digital phantom data.

Contact

siming.bayer@fau.de
 https://www5.cs.fau.de/en/our-team/bayer-siming



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Pattern Recognition Lab, Department of Computer Science, Friedrich-Alexander-Universität Erlangen-Nürnberg

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