

# Generation of Personalized Computational Phantoms Using Only Patient Metadata

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## Introduction

- Computational phantoms have found widespread use for
  - Optimization of acquisition settings and workflow for diagnostic CT [1]
  - Skin dose monitoring and positioning of interventional X-ray system [2]
- Existing approaches include template deformation using
  - Scaling factors from anatomical measurement [3]
  - Scaling factors from body mass index [4]
- We generate personalized computational phantoms including internal organs using **a learning based approach**

## Materials and Methods

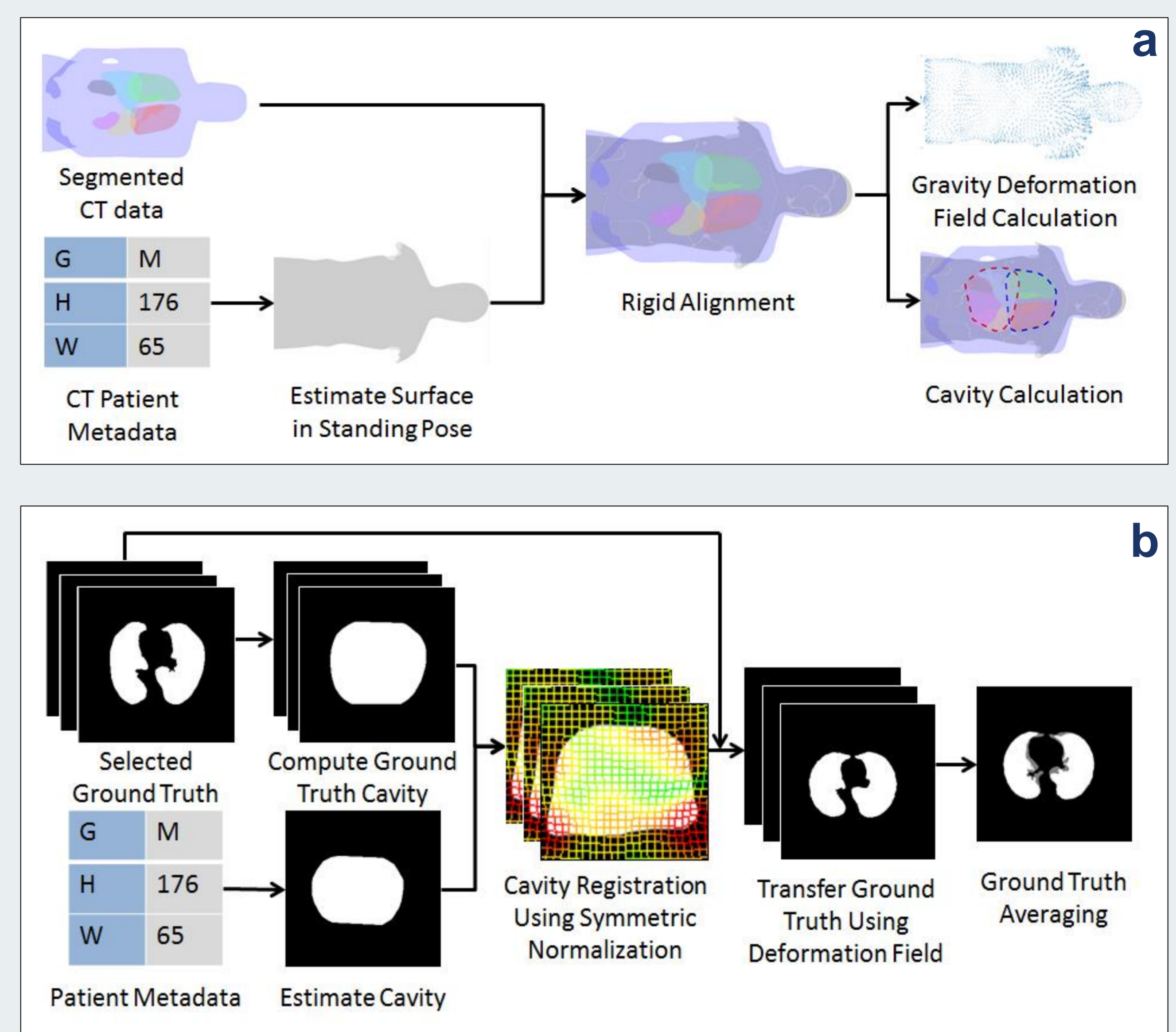
- **Training data**
  - Patient surface models: avatar database (standing)
  - Segmented 3D diagnostic scans: Visceral Anatomy3 (supine) [5]
- **Internal body cavities**
  - Minimal convex hull around selected organs
  - Modeled individually, e.g. for thorax and abdomen
- **Patient-dependent boundary determination** (Fig. 1a)
  - Learn a joint subspace linking patient metadata with patient surface model in standing pose based on the avatar database
  - Estimate for each CT scan a surface model using associated metadata and align the estimate with segmented surface
  - Calculate gravity deformation field to adjust the estimates from standing to supine pose
  - Learn a joint subspace linking patient metadata with surface models in standing pose, gravity deformation fields, and cavities model.
- **Internal organ model filling** (Fig. 1b)
  - Estimate internal cavity based on associated patient metadata
  - Select N closest datasets from database matching metadata
  - Non-rigid registration between estimated and selected cavities
  - Propagate the segmentation of selected ground truth to the estimates
  - Merge and extract segmentation using popular vote

## Results and Discussion

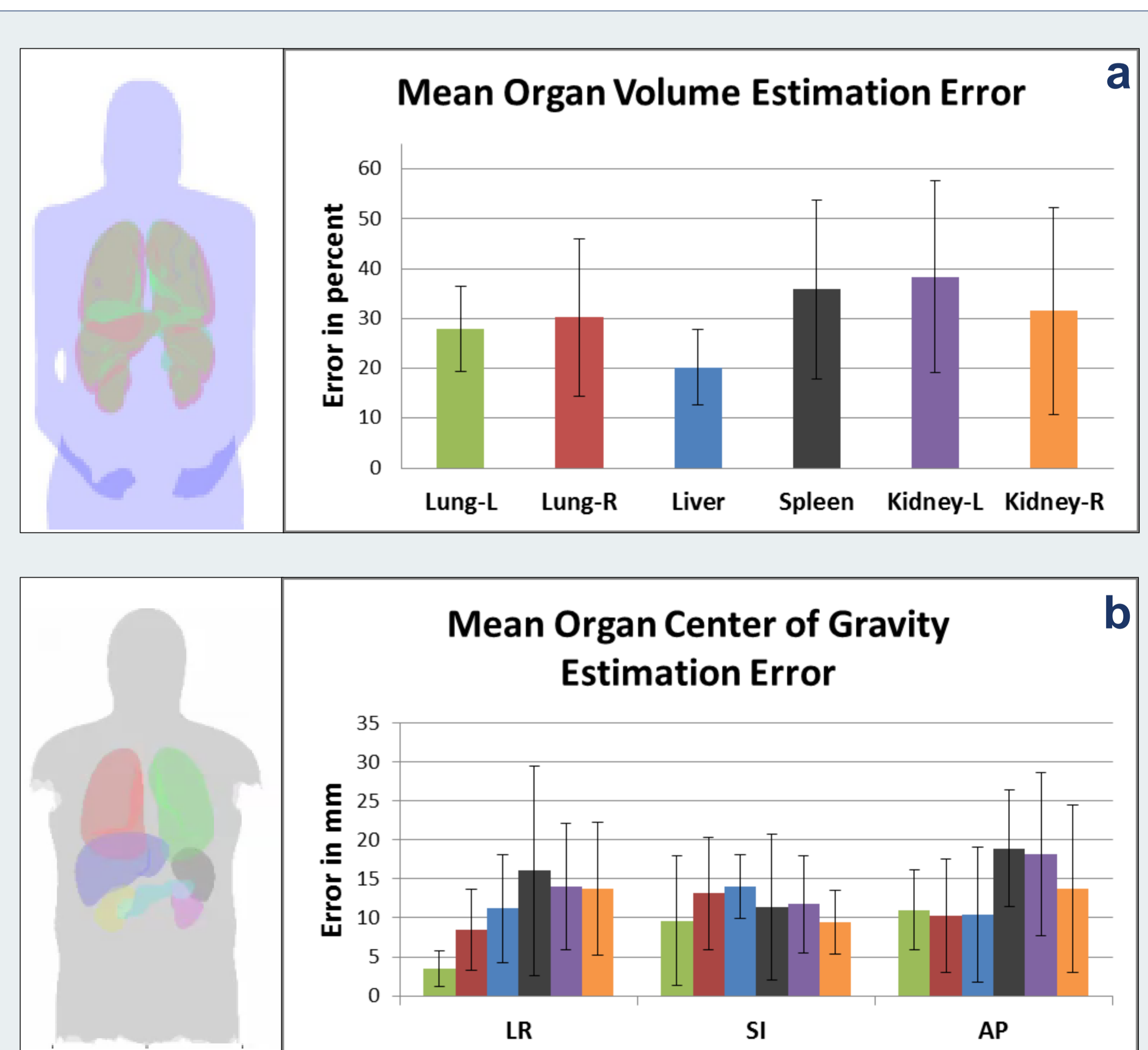
- **Evaluation data**
  - Validation dataset: Visible Human [6] and Golem [7].
  - Test dataset: 9 full-body CTs
- Organ volume estimation error:  $30.6 \pm 15.0$  % (Fig. 2a)
- Mean center of gravity error over all organs:  $23.5 \pm 8.3$  mm (Fig. 2b)
- High inter-patient variation for organ mass and dimensions
- Planned: convert computational model to attenuation maps and evaluate for attenuation correction

## Conclusions

- We proposed a learning based method to estimate personalized computational phantoms (including organs) using patient metadata only.
- Good accuracy for organ localization



**Figure 1:** Workflow for generation of personalized patient model. (a) Estimation of gravity deformation field and cavities. (b) Internal organ estimation via cavity determination.



**Figure 2:** Evaluation results of proposed algorithm. (a) Internal organs of the Visible Human (green) and estimated corresponding model (red) overlaid in AP view (left). Organ volume estimation errors (right). (b) An example of personalized computational phantom estimate (left). Mean organ center of gravity displacement error from evaluation dataset (right).

## References

- [1] Singh et al., MICCAI (2014)
- [2] Schaller et al., MICCAI (2009)
- [3] Broggio et al., PMB (2011)
- [4] Johnson et al., Proc. IEEE (2009)
- [5] del Toro et al., ISBI (2014)
- [6] Spitzer et al., JAMIA (1996)
- [7] Petoussi-Henss et al., PMB (2002)

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