

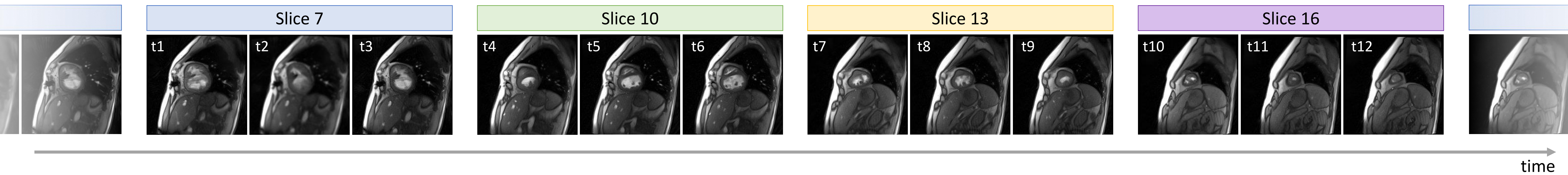
AI-based Evaluation of cardiac real-time MRI with congenital heart disease

Philipp Rosauer¹, Wadim Koslow¹, Anja Bach², Alex Hoff², Jens Tank², Darius A. Gerlach²

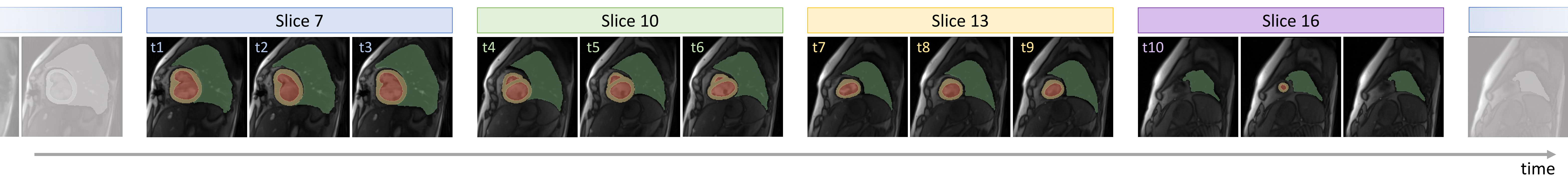
¹ Institute for Software Technology, German Aerospace Center (DLR), Cologne, Germany

² Institute of Aerospace Medicine, German Aerospace Center (DLR), Cologne, Germany

Real-time Magnetic Resonance Images of a univentricular heart recorded at 30fps



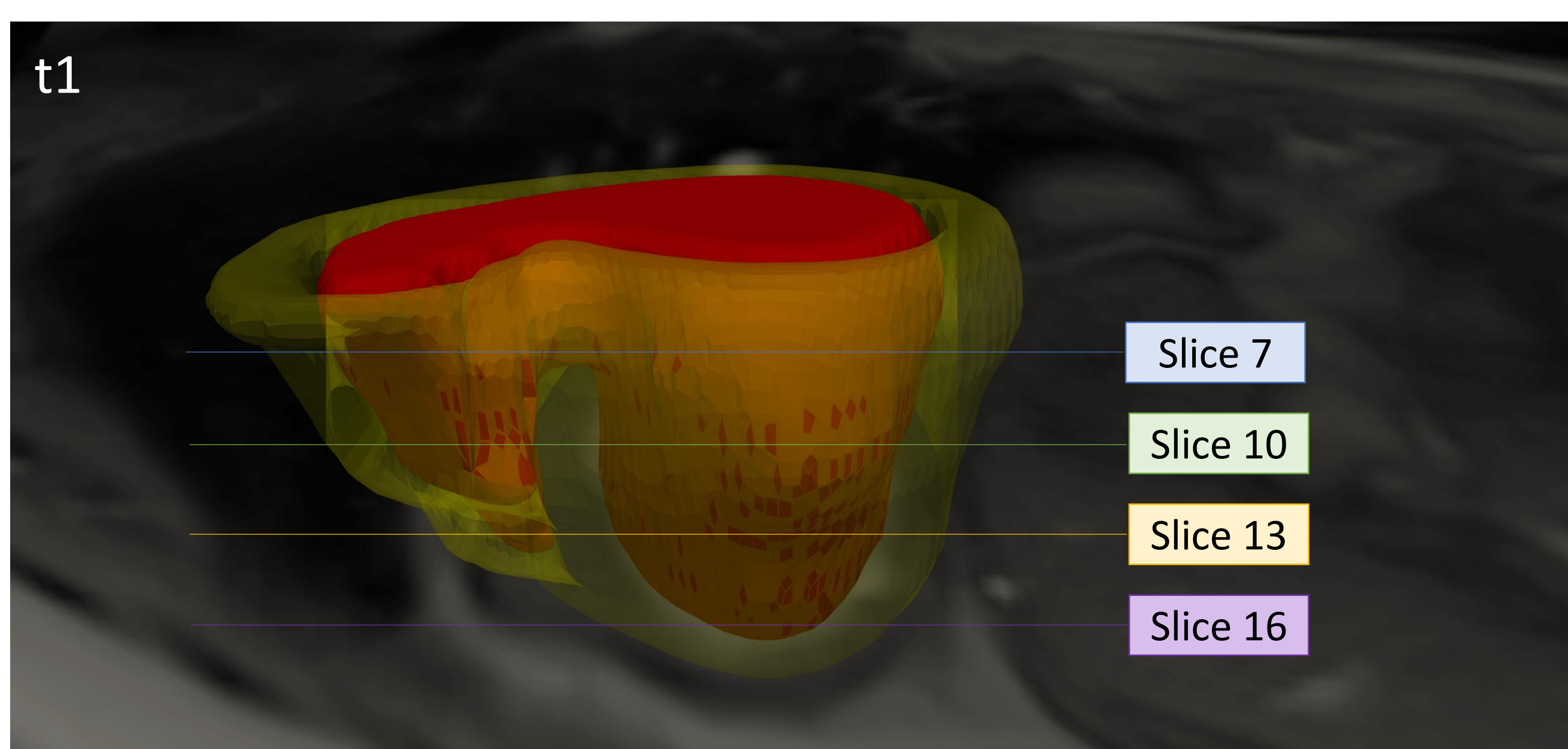
Automatic Segmentation of the Left Ventricle, Myocard and Lung



Automatic Synchronization of Respiration and Cardiac Cycle



Visualization of a beating heart with segmented Left Ventricle and Myocard



Methods

- **Semantic Segmentation** of Left Ventricle and Myocard using U-Net architecture
- Pretrained nnU-net [1] with healthy hearts in short axis view
- Expert Annotations of CINE and Real-time MRI of univentricular hearts for transfer learning

Results

- Automatic breath- and cardiac cycle synchronization
- **Determination of diastolic and systolic phases** based on segmentations
- **Determination of respiratory phases** based on lung, liver and thorax movements
- Evaluation of vital parameters
- Computation of stroke volume at various pairs of respiratory and cardiac phases
- **Influence of respiration on stroke volume and myocardial function**

References

[1] Isensee, F., Jaeger, P.F., Kohl, S.A.A. *et al.* nnU-Net: a self-configuring method for deep learning-based biomedical image segmentation. *Nat Methods* 18, 203–211 (2021).

Background

- Cardiac MRI is an important diagnostic tool in heart diseases
- Assessment of **vital parameters like stroke volume or myocardial function**
- Conventional CINE MRI limited in congenital heart diseases due to limited breath hold capabilities and irregular heart beat
- **Real-time MRI** with framerates over 30fps allows **recording during spontaneous breathing**
- High framerates yield large amounts of data that need to be evaluated highly efficiently and with as little manual intervention as possible

Requirements

- Comprehensible predictions, easy to verify by medical staff
- **Editable interim results**
- **High performance** on standard hardware
- High expendability for further research

Key challenges

- Deep Learning with **small and noisy data sets**
- Very specific heart anatomies
- Anomalies in data due to metal artifacts
- Limited amount of patients, only few examined with real-time MRI
- **Acquisition of 2D** videos, layer by layer ↔ **Evaluation of 4D** (3D+time) data
- Patients are able to breath freely, respiration dependency
- **Reliability** and Uncertainty Quantification

Evaluation of vital parameters like the stroke volume subject to respiration

