

# Introduction to Coherent Depth Fields for Dense Monocular Surface Recovery

## Supplementary Material

Vladislav Golyanik<sup>1,2</sup>  
vladislav.golyanik@dfki.de

Torben Fetzer<sup>1</sup>  
torben.fetzer@dfki.de

Didier Stricker<sup>1,2</sup>  
didier.stricker@dfki.de

<sup>1</sup> Department of Computer Science  
University of Kaiserslautern

<sup>2</sup> Department Augmented Vision  
German Research Center for Artificial  
Intelligence (DFKI)

The present exposition is intended to complement the paper "*Introduction to Coherent Depth Fields for Dense Monocular Surface Recovery*". An interested reader might find here further useful insights. Appendix A details the choice of  $G$ . Appendix B provides more details about several additional sequences used in the experiments.

## A On the choice of $G$

There are several ways to choose  $G$ . To facilitate well-posedness, second and higher order derivatives of  $G$  must exist. At the same time, dependency from the measurements at large distances must vanish [9]. Gaussian, Lorentzian and Pearson type VII kernels satisfy the conditions mentioned above, among others. Note that CT regulariser expresses a process approaching a smoothing by a respective kernel. Thus, high-frequency structures can still be reconstructed (the sharpness is guided by a regularisation parameter).

## B Experiments on real-world image sequences

Apart from the *heart*, *face*, *back* and *laparoscopy* sequences, we tested CDF on four other image sequences. Three of them are real-world image sequences — the new *face* [9], *music notes* [9] and *owl* [9] — and one is the synthetic SINTel's *bandage2* [9].

The *face* sequence contains 120 frames and shows a face of a speaking person on a static background with sufficient rotation and deformations. The *music notes* sequence contains 139 frames and shows a bent and twisted music sheet. Note that this sequence contains large deformations. CDF tends to explain observed deformations as much as possible through rotation of rather a mean shape with smaller deformations. This effect is a common limitation of NRSfM methods which do not explicitly model large deformations. The *owl* sequence is a video depicting a quiet barn owl watching the environment. The head movements are slow and gracious and are accompanied by frequent changes of sight direction. In total, there are 202 frames. The *shaman2* sequence is a fragment of an open-source animated movie [9]. In the foreground, it depicts an amply textured human face which we have reconstructed.

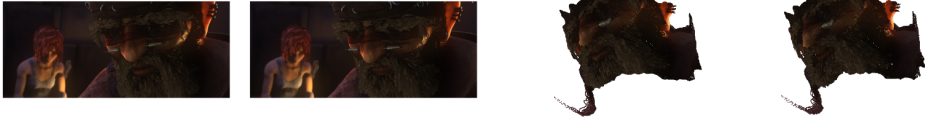
face



music notes



shaman2



owl



Figure 1: Selected frames and reconstructions by CDF from the new *face* [4], *music notes* [4], *owl* [4] and SINTEL’s *bandage2* [4] sequences.

Fig. 1 contains selected frames from all four sequences (on the left side) and exemplary reconstructions (on the right side). Moreover, this supplementary material includes a video visualising the reconstructions.

## References

- [1] D. J. Butler, J. Wulff, G. B. Stanley, and M. J. Black. A naturalistic open source movie for optical flow evaluation. In *European Conference on Computer Vision (ECCV)*, pages 611–625, 2012.
- [2] P. Dinning. *Barn Owl at Screech Owl Sanctuary*. <https://www.youtube.com/watch?v=xmou8t-DHh0>, 2014. [online; accessed on 25.04.2017].
- [3] V. Golyanik, A. S. Mathur, and D. Stricker. NRSfM-Flow: Recovering non-rigid scene flow from monocular image sequences. In *British Machine Vision Conference (BMVC)*, 2016.
- [4] V. Golyanik, T. Fetzer, and D. Stricker. Accurate 3d reconstruction of dynamic scenes from monocular image sequences with severe occlusions. In *Winter Conference on Applications of Computer Vision (WACV)*, 2017.
- [5] A.L. Yuille and N.M. Grzywacz. The motion coherence theory. In *International Conference on Computer Visions (ICCV)*, 1988.