

Learning 3D Shape Completion from Laser Scan Data with Weak Supervision

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Problem

Learning **3D shape completion** of sparse and noisy point clouds **without ground truth**.



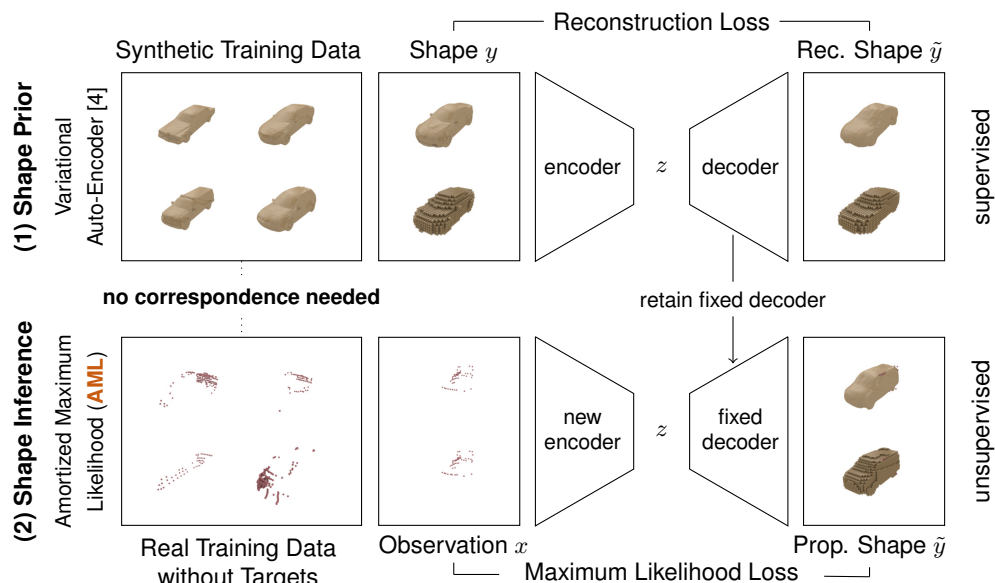
Related Work

- Data-driven [1] – slow inference;
- Learning-based [2] – supervision required.

Contributions

- Learning-based, weakly-supervised, amortized maximum likelihood (**AML**) approach.
- Synthetic and real benchmarks.
- Improved results published in [3].

Method



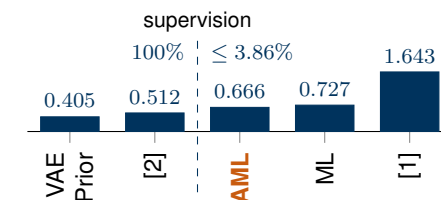
References

- [1] F. Engelmann et al. "Joint Object Pose Estimation and Shape Reconstruction in Urban Street Scenes Using 3D Shape Priors". In: *GCPR*. 2016.
- [2] A. Dai et al. "Shape Completion using 3D-Encoder-Predictor CNNs and Shape Synthesis". In: *CVPR*. 2017.
- [3] D. Stutz and A. Geiger. "Learning 3D Shape Completion under Weak Supervision". In: *arXiv.org abs/1805.07290* (2018).
- [4] D. P. Kingma and M. Welling. "Auto-Encoding Variational Bayes". In: *ICLR* (2014).
- [5] A. Geiger et al. "Are we ready for Autonomous Driving? The KITTI Vision Benchmark Suite". In: *CVPR*. 2012.
- [6] A. X. Chang et al. "ShapeNet: An Information-Rich 3D Model Repository". In: *arXiv.org 1512.03012* (2015).
- [7] Z. Wu et al. "3D ShapeNets: A deep representation for volumetric shapes". In: *CVPR*. 2015.
- [8] B. Yang et al. "3D Object Dense Reconstruction from a Single Depth View". In: *arXiv.org abs/1802.00411* (2018).

Results

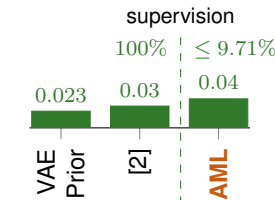
ShapeNet

Accuracy and Completeness [vx] ↓



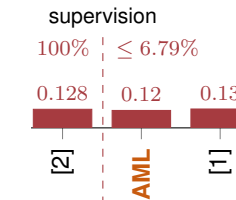
ModelNet10

Occupancy Error ↓



KITTI

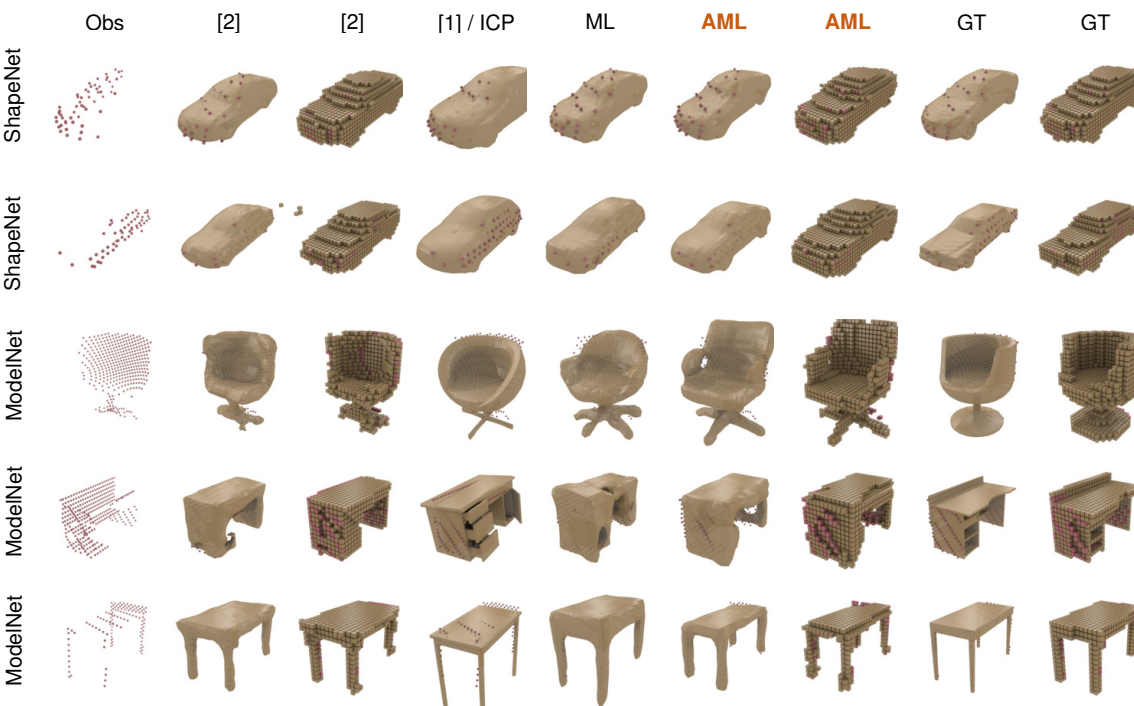
Completeness [m] ↓



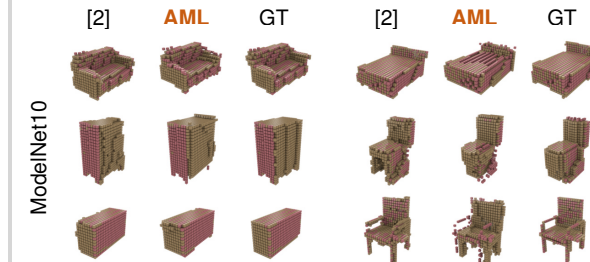
Runtime

~ 2ms (**AML** and [2], GPU) vs. > 168ms ([1], CPU with Multi-Threading) and > 30s (ML, GPU).

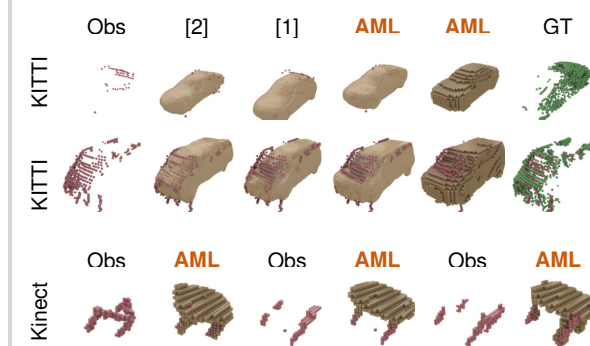
ShapeNet and ModelNet (Synthetic, Single Categories)



ModelNet10 (Synthetic, Multiple Categories)



KITTI and Kinect (Real)



Conclusion

We proposed a weakly-supervised, learning-based, amortized maximum likelihood (**AML**) approach to 3D shape completion.

- Outperforms data-driven approach [1]; but 84 times faster.
- Competitive to learning-based approach [2]; but up to 96% less supervision.

Paper, Code and Data:
davidstutz.de/cvpr2018

