

Point Completion By Unsupervised Skeleton Learning

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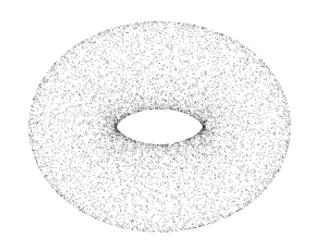


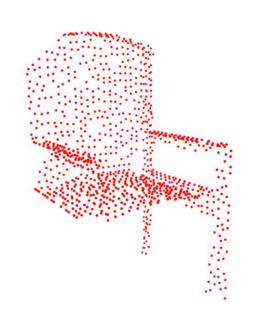


Introduction

Point clouds are used for many purposes, including creating 3D CAD models for manufactured parts and for a multitude of visualization and rendering.

Due to the limitation of view angles or occlusions of devices, raw point cloud is usually incomplete. A point cloud completion method is needed.





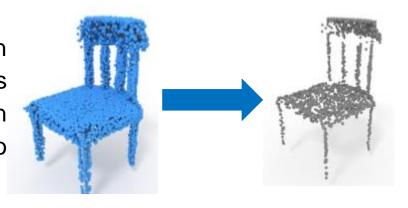


Research Basis

The most common approach is to use an **autoencoder framework**, but the unordered nature of point cloud make it difficult to keep the shape topology.

SK-PCN, a novel point cloud completion model proposed by Nie et.al[1], outperforms the autoencoder-based method by using an intermediate modality meso-skeleton to capture the features of original point cloud.

Limitations: Need both the full point cloud and full skeleton as the ground truth.



Full point cloud

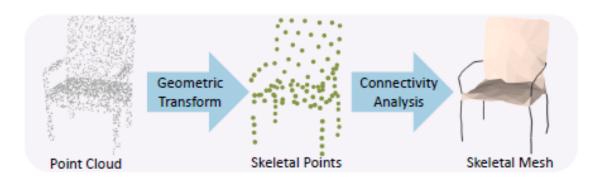
Skeleton points



Research Basis

Lin[2] proposed an unsupervised method **Point2Skeleton** to generate skeletal mesh from point cloud for complex structures.

We hope to use Point2Skeleton to generate skeleton points conveniently and use the idea of SK-PCN and other methods to obtain a complete point cloud through sparse skeleton points.



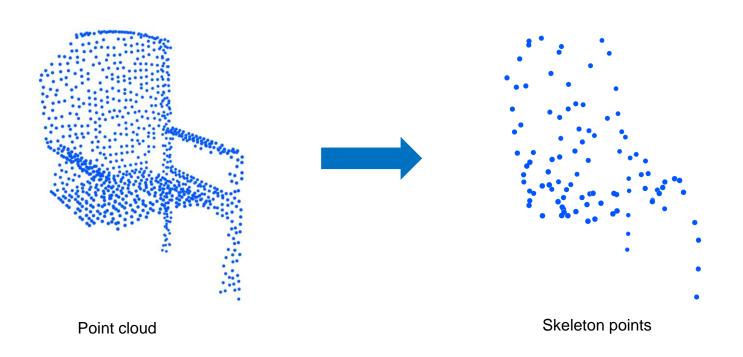
Steps of Point2skeleton



Proposed Network Architecture

Skeleton Generation

Based on **Point2Skeleton**, we can generate skeleton points from its original point cloud. This transformation architecture consists of one PointNet++ encoder and MLPs.

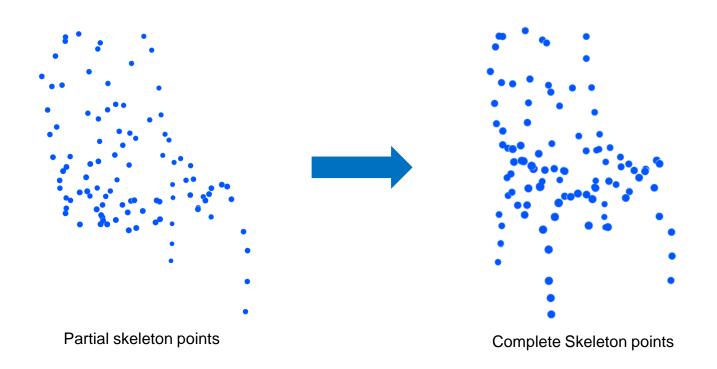




Proposed Network Architecture

Skeleton Completion

Build a sub-network which takes the partial skeleton as input and predicts the complete skeleton given the ground truth. **PU-Net[3]** will be the basic framwork.

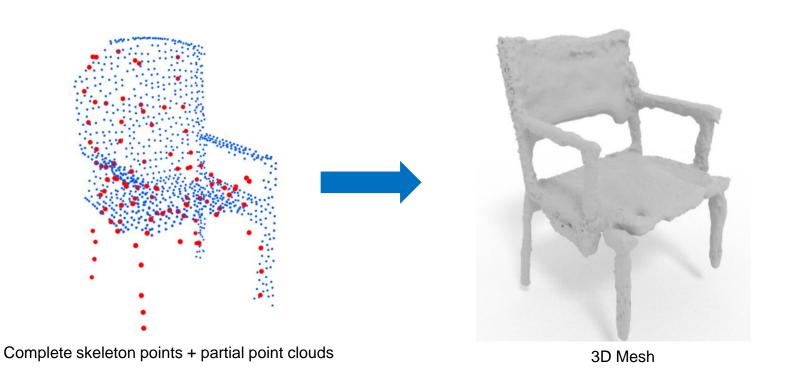




Proposed Network Architecture

Skeleton2surface

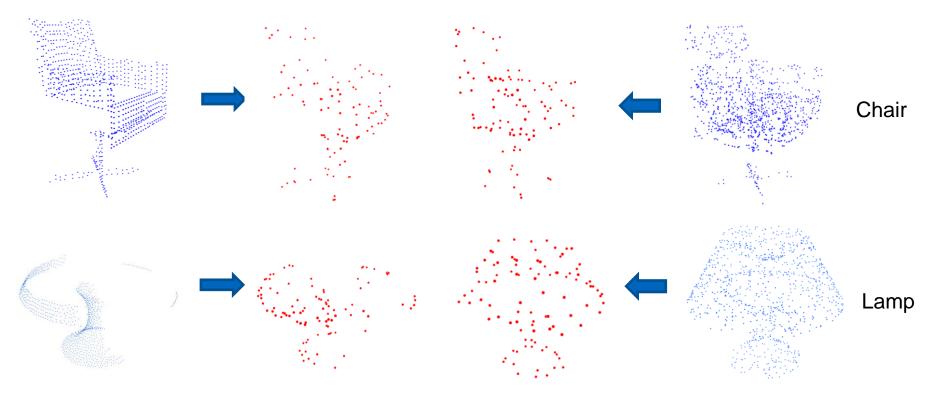
By combining the full skeleton and partial point cloud, we will refer to **P2P-NET[4]** and the completion part of **SK-PCN** to generate our final result, a 3D Mesh.





Our Progress

We have studied the source code of **Point2Skeleton** and achieved the first part of our network: skeleton generation from point cloud based on Point2Skeleton for both partial skeleton and complete ground truth skeleton.





Challenge

- The skeletons generated from partial point cloud are not ideal, somtimes they can't capture the features from point cloud.
- Possible Solutions: increase the number of skeleton points
- Need more training data
- It is not clear whether PU-net as an upsampling method will be helpful for our task



What we do next...





References

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- 2. Cheng Lin, Changjian Li, Yuan Liu, Nenglun Chen, Yi-King Choi, and Wenping Wang. Point2skeleton: Learning skeletal representations from point clouds. arXiv preprint arXiv:2012.00230, 2020.
- 3. Lequan Yu, Xianzhi Li, Chi-Wing Fu, Daniel Cohen-Or, and Pheng-Ann Heng. Pu-net: Point cloud upsampling network. In *Proceedings of IEEE Conference on Computer Vision and Pattern Recognition (CVPR), 2018.*
- 4. Kangxue Yin, Hui Huang, Daniel Cohen-Or, and Hao Zhang. P2p-net: Bidirectional point displacement net for shape transform. *ACM Transactions on Graphics(Special Issue of SIGGRAPH)*, 37(4):152:1–152:13, 2018.



Thank You!

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