Neural Hierarchical Decomposition for Single Image Plant Modeling.

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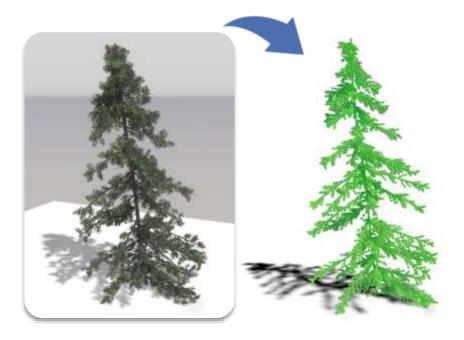


- <u>Vegetation</u> is an important part of natural scenes.
- However, plants usually feature extremely <u>complex topological structures</u>, so obtaining their high-quality, realistic 3D assets remains a very challenging task.







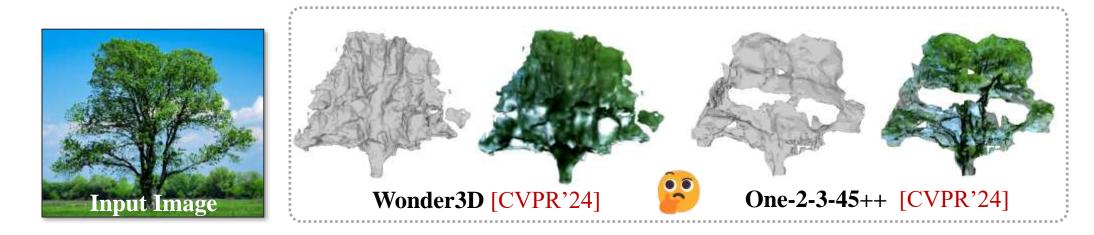


> Single-image-to-3D approaches in Al.

• In Generative AI fields, a series of methods [1,2,3] have been recently proposed to produce 3D geometries from just a single image based on novel view synthesis.



• However, these approaches mainly works well on smooth-surfaced objects with water-tight geometries. When it comes to <u>plants</u>, their outputs are usually problematic, noisy meshes (as follows).



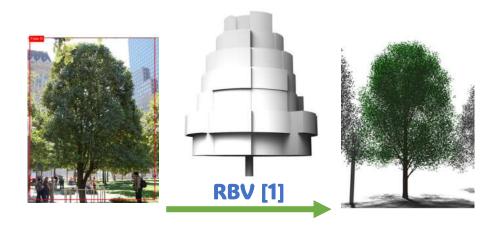
^{[1] &}quot;One-2-3-45: Any single image to 3d mesh in 45 seconds without per-shape optimization.", NeurIPS (2023).

^{[2] &}quot;Wonder3D: Single image to 3d using cross-domain diffusion.", CVPR (2024).

^{[3] &}quot;One-2-3-45++: Fast single image to 3d objects with consistent multi-view generation and 3d diffusion.", CVPR (2024).

> Single-image plant modeling in CG.

• In order to obtain practically-usable 3D plant models, CG researchers also explored combining deep learning and procedural modeling to tackle this problem [1,2].



Limitations of existing works:

- (1) Limited flexibility to depict arbitrary tree shapes.
- (2) Cannot adapt to houseplants.







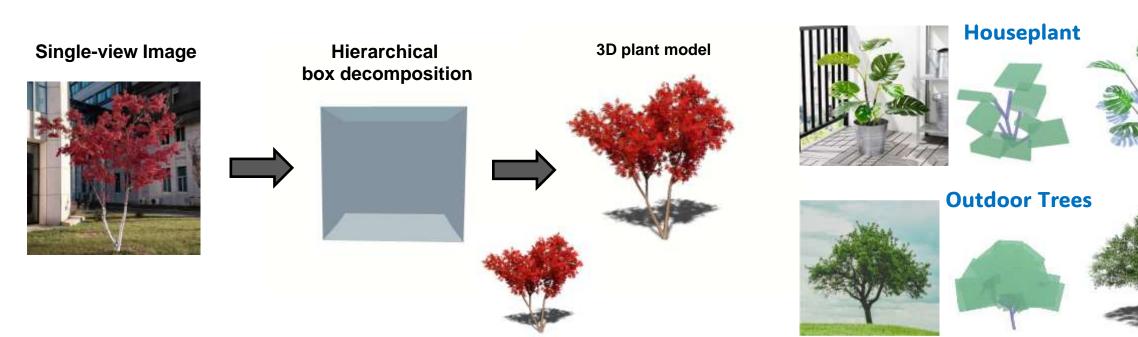


^{[1] &}quot;Learning to reconstruct botanical trees from single images.", ACM Trans. Graph. (2021)

^{[2] &}quot;SVDTree: Semantic Voxel Diffusion for Single Image Tree Reconstruction.", CVPR (2024).

Our Solution:

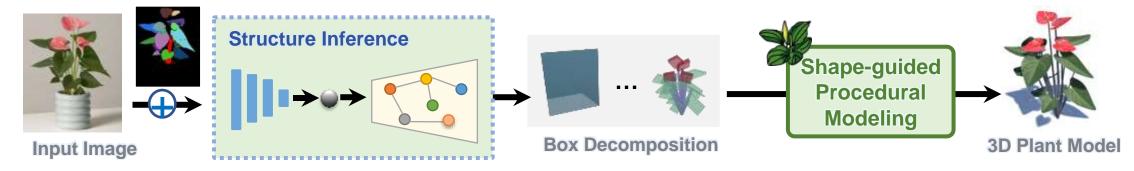
- Here, we introduce a new method for generating high-quality 3D plant models from single images through a systematic combination of *hierarchical box decomposition* and *shape-driven procedural modeling*.
- In the field of plant modeling, our method is the first approach that can handle two distinct plant categories: houseplant and outdoor trees.



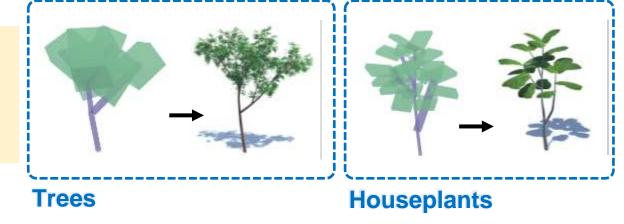
Methodology

Brief Methodology

- □ [Step-1] Given an image with its segmentation mask, we first employ an inference module to progressively yield a hierarchical box decomposition from coarse to fine.
- □ [Step-2] Then, constrained by the last level of boxes, we leverage a <u>shape-guided parametric modeling</u> algorithm to biologically synthesize the corresponding geometric details for each local part.

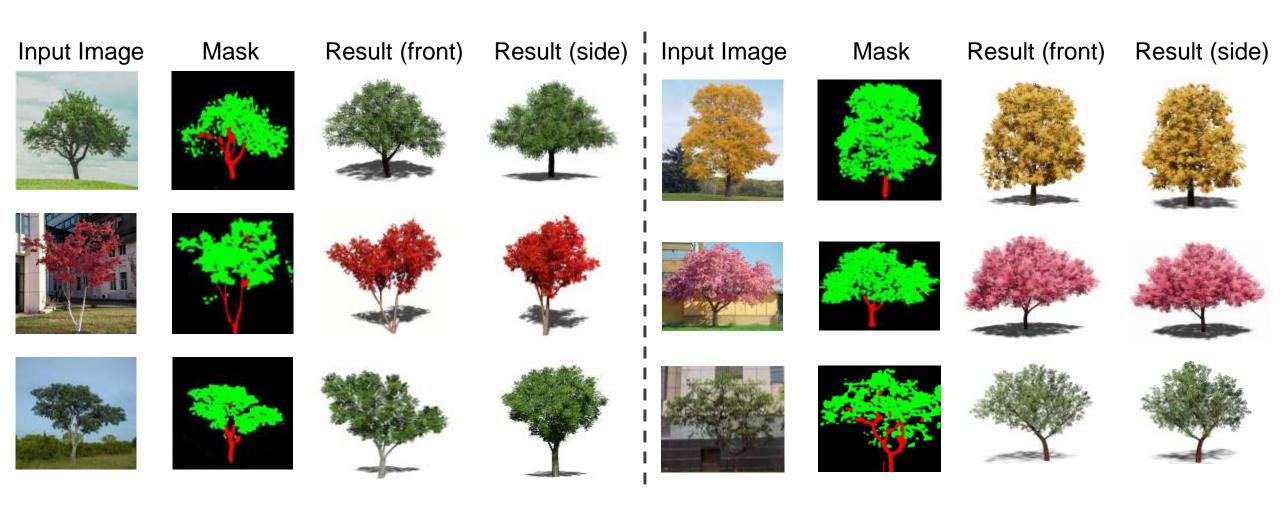


Through learning the decomposition in <u>different levels of</u> <u>details (LOD)</u>, our method can easily adapt to two distinct plant categories: <u>trees</u> and <u>small houseplants</u>, each with unique appearance features.



Results

Reconstruction of Outdoor Trees.



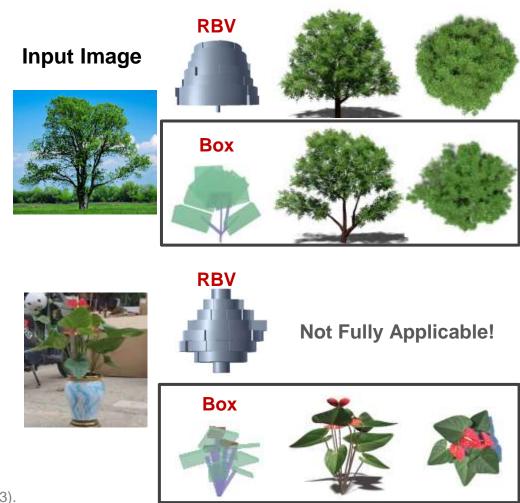
Reconstruction of Houseplants.

Input Image Result (front) Result (side) Input Image Result (front) Result (side) Mask Mask

Comparison with Pure Al-based Methods

One-2-3-45[1] Wonder3D[2] Input Image **Ours**

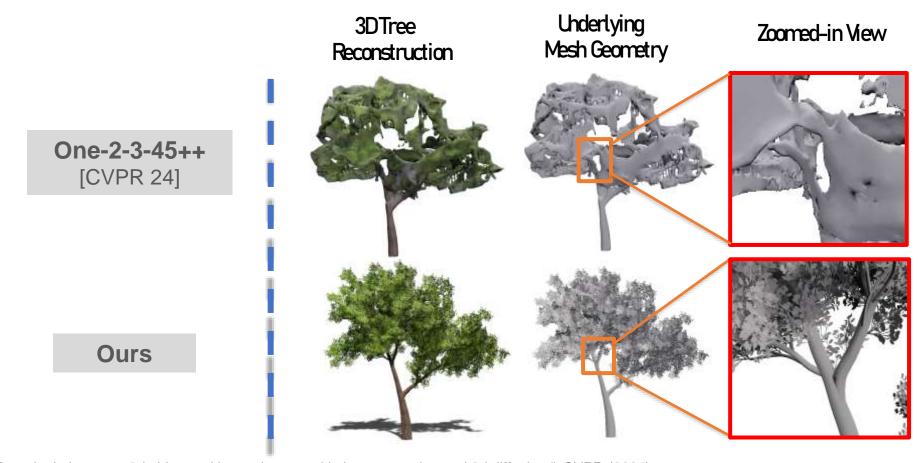
Comparison with a Tree-Specialized method[3]



- [1] "One-2-3-45: Any single image to 3d mesh in 45 seconds without per-shape optimization.", NeurIPS (2023).
- [2] "Wonder3D: Single image to 3d using cross-domain diffusion.", CVPR (2024).
- [3] "Learning to reconstruct botanical trees from single images.", ACM Trans. Graph. (2021)

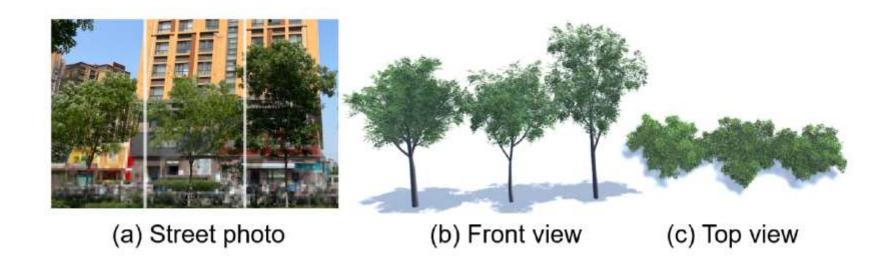
Comparison of Underlying Geometries.

• Our methods can produce high-quality 3D geometries supporting for instant use in downstream CG applications,.



[1] "One-2-3-45++: Fast single image to 3d objects with consistent multi-view generation and 3d diffusion.", CVPR (2024).

Applications.



Reconstruct a scene of street trees.

Applications.





Assemble our results into a 3D scene.

Thanks for watching!