

# Learning Unsupervised Hierarchical Part Decomposition of 3D Objects from a Single RGB Image

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[http://superquadrics.com/hierarchical\\_primitives](http://superquadrics.com/hierarchical_primitives)

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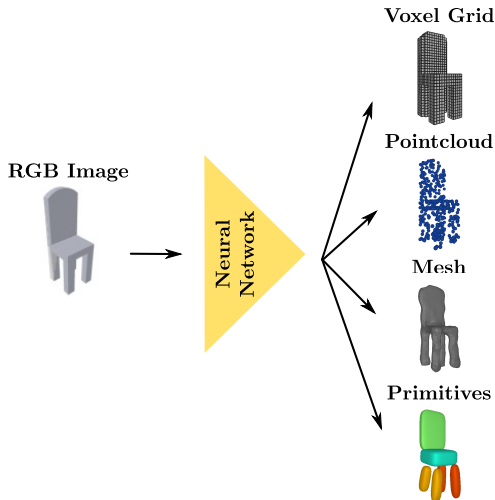


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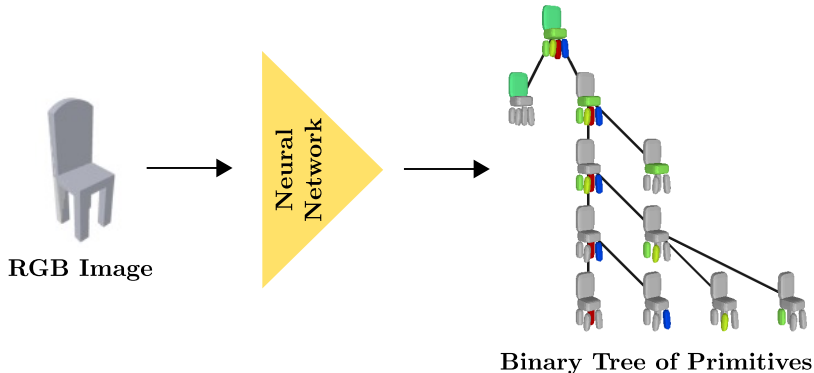
# Motivation

Existing shape representations **focus only on reconstructing the geometry** of a 3D object **without considering its part-based decomposition or relations between parts.**



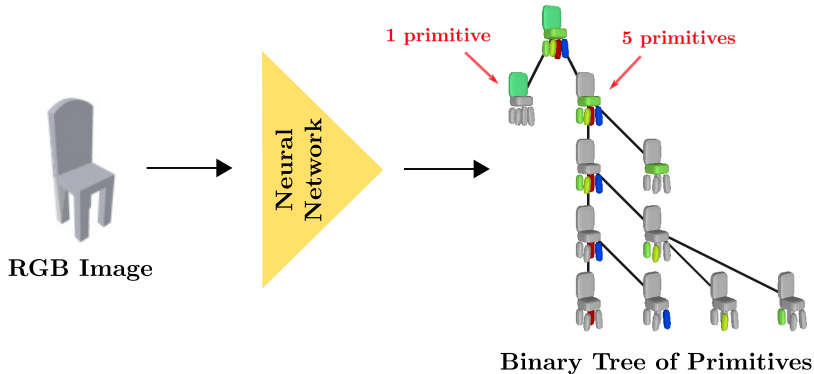
# Our Structure-aware Representation

**Jointly** recovers **the geometry** and **the latent hierarchical layout** of an object as an **unbalanced binary tree of primitives**



# Our Structure-aware Representation

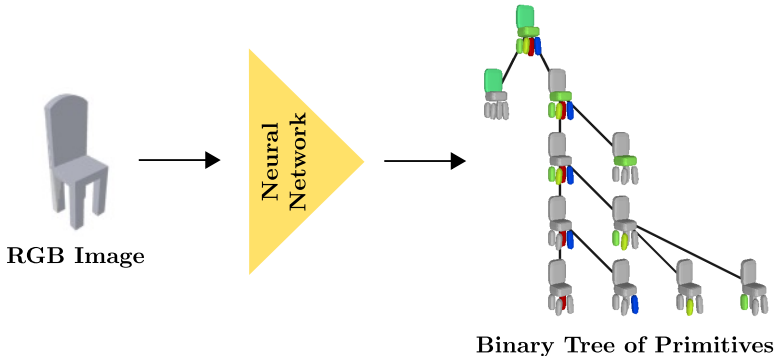
**Jointly** recovers the **geometry** and the **latent hierarchical layout** of an object as an **unbalanced binary tree of primitives**



where **simple parts** are represented with fewer primitives and **complex parts** with more components.

# Our Structure-aware Representation

**Jointly** recovers **the geometry** and **the latent hierarchical layout** of an object as an **unbalanced binary tree of primitives**.

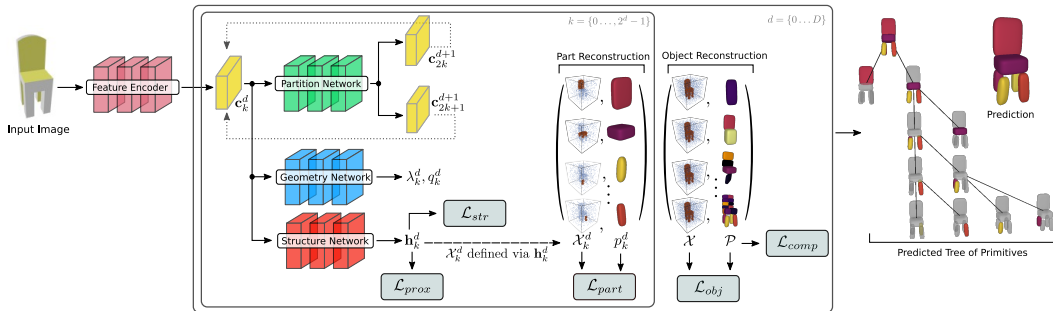


The hierarchical part decomposition is learned **without any supervision** neither on the object parts nor their structure.

# Our Structure-aware Representation

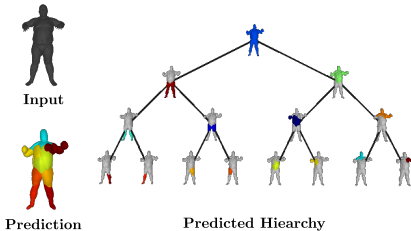
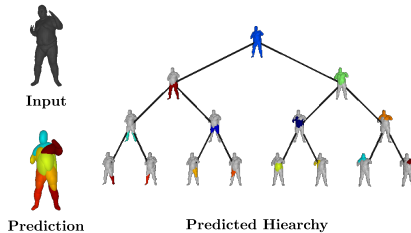
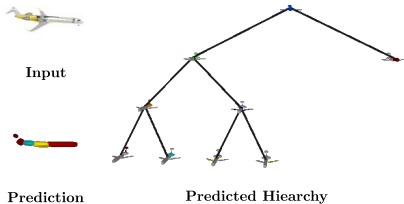
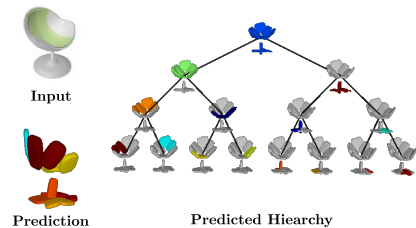
Given an **input** and a **target mesh** represented as a set of occupancy pairs  $\mathcal{X} = \{(\mathbf{x}_i, o_i)\}_{i=1}^N$ , our network predicts a **binary tree of primitives**. For each primitive the network regresses:

- A set of 11 parameters  $\lambda_k^d$  that define the shape, size and position in 3D space of its primitive at each depth level  $d$ .
- A reconstruction quality  $q_k^d$ .



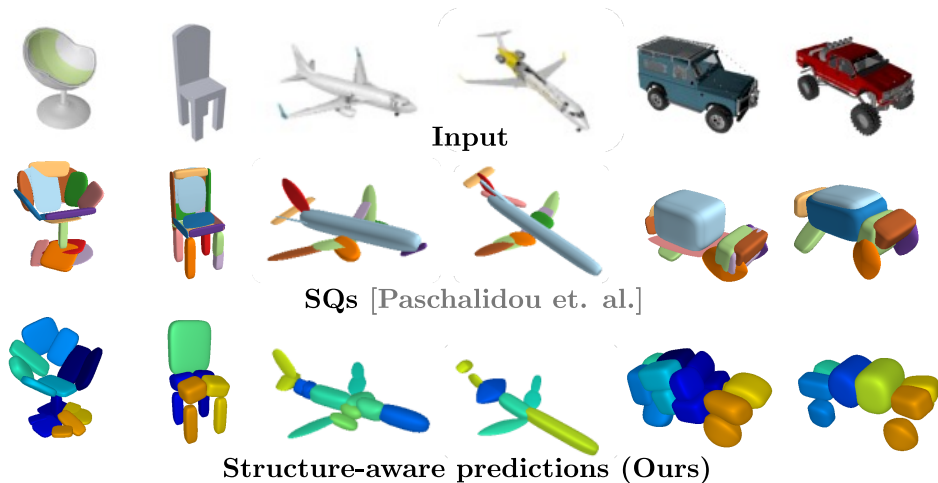
# Expressive Shape Abstractions

We evaluate our model on the single view 3D reconstruction task on ShapeNet and D-FAUST.



# Expressive Shape Abstractions

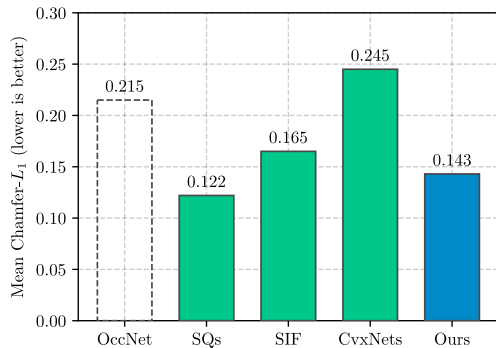
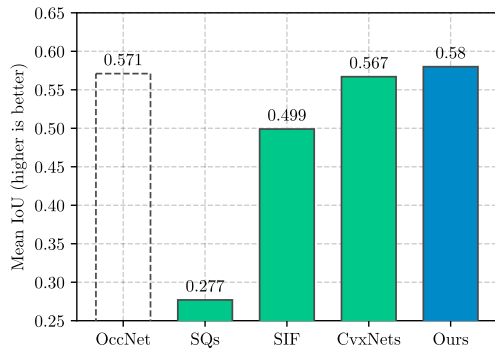
We show that considering the part decomposition improves the reconstruction quality.





# Expressive Shape Abstractions

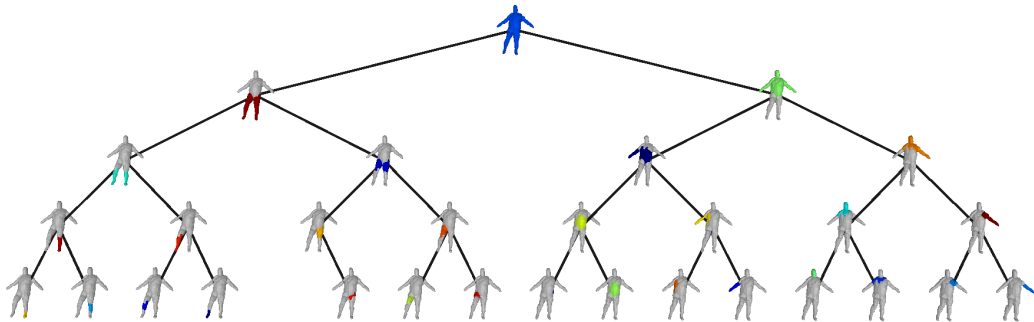
We show that considering the part decomposition improves the reconstruction quality.



 Implicit Shape Representations    Primitive-based Representations    Ours

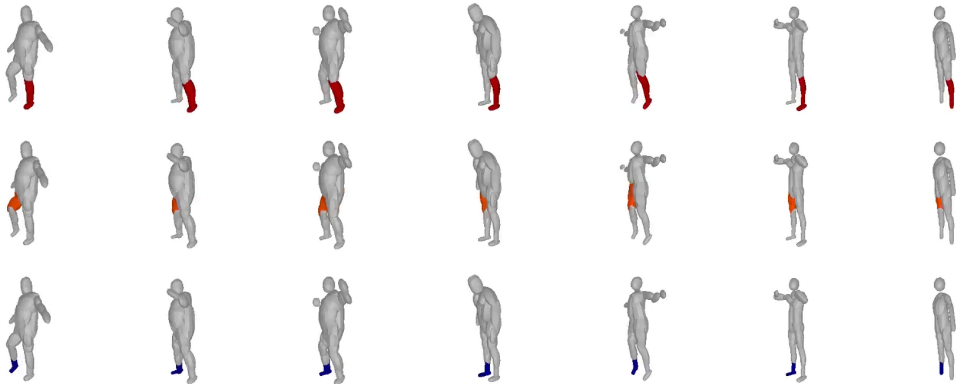
# Semantic Interpretation of Learned Hierarchies

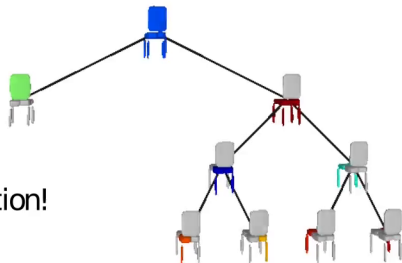
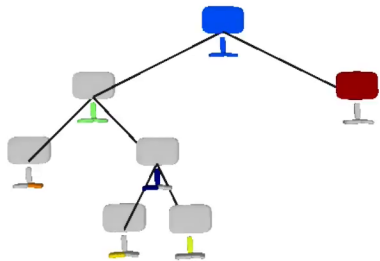
We show that our model recovers **semantic hierarchies** as the same node is consistently used for representing the same object part.



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Thank you for your attention!

Project Page: [http://superquadrics.com/hierarchical\\_primitives](http://superquadrics.com/hierarchical_primitives)

