

# A Study on Different Architectures on a 3D Garment Reconstruction Network

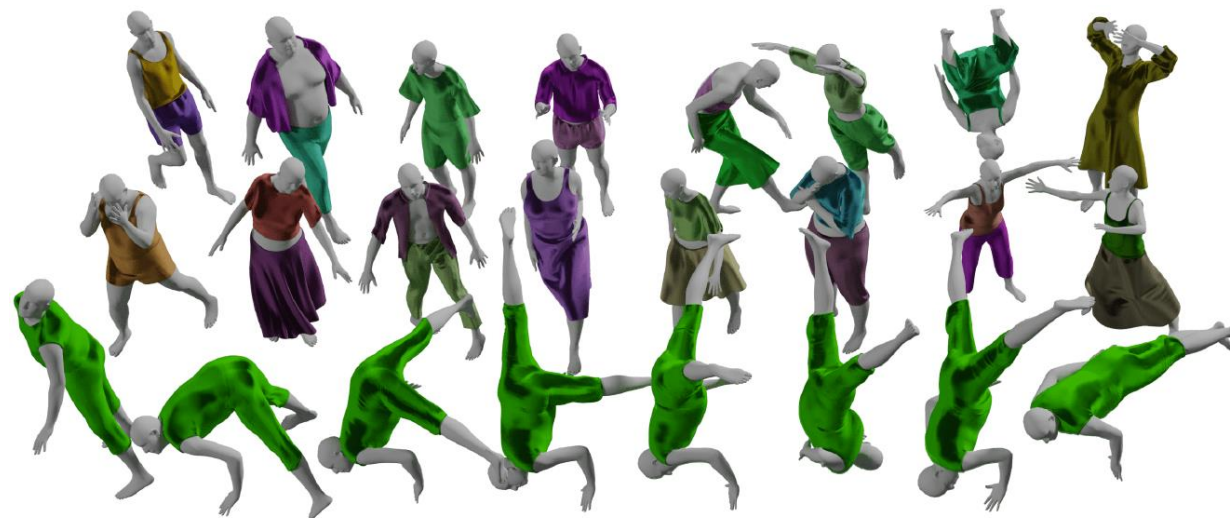
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- Introduction
- Overview of CLOTH3D Architecture
- Network Changes
- Experiments
- Conclusions

- Applications
  - CGI Characters
  - Fashion
- Realistic garments
  - Proper fitting
  - Cloth dynamics
- 3D Garment reconstruction
  - Create realistic 3D garments

- Challenge:
  - Lack of 3D garment data
- 3 Methods to obtain data:
  - 2D Image to 3D model
  - 3D Scans
  - Synthetic Generation
- CLOTH3D Dataset
  - Large-scale synthetic dataset of 3D clothed human sequences
  - 2M+ samples
  - Large variety of garment types, topologies, shapes, sizes, and fabrics
  - SMPL Human body sequences



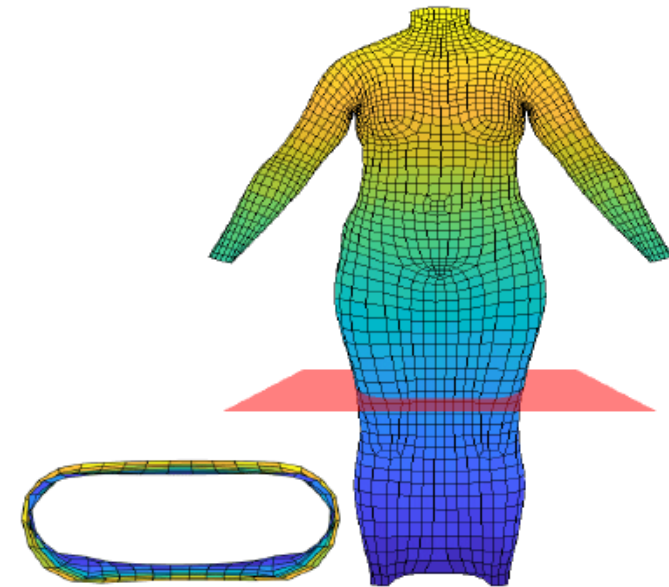
- CLOTH3D Network
  - GCVAE
  - Offsets
  - Skirt/Dress Topology
- Goal:
  - Achieve better understanding of GNNs for clothes simulation
  - Study different architectures for 3D garment reconstruction
- Network Modifications:
  - Pooling
  - Central Difference Convolution
  - Octave Convolution

- CLOTH3D dataset + DL Network
- Challenge:
  - Large variability in Garment type and topology
- Solution:
  - Encode garments as offsets
- Graph Conditional Variational Auto-Encoder (GCVAE)

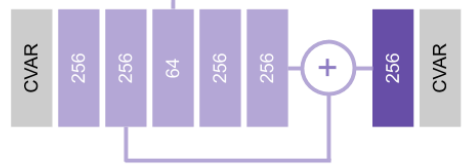
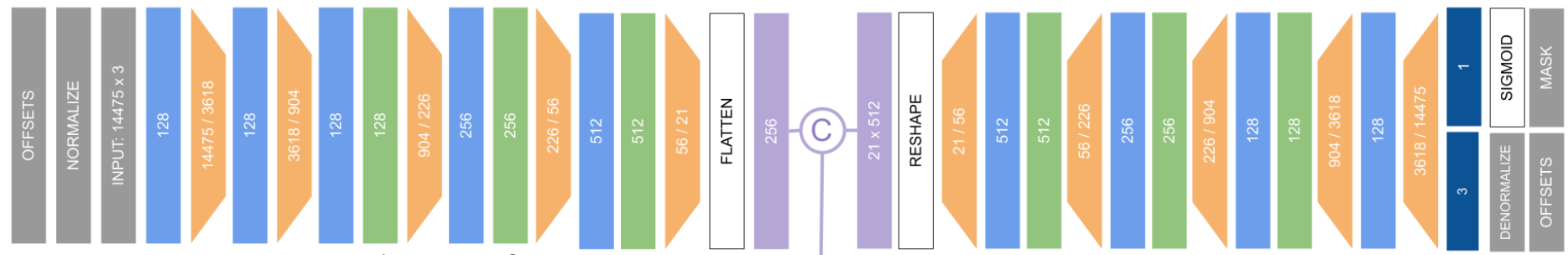


## *Pre-Processing*

- Represent garments as offsets from SMPL body
- Mask used to highlight garment vertices
- Body-garment matching: non-rigid ICP
- Body topology for skirt/dress



# Overview of CLOTH3D Architecture





## *Graph Convolution*

- Spectral Graph Convolution

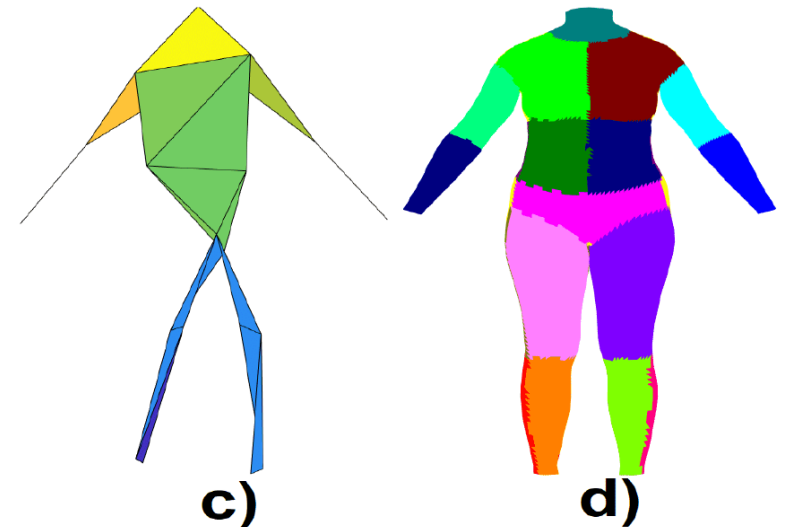
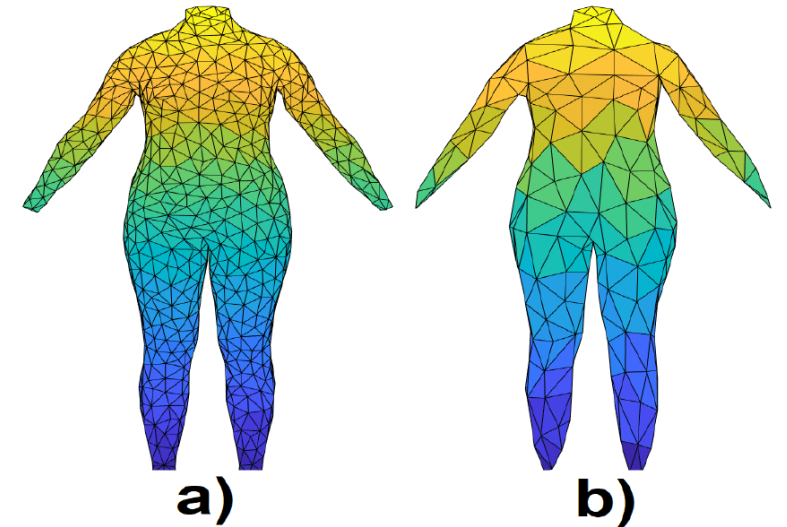
$$y = \sum_{i=0}^K w_i T_i(\hat{L})x$$

- $w_i$ : learnable weights
- $T_i(\hat{L})$ :  $i^{\text{th}}$  Chebyshev polynomial order
- $K$ : node receptive field size
- $K = 1$  is used here

# Overview of CLOTH3D Architecture

## *Pooling*

- Reduce number of vertices in the graph
- Body segmented into 21 regions
- Mesh Hierarchy
  - $14475 \rightarrow 3618 \rightarrow 904 \rightarrow 226 \rightarrow 56 \rightarrow 21$



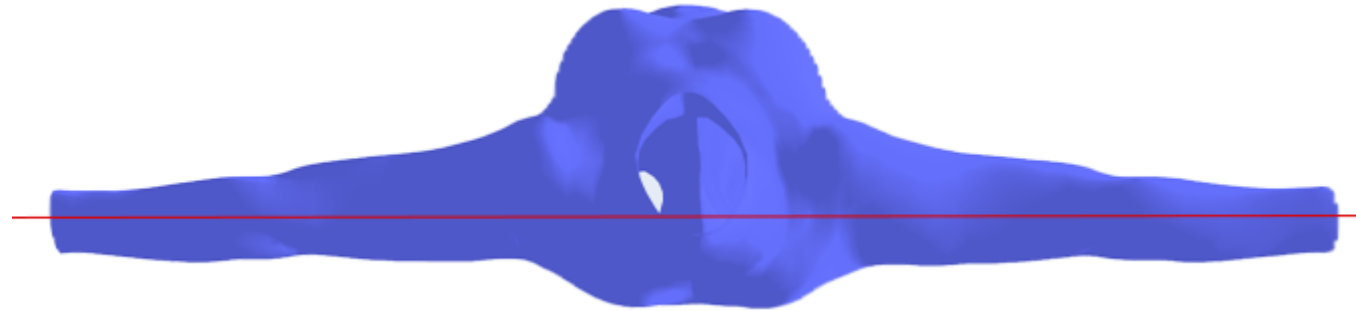
- Goal:
  - Study different architectures for 3D garment reconstruction
- Modifications to CLOTH3D architecture:
  - Pooling
  - Central difference convolution
  - Octave convolution

## *Pooling*

- Objectives:
  - 6 vertices in final layer
    - 2 arms, 2 legs, front and back torso
  - Increase vertex pop. in lower arms and legs
    - These regions have high variability in cloth dynamics
  - Observe effects of alternate pooling
- Four steps:
  1. Split the vertices into front and back
  2. Recursive Spectral clustering
  3. Manually merge final layer
  4. Create the edge connections

## *Pooling*

- 1) Split the vertices into front and back



## *Pooling*

### 2) Spectral Clustering (SC)

- Cluster both groups (front and back) into 5 (total of 10)
  - 5 clusters: arms, legs, torso
- Recursively apply SC to each group
  - Base case: single vertex remaining
- Hierarchy:
  - $14475 \rightarrow 8648 \rightarrow 2558 \rightarrow 640 \rightarrow 160 \rightarrow 40 \rightarrow 10$

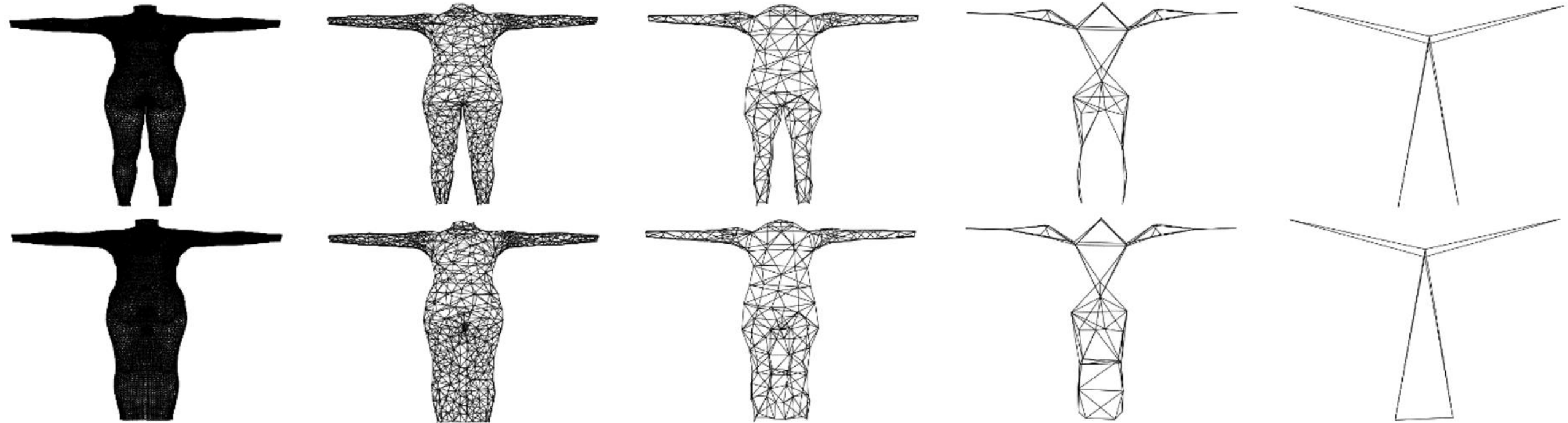
### 3) Manually merge final layer

- Combine arm and leg vertices
- Final Hierarchy:
  - $14475 \rightarrow 8648 \rightarrow 2558 \rightarrow 640 \rightarrow 160 \rightarrow 40 \rightarrow 10 \rightarrow 6$

## *Pooling*

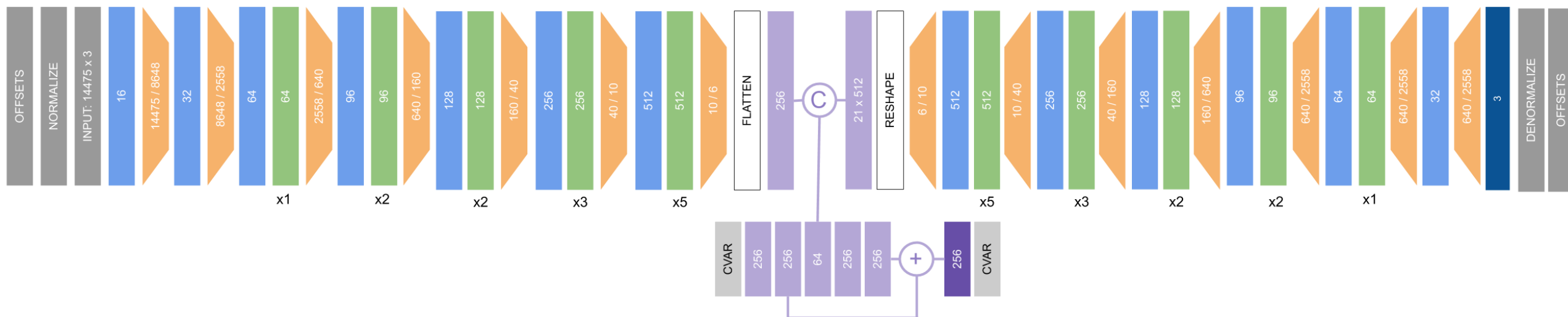
### 4) Create edge connections

- Start with initial edges (from 14475 vertex graph)
- Consider each vertex as a group of vertices pooled together
- If an edge exists between vertices from separate groups, then an edge exists between groups



## Pooling

- Modify architecture layers
  - CLOTH3D: 5 pooling layers
  - New Pooling: 7 pooling layers





## *Central Difference Convolution (CDC)*

- Goal:
  - Enhance vanilla convolution using center oriented gradient of samples
- Vanilla convolution:
  - Weighted sum of sampled region
- Central Difference:
  - Weighted sum of difference between sampled region and center
- Vanilla + central difference = intensity + gradient learning

- CNNs

$$y(p_0) = \underbrace{\sum_{p_n \in R} w(p_n) \cdot (x(p_0 + p_n))}_{\text{Vanilla}} - \theta \cdot x(p_0) \cdot \underbrace{\sum_{p_n \in R} w(p_n)}_{\text{Central Difference}}$$

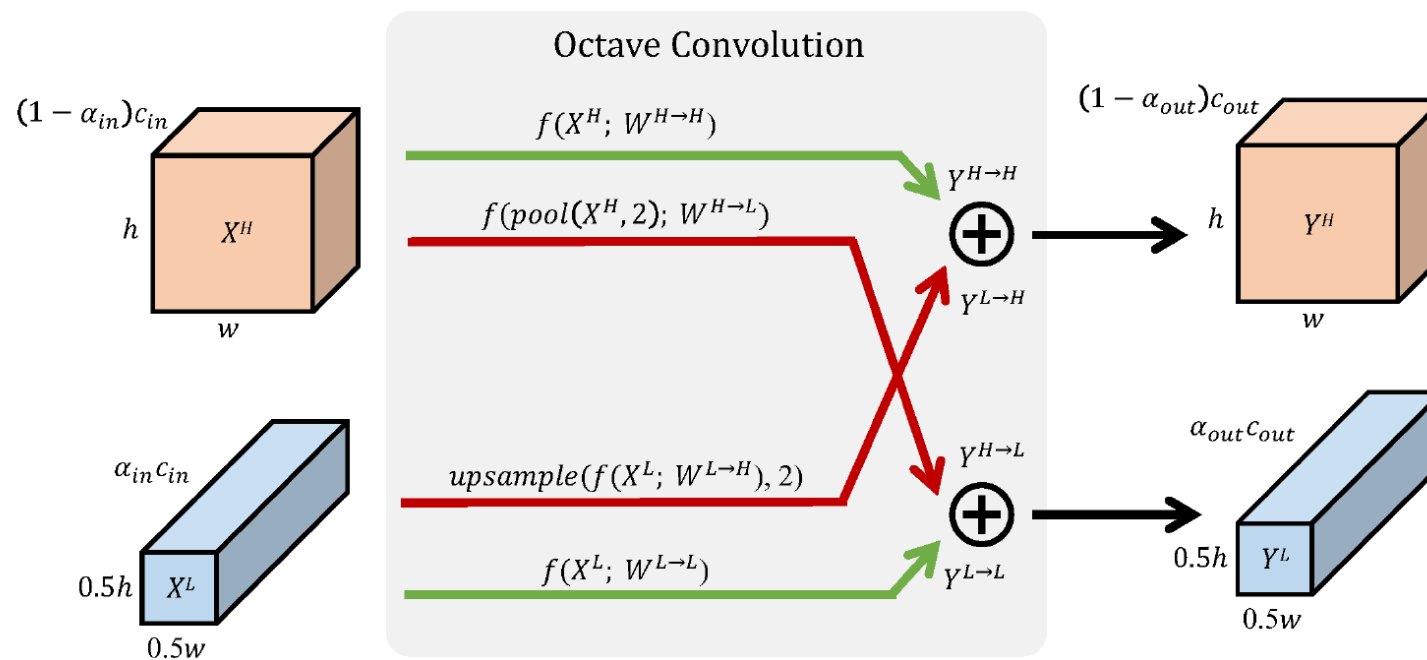
- GNNs

- $\theta = 0.7$
- $K = 1$

$$y = \underbrace{\sum_{i=0}^K w_i T_i(\hat{L}) x}_{\text{GCN}} - \theta \cdot x \cdot w_1 \underbrace{\hspace{1.5cm}}_{\text{Central Difference}}$$

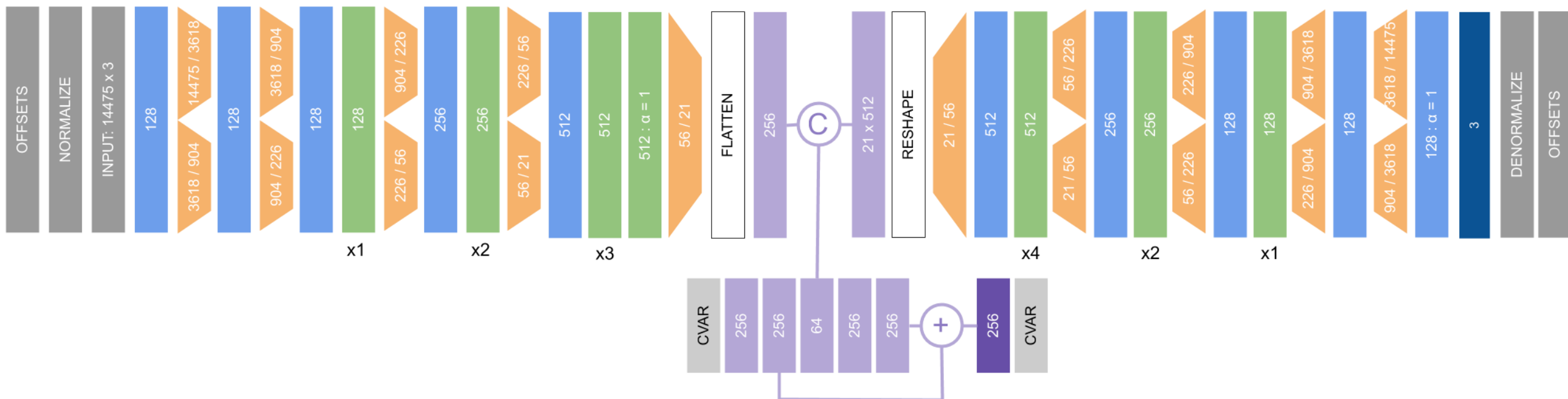
## Octave Convolution

- Goal:
  - Increase efficiency by separating features into high and low components
- $\alpha = 0.875$



# Network Changes

## Octave Convolution



## *Networks*

- Five networks tested:
  - Default
  - CDC
  - Pooling
  - Octave
  - CDC-Pooling
- Quantitative and Qualitative Results

## *Data*

- Subset of CLOTH3D Dataset
- Training: 647 outfits
- Validation: 60 outfits
- 300 frames per outfit
- 6 Garment types
- Fabrics:
  - Silk, denim, cotton, leather



Tshirt



Trousers



Jumpsuit



Dress



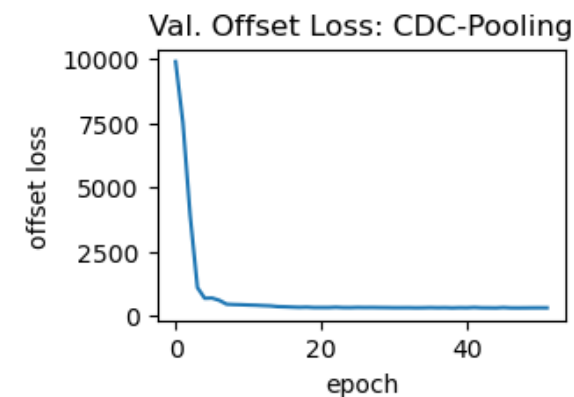
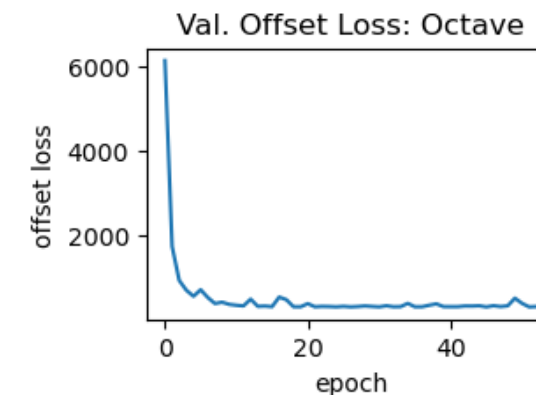
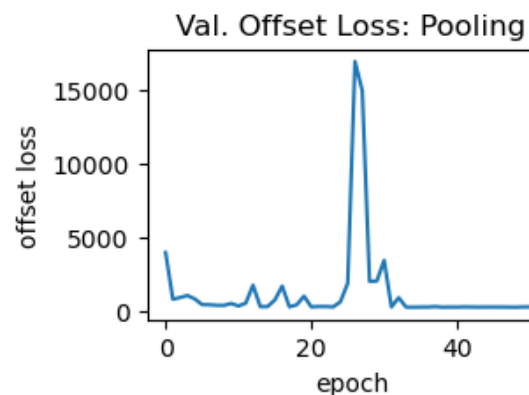
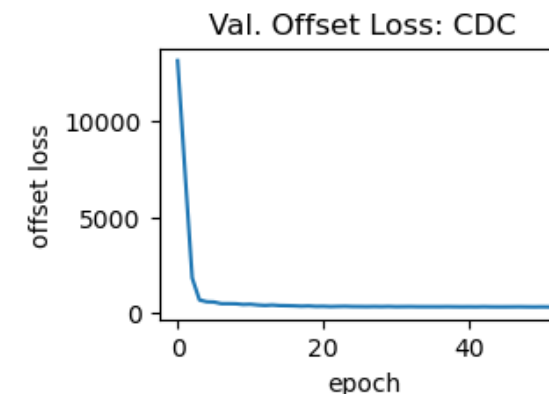
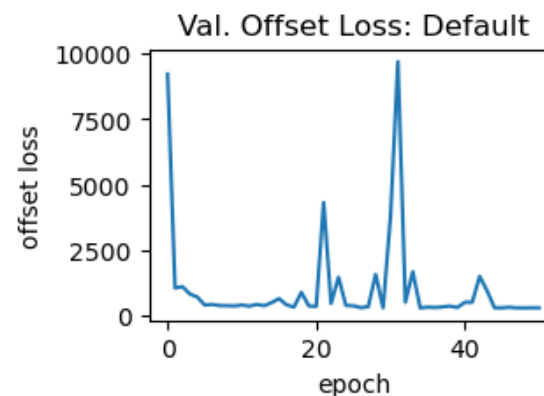
Top



Skirt

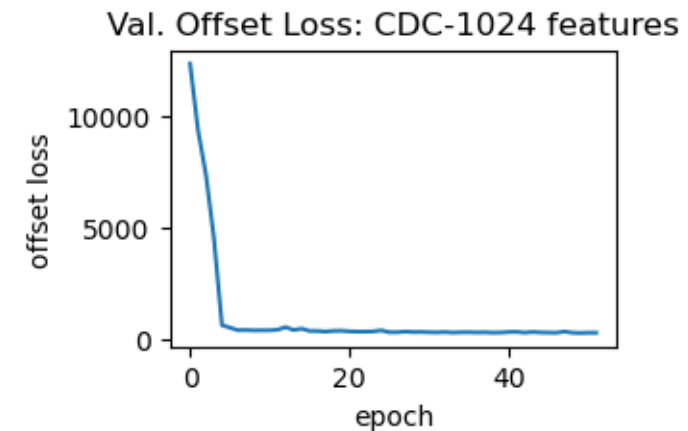
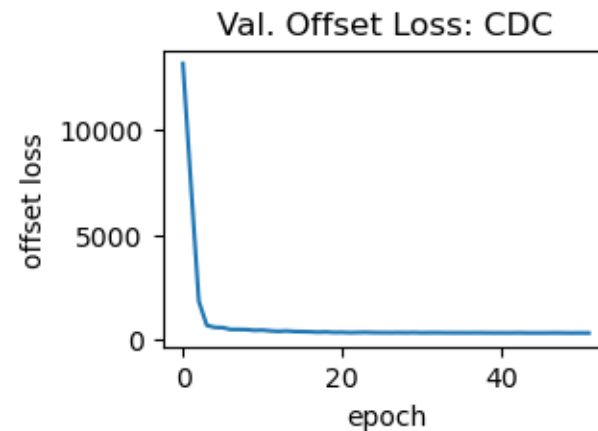
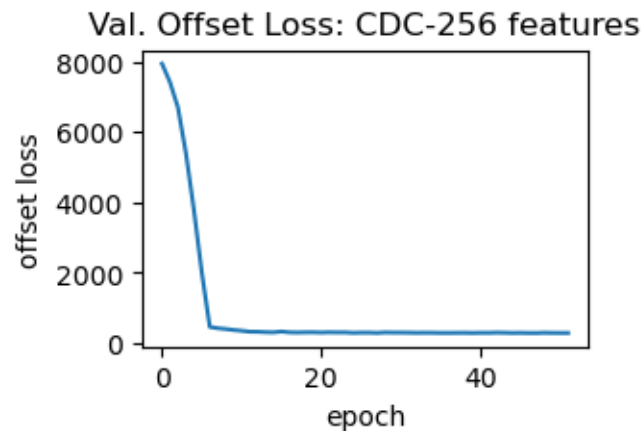
## Training

- NVIDIA GeForce GTX 1080Ti
- Epochs: 52
- LR: 0.0001
- Optimizer: Adam
- Batch Size: 16
- Incremental Training by 200 steps
- Offset Loss



## *Ablation Study*

- Capacity of the network
  - CDC Network
  - 256, 512, 1024 features
- 512 has fastest convergence
- 512 lowest overall loss





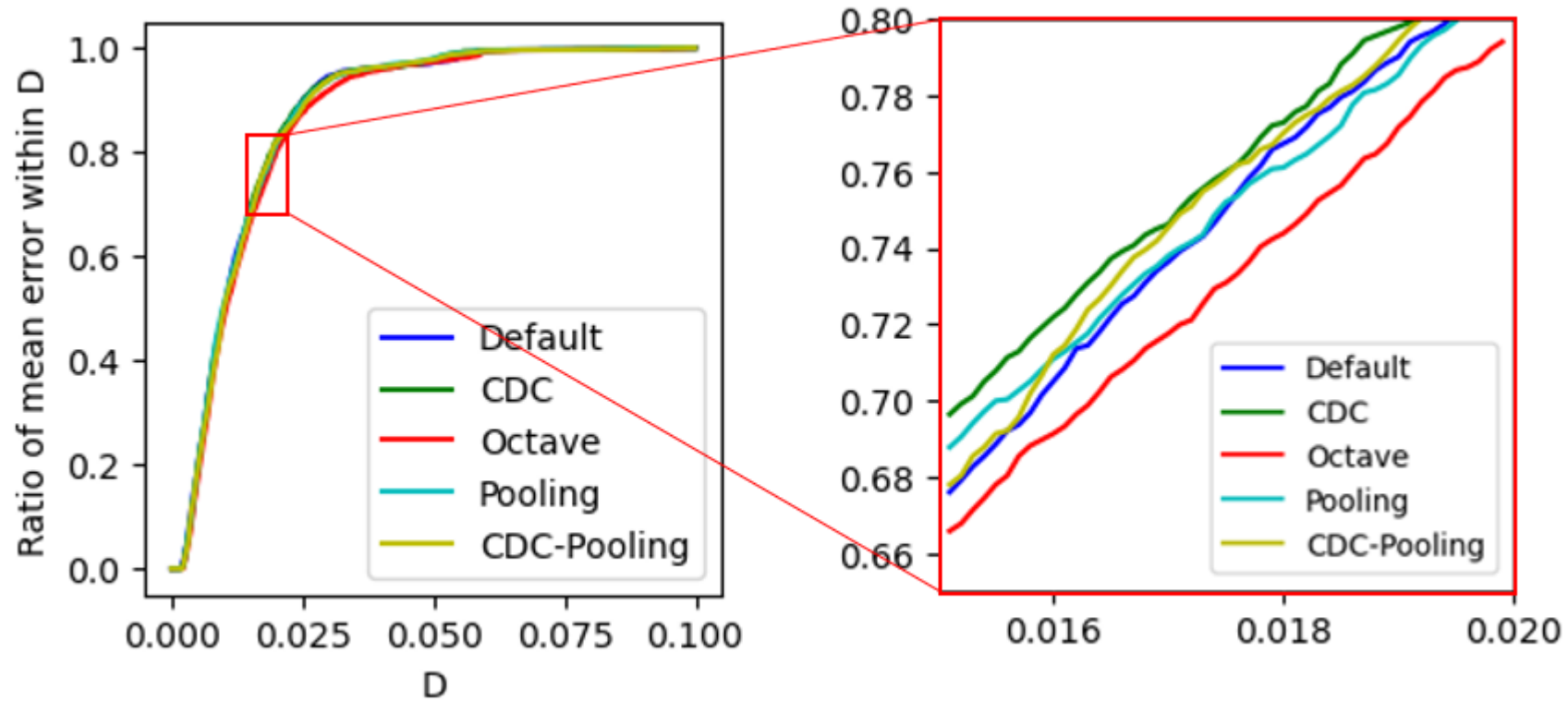
## *Quantitative Results*

	Default	CDC	Pooling	Octave	CDC-Pool
Tshirt	0.0282	0.0281	<b>0.0271</b>	0.0293	0.0282
Trousers	0.0245	0.0254	0.0254	0.0266	<b>0.0241</b>
Jumpsuit	0.0202	0.0204	<b>0.0201</b>	0.0210	0.0202
Dress	0.0375	<b>0.0371</b>	0.0374	0.0404	0.0381
Top	0.0212	0.0211	<b>0.0209</b>	0.0219	0.0230
Skirt	0.0639	<b>0.0610</b>	0.0624	0.0650	0.0641
silk	0.0397	0.0395	<b>0.0394</b>	0.0415	0.0408
denim	0.0306	<b>0.0303</b>	0.0304	0.0316	<b>0.0303</b>
cotton	0.0246	0.0248	<b>0.0245</b>	0.0262	0.0256
leather	0.0280	0.0270	0.0277	0.0297	<b>0.0266</b>
<b>TOTAL</b>	<b>0.0304</b>	<b>0.0302</b>	<b>0.0302</b>	<b>0.0320</b>	<b>0.0308</b>


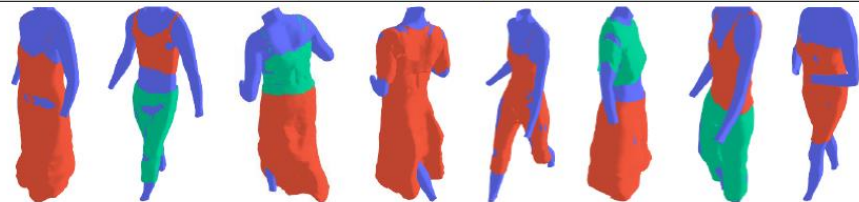

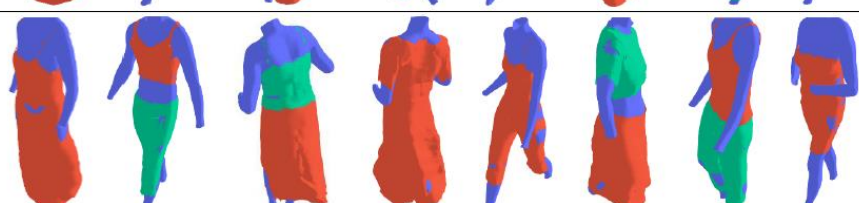
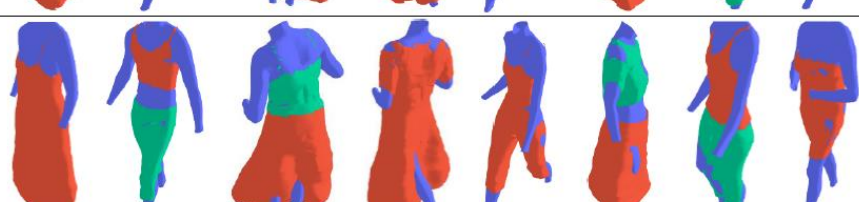
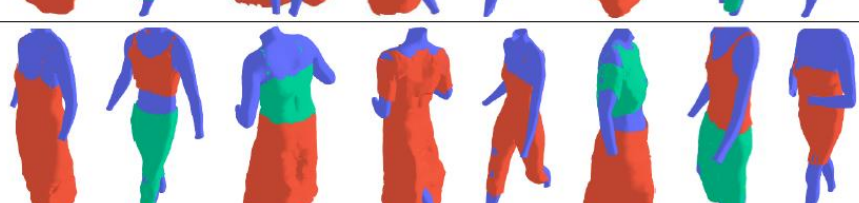
Average per vertex Euclidean error of garment vertices, for each architecture.

## Quantitative Results

- Frame Ratio above Distance Threshold  $D$



## *Qualitative Results*

GT	
Def.	
CDC	
Pool	
Oct.	
CDC + Pool	

- Study of architectures for 3D garment reconstruction
- Five networks compared
- Quantitative results:
  - CDC and Pooling modifications are positive
  - Octave modification negative
- Qualitative Results:
  - Similar results for: Tshirt, Trousers, Top, Jumpsuit
  - Pooling captured the cloth dynamics of the skirt and dress garments best
  - CDC-Pooling had best overall performance
- Future Considerations:
  - Explore and improve the pooling
  - Account for penetrations

Questions?