

## **Point Cloud Upsampling Using Frequency Models**



Friedrich-Alexander-Universität Technische Fakultät Viktoria Heimann viktoria.heimann@fau.de Chair of Multimedia Communications and Signal Processing



## Virtual and Augmented Reality



- Zoom in 3D objects
- Point Cloud Upsampling is necessary



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## **Point Cloud**





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# **Point Clouds**



- Single points
  - Geometry (x,y,z)
  - Attribute (RGB)

→ Geometry Upsampling

→ Color Upsampling



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#### Input Point Cloud





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## **Block Partitioning**



![](_page_5_Picture_2.jpeg)

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![](_page_5_Picture_5.jpeg)

![](_page_5_Picture_6.jpeg)

## Input Block

![](_page_6_Figure_1.jpeg)

![](_page_6_Picture_2.jpeg)

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![](_page_6_Picture_5.jpeg)

## Geometry

Aim:

Generate model that estimates the surface of the point cloud

$$q = f[o, p] = \sum_{k,l \in \mathcal{K}} c_{k,l} \varphi_{k,l}[o, p]$$
  
Expansion coefficient

![](_page_7_Picture_4.jpeg)

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![](_page_7_Picture_7.jpeg)

![](_page_7_Picture_8.jpeg)

## Frequency-Selective Geometry Upsampling

![](_page_8_Figure_1.jpeg)

![](_page_8_Picture_2.jpeg)

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![](_page_8_Picture_5.jpeg)

#### **Model Generation**

![](_page_9_Figure_1.jpeg)

![](_page_9_Picture_2.jpeg)

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![](_page_9_Picture_6.jpeg)

#### **Continuous Model and Input Points**

![](_page_10_Figure_1.jpeg)

![](_page_10_Picture_2.jpeg)

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![](_page_10_Picture_5.jpeg)

![](_page_10_Picture_6.jpeg)

### **Delaunay Triangulation**

![](_page_11_Figure_1.jpeg)

![](_page_11_Picture_2.jpeg)

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![](_page_11_Picture_6.jpeg)

![](_page_11_Picture_7.jpeg)

### **Delaunay Triangulation**

![](_page_12_Figure_1.jpeg)

![](_page_12_Picture_2.jpeg)

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![](_page_12_Picture_6.jpeg)

#### Continuous Model with In- and Output Points

![](_page_13_Figure_1.jpeg)

![](_page_13_Picture_2.jpeg)

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![](_page_13_Picture_5.jpeg)

![](_page_13_Picture_6.jpeg)

## **Frequency-Selective Geometry Upsampling**

![](_page_14_Figure_1.jpeg)

V. Heimann et al., "Frequency-Selective Geometry Upsampling of Point Clouds," accepted for ICIP 2022.

![](_page_14_Picture_3.jpeg)

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![](_page_14_Picture_6.jpeg)

## **Error Metrics**

![](_page_15_Figure_1.jpeg)

$$e_{\mathbf{A},\mathbf{B}}^{\text{P2Point}} = \frac{1}{N_A} \sum_{\forall a_j \in \mathbf{A}} \|E(i,j)\|_2^2$$

$$e_{A,B}^{P2Plane} = \frac{1}{N_A} \sum_{\forall a_j \in A} (E(i,j) \cdot N_j)^2$$

Tian et al., "Geometric distortion metrics for point cloud compression," ICIP 2017.

![](_page_15_Picture_5.jpeg)

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![](_page_15_Picture_9.jpeg)

![](_page_15_Picture_10.jpeg)

## Results

Scale factor	Metric	Method	Dragon	Duck	Jaguar	Rabbit	camel	elephant	kitten	star	6	21	45	76	Avg
2	$\begin{array}{c} \textbf{P2Point} \\ \times 10^{-3} \end{array}$	EAR [1]	11.3	9.5	14.4	29.6	13.6	8.7	8.3	7.4	7.5	7.4	7.9	7.6	11.1
		PU [2]	13.4	12.7	12.4	9.0	15.3	17.1	18.0	22.4	13.3	13.6	14.0	13.0	14.5
		FSGU (Ours)	1.9	3.8	1.5	2.9	3.5	3.6	3.6	3.8	3.1	2.5	3.3	2.6	3.0
	P2Plane $\times 10^{-3}$	EAR [1]	5.0	4.3	6.6	9.2	2.8	2.0	1.7	0.7	3.0	3.6	3.5	4.0	3.9
		PU [2]	11.3	8.3	10.1	7.3	7.5	8.6	7.0	7.6	7.1	8.9	9.4	8.6	8.5
		FSGU (Ours)	0.5	0.6	0.4	1.3	0.5	0.5	0.4	0.5	0.5	0.5	0.6	0.4	0.6
4	$\begin{array}{c} \text{P2Point} \\ \times 10^{-3} \end{array}$	EAR [1]	27.8	12.7	16.6	12.1	17.9	20.6	16.4	15.2	18.9	14.6	14.0	14.3	16.8
		PU [2]	11.7	11.4	11.4	8.7	16.6	16.6	16.7	16.6	12.9	12.9	14.0	12.6	13.5
		FSGU (Ours)	1.5	3.5	1.2	2.8	3.4	3.5	3.8	4.0	3.7	3.0	4.1	3.0	3.1
	$\begin{array}{c} \textbf{P2Plane} \\ \times 10^{-3} \end{array}$	EAR [1]	9.0	5.0	6.6	3.7	4.0	5.4	4.5	4.5	6.3	5.4	4.9	5.8	5.4
		PU [2]	6.0	6.3	6.5	4.5	6.4	6.8	7.2	7.3	5.8	6.1	6.4	7.2	6.4
		FSGU (Ours)	0.4	0.4	0.4	1.4	0.4	0.4	0.4	0.5	0.5	0.6	0.5	0.5	0.5

[1] H. Huang et al., "Edge-aware point set resampling," ACM Transactions on Graphics, 2013. [2] L. Yu et al. "PU-Net: Point Cloud Upsampling Network," CVPR 2018. Point Clouds taken from: Nouri et al., "Technical report : Greyc 3D colored mesh database," 2017., [2], and Z.Lian et al., "SHREC'11 track: shape retrieval on non-rigid 3d watertight meshes," 2011. Metric taken from Tian et al., "Geometric distortion metrics for point cloud compression," ICIP 2017.

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![](_page_16_Picture_5.jpeg)

![](_page_16_Picture_6.jpeg)

## Visual Results – Dragon

![](_page_17_Figure_1.jpeg)

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![](_page_17_Picture_4.jpeg)

# Conclusion

Frequency-selective model estimation

- uses
  - Expansion coefficients
  - Basis functions
- works
  - Block-based
  - Iteratively
- is successfully applied to
  - Geometry Upsampling

![](_page_18_Picture_10.jpeg)

![](_page_18_Picture_11.jpeg)

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![](_page_18_Picture_15.jpeg)