

BlendPCR Seamless and Efficient Rendering of Dynamic Point Clouds captured by Multiple RGB-D Cameras



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- Rendering dynamic point cloud is crucial for many VR and XR applications, e.g.:
 - Telepresence
 - Point cloud avatars
 - XR telemedicine
 - Performance capture and live performances
 - In general: Visualization of dynamic RGB-D data



Yu et al. (2021)



Gasques et al. (2021)

Results

Conclusion









Multiple RGB-D cameras are often used to capture a scene more completely



Motivation

Problem

Method





Results

Conclusion







• However, rendering point clouds by multiple RGB-D cameras currently leads to visible artifacts which we call **seam-flickering**



Uniform Splatting

Motivation

Problem

Method

Separate Meshes

Results

Conclusion







- - Due to (a) different specular reflections, (b) white balance and (c) slightly different color gamuts of each camera



Splats

Separate Meshes

Motivation

Problem

Method

Seam-flickering also occurs in leading-edge rendering techniques



Pointersect (2023) Apple Machine Learning Research



P2ENet (2024) based on Gaussian Splatting

Results

Conclusion







Our Method

Motivation

Problem

Method

Results

Conclusion





- BlendPCR
 - A two-step approach to render dynamic point clouds
 - Step 1: Create separate surfaces for each camera and render them to individual framebuffers
 - Step 2: Selectively blend these individual framebuffers
 - This avoids seams and seam-flickering

Results

Conclusion









Motivation

Problem

Method



Results

Conclusion







- a) Create a mesh in naïve order of the depth image
 - High precision & very performant



(disjoined)



Results

Conclusion







- a) Create a mesh in naïve order of the depth image
 - High precision & very performant
- b) Smooth edges via Moving Least Squares filter
 - To reduce sharp edges and noise



Results

Conclusion







- a) Create a mesh in naïve order of the depth image
 - High precision & very performant
- b) Smooth edges via Moving Least Squares filter
 - To reduce sharp edges and noise
- **c)** Estimate surface normal
 - Using Cholesky decomposition [1]

[1] J. Klein, G. Zachmann, "Proximity Graphs for Defining Surfaces over Point Clouds," SPBG 2004.



Results

Conclusion

Future Work



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- d) Estimate Accuracy for each vertex, based on:
 - Distance to camera
 - Surface normal
 - Proximity to the edge



Results

Conclusion







- d) Estimate Accuracy for each vertex, based on:
 - Distance to camera
 - Surface normal
 - Proximity to the edge
- e) Render to individual framebuffer





Results

Conclusion









Step 2: Selectively blend Separate Surfaces



Motivation

Problem

Method



Conclusion

Future Work



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Step 2: Selectively blend separate surfaces a) Choose the major camera based on estimated accuracy



Motivation

Problem

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Results

Conclusion









Step 2: Selectively blend separate surfaces

- a) Choose the major camera based on estimated accuracy
- b) Calculate camera weights in screen space
 - For seamless transitions

Problem



Results

Conclusion









- **Step 2:** Selectively blend separate surfaces
 - a) Choose the major camera based on estimated accuracy
 - b) Calculate camera weights in screen space
 - For seamless transitions
 - **c)** Blend separate meshes based on **(b)**
 - Prevents seam-flickering, only draws the information with the highest accuracy, and no blurring

Motivation

Problem

Method



Results

Conclusion





U Overview of our Pipeline



Motivation

Problem

Method

Conclusion





U High Resolution (HR) Encoding

Further improving quality

Motivation

Problem

Method

Results

Conclusion





High Resolution (HR) Encoding

- When working with RGB-D cameras
 - Depth image can be mapped onto the color image
 - > Point Cloud gets huge; bad for performance
 - Color image can be mapped onto the depth image

> Color resolution is (significantly) reduced





2048 x 1536



2048 x 1536

2048 x 1536



640 x 576 640 x 576

Results

Conclusion









High Resolution (HR) Encoding

- Texture mapping
 - Using high resolution color image as texture for low resolution point cloud
 - > Is not a problem, since we work with meshes
- But how to get the UV coordinates?



2048 x 1536

Results

Conclusion

Future Work

use as







U High Resolution (HR) Encoding

• Obtain UV coordinates:

> We can transform an image of UV coordinates onto the depth image

> However, the Azure Kinect SDK only allows

to transform RGBA32 images onto depth

> We can encode 2x 16 bit UV into 4 x 8 bit RGBA

 \geq However, due to linear interpolation, this information is destroyed







Conclusion







Interpolation-Resistant Encoding Scheme

- We created a novel interpolationresistant encoding scheme
 - To encode 2 x 16 bit UV coordinates into 4 x 8 bit RGBA values
 - This finally allows us to generate UV coordinates using arbitrary SDK's colorto-depth transformation functionality.
 - Details in the paper



Conclusion







U High Resolution (HR) Encoding



Low Resolution Color, 640 x 576

Motivation

Problem

Method



High Resolution Color, 2048 x 1536

Results

Conclusion







Results

Motivation

Problem

Method

Results

Conclusion





- Tested on CWIPC-SXR dataset [2]
 - Used seven Microsoft Azure Kinect
 - 45 social XR scenarios



[2] I. Reimat et al., "CWIPC-SXR: Point Cloud Dynamic Human Dataset for Social XR," ACM MMSys 2021.

Motivation

Problem

Method

[2]

Results

Conclusion







• CWIPC-SXR, S13 Card Trick Scene



Separate Meshes

Motivation

Problem

Method

BlendPCR

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Conclusion







CWIPC-SXR, S7 Scarf Dressing



[3] From: Y Hu et al., "Low Latency Point Cloud Rendering with Learned Splatting," CVPR Workshop 2024.

Motivation

Problem

Method





Splats



Separate Meshes



Apple Machine Learning Research



based on Gaussian Splatting





Results

Conclusion







• CWIPC-SXR, S3 Flight Attendant

Uniform Splats

Separate Meshes

TSDF 512³

BlendPCR (HR)



[3] From: Y Hu et al., "Low Latency Point Cloud Rendering with Learned Splatting," CVPR Workshop 2024.

Motivation

Problem

Method



Splats



Separate Meshes



TSDF 512



Pointersect (2023) [3] Apple Machine Learning Research



P2ENet (2024) [3] based on Gaussian Splatting



Ours (CGVR)

Results

Conclusion









Using an NVIDIA GeForce RTX 4090 @ 3580 x 2066

Method





Performance



Method

Results

Conclusion





U Source Code

• Github: <u>https://github.com/muehlenb/blendpcr</u>

Pure OpenGL 3.3 implementation > Pre-built binaries available



Motivation

Problem

Method



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12 main • 2 Branches 🛇 0 Tags	Q Go to file t Add file •	<> Code •
ig muehlenb Fixed typo.	0e22bc2 · last week	3 71 Commits
ata/model	Started to transform our PCR framework into a BlendPCR-on	last month
images	Better screenshot.	3 weeks ago
🖿 lib	Improving integration of libraries.	3 weeks ago
shader	GLSL compatibility fix for some graphics cards & Streaming f	3 weeks ago
src	GLSL compatibility fix for some graphics cards & Streaming f	3 weeks ago
🗋 .gitignore	Added build folder to gitignore.	3 weeks ago
CMakeLists.txt	Updated CMakeLists.txt (k4a) and Readme.md	3 weeks ago
	Added EUROPEAN UNION PUBLIC LICENCE v. 1.2 license	2 weeks ago
C Readme.md	Fixed typo.	last week

README IVPL-1.2 license

BlendPCR: Seamless and Efficient Rendering of Dynamic Point Clouds captured by Multiple RGB-D Cameras

Video (available soon) | Paper (available soon)

C++/OpenGL implementation of our real-time renderer BlendPCR for dynamic point clouds derived from multiple RGB-D cameras. It combines efficiency with high-quality rendering while effectively preventing common z-fightinglike seam flickering. The software is equipped to load and stream the CWIPC-SXR dataset for test purposes and comes with a GUI

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Pre-built Binaries

If you only want to test the BlendPCR renderer, without editing the implementation, we also offer pre-built binaries:

• Download Windows (64-Bit), without CUDA for all graphic cards.

Build Requirements

Required:

- CMake ≥ 3.11
- **OpenGL** ≥ 3.3
- C++ Compiler, e.g. MSVC v143
- Azure Kinect SDK 1.4.1: Required to load and stream the CWIPC-SXR dataset.

Results

Conclusion





U Conclusion

> Novel rendering technique for dynamic point clouds of multiple RGB-D camera

No seam-flickering or seams

> Always uses the most accurate data

Very performant & applicable for VR

Encoding scheme for high resolution textures

Motivation

Problem

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Results

Conclusion





U Future Work

> Objective measurement to state-of-the-art techniques



Pointersect (2023) Apple Machine Learning Research



Subjective comparison study in VR environment

Motivation

Problem

Method

based on Gaussian Splatting



Results

Conclusion





Future Work

Finally: Integrate with other State-of-the-Art techniques



Motivation

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U Discussion

Thank you for your attention!



Separate Meshes

Contact: <u>muehlenb@uni-bremen.de</u>

Motivation

Problem

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BlendPCR

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