Neural Scene Graph Rendering

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NVIDIA

Credit to CS348i Journalist Slides



The G

The Key Parts

- Neural graph representation: Enables the design of scenes composed of neural primitives
- Lifted transformation space: Allows independent manipulation of orthogonal material and geometry attributes of the neural primitives
- Streaming Renderer: Processes neural scene graph to render RGB images with differentiable operations

Neural Graph Representation

Traditional Scene Graph



Neural Scene Graph





Lifted Transformation Space

Forming Object Representations



Lifted Translation



Lifted Hue Shift



Lifted Color Transformation



Streaming Renderer

Neural Scene Graph





Input to Streaming Renderer



Streaming Renderer



Optimization



Results

Results



Orthogonality



Temporally Stable Interpolation



3D Results



Occlusion



Scalability





Generalizing to New Material-Geometry Pairings

Training Data Examples



Diffuse Prediction



Volumetric Prediction



Summary

- Neural graph representation: Enables the design of scenes composed of neural primitives
- Lifted transformation space: Allows independent manipulation of orthogonal material and geometry attributes of the neural primitives
- Streaming Renderer: Processes neural scene graph to render RGB images with differentiable operations
- Results demonstrated many desirable properties on 2D and 3D

Challenges

- Limited to training data and simple scenes
- Struggles with interactions between dynamic scene elements
- Currently only handles static lighting
- Does not integrate any graphics priors into rendering