



TRAVERSE RESEARCH

**Machine Learning-Enhanced Graphics:
an overview**

NEXT SLIDE



MACHINE LEARNING-ENHANCED GRAPHICS: AN OVERVIEW

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GOAL

- 01 Photorealistic graphics
- 02 High resolution
- 03 High refresh rate
- 04 Scalable to a variety of devices



HARDWARE SOLUTIONS

01

- Specialized hardware for graphics
- Nvidia RT cores

02

- Cloud solution
- Nvidia CLOUD XR

03

- Specialized hardware for Machine Learning
- Nvidia Tensor Cores
 - Google Tensor



NVIDIA CLOUDXR

Streaming for Extended Reality



NEURAL NETWORKS

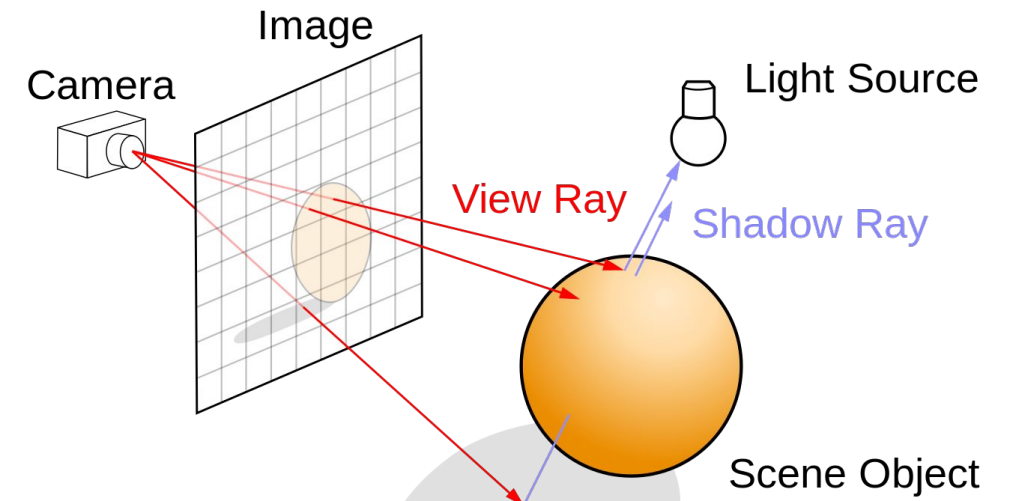
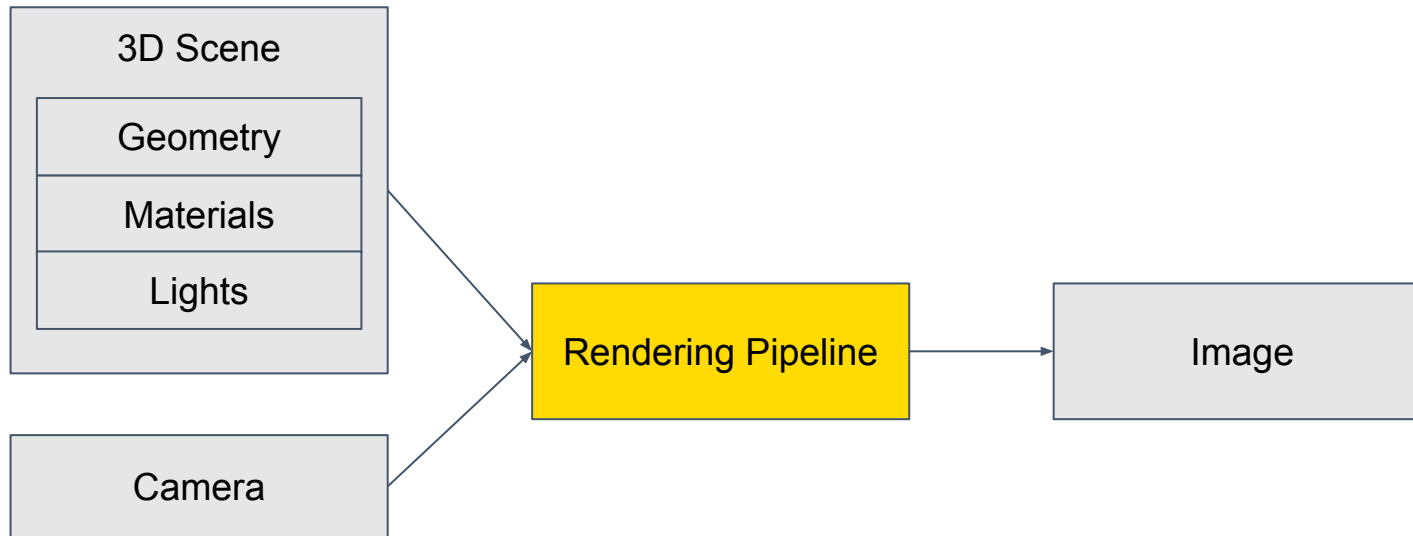
- 01 Black box
- 02 Fixed size
- 03 Fixed compute

People with no idea about AI, telling me my AI will destroy the world

Me wondering why my neural network is classifying a cat as a dog..



RENDERING PIPELINE



Ray Tracing
[https://en.wikipedia.org/wiki/Ray_tracing_\(graphics\)](https://en.wikipedia.org/wiki/Ray_tracing_(graphics))

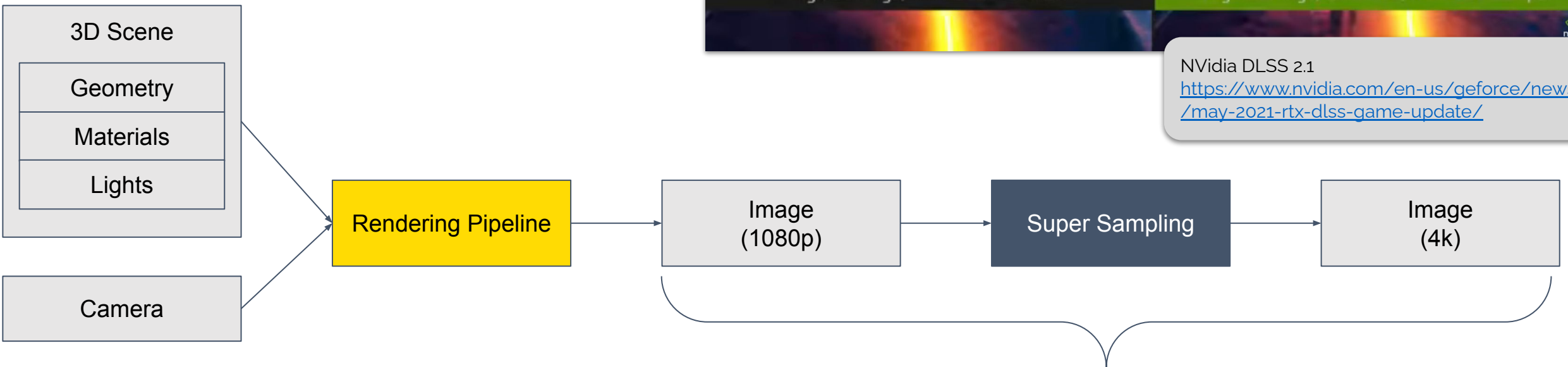
SUPER SAMPLING

01 Render at lower resolution = higher frame rate

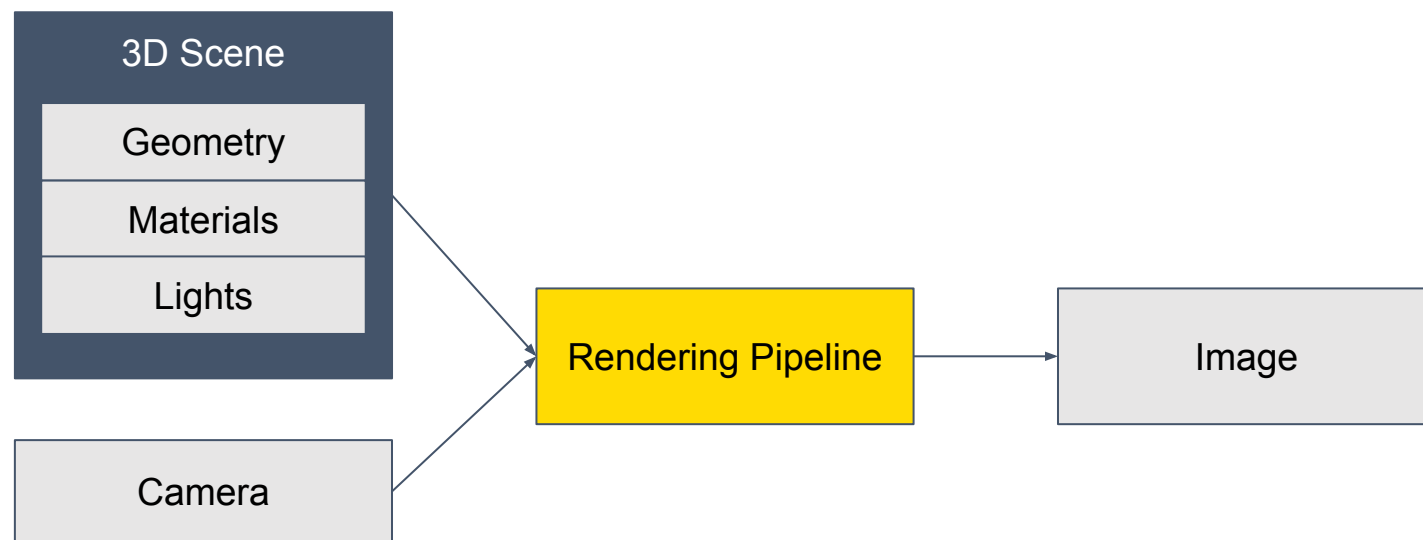
02 Neural Network increases the resolution



NVIDIA DLSS 2.1
<https://www.nvidia.com/en-us/geforce/news/may-2021-rtx-dlss-game-update/>



NEURAL SCENE



01 inverse graphics: 3D scene from an image

02 Novel view synthesis: given multiple images create a 3D scene

03 NeRF

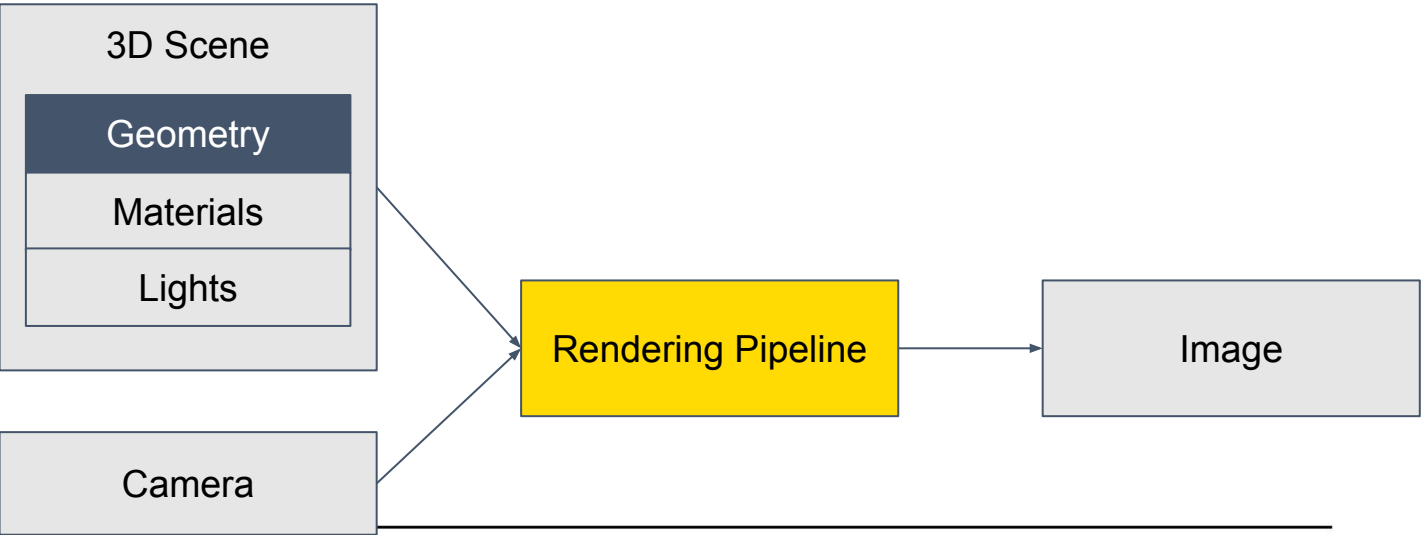


Mildenhall, B., (2020).
NeRF: Representing scenes as neural radiance fields for view synthesis.

NEURAL GEOMETRY

01 ACORN:
compress signal with neural network

02 autoLOD
reduces the amount of triangles



		Uncompressed	Compressed	ACORN
img.	Pluto (8192×8192)	n/a	192 MB	38 MB
	Tokyo (19456×51200)	2.8 GB	169 MB	670 MB
shapes	Lucy (14 M vert.)	1.2 GB	380 MB	68 MB
	Dragon (930 K vert.)	66 MB	27 MB	68 MB
	Thai Statue (5 M vert.)	424 MB	126 MB	68 MB
	Engine (308 K vert.)	16 MB	4.4 MB	68 MB



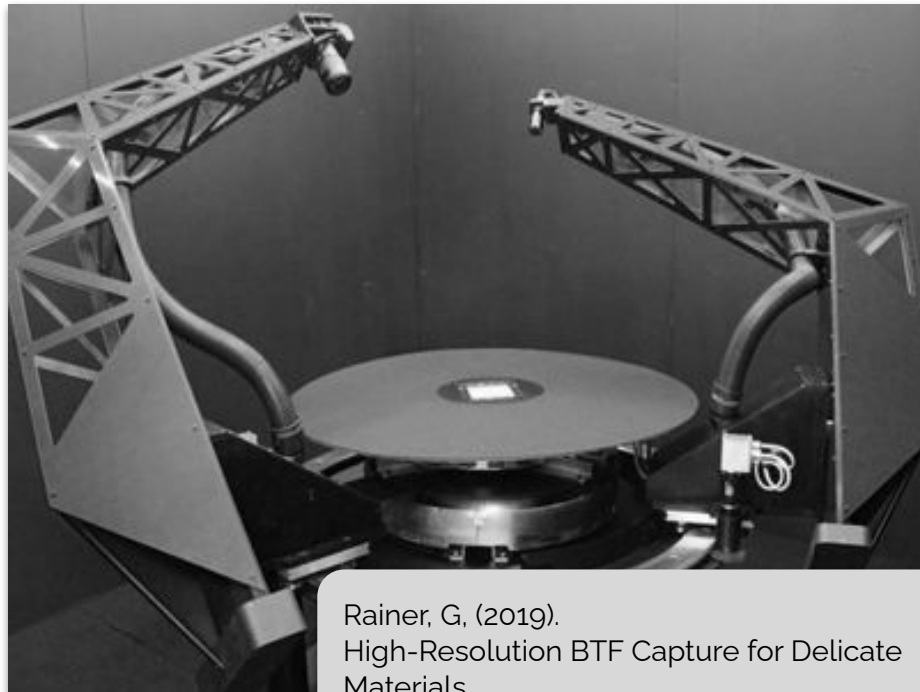
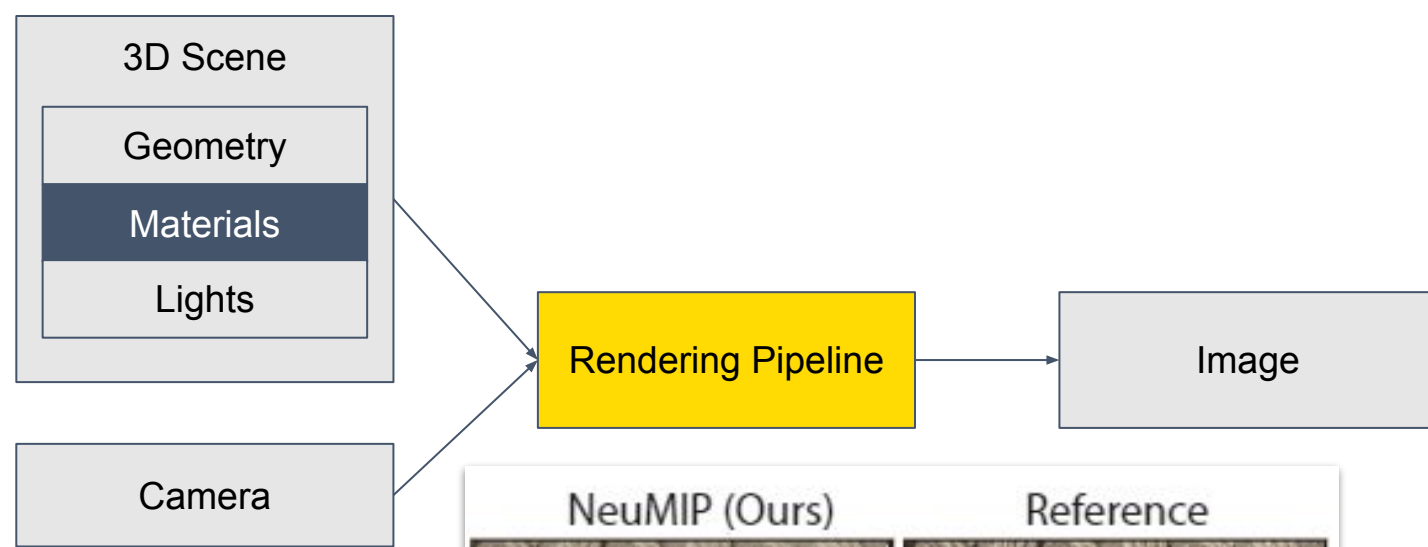
Hasselgren, J., (2021).
Appearance-driven automatic 3D model
simplification.

Martel, J. N. P., (2021).
ACORN: Adaptive Coordinate Networks for
Neural Scene Representation.

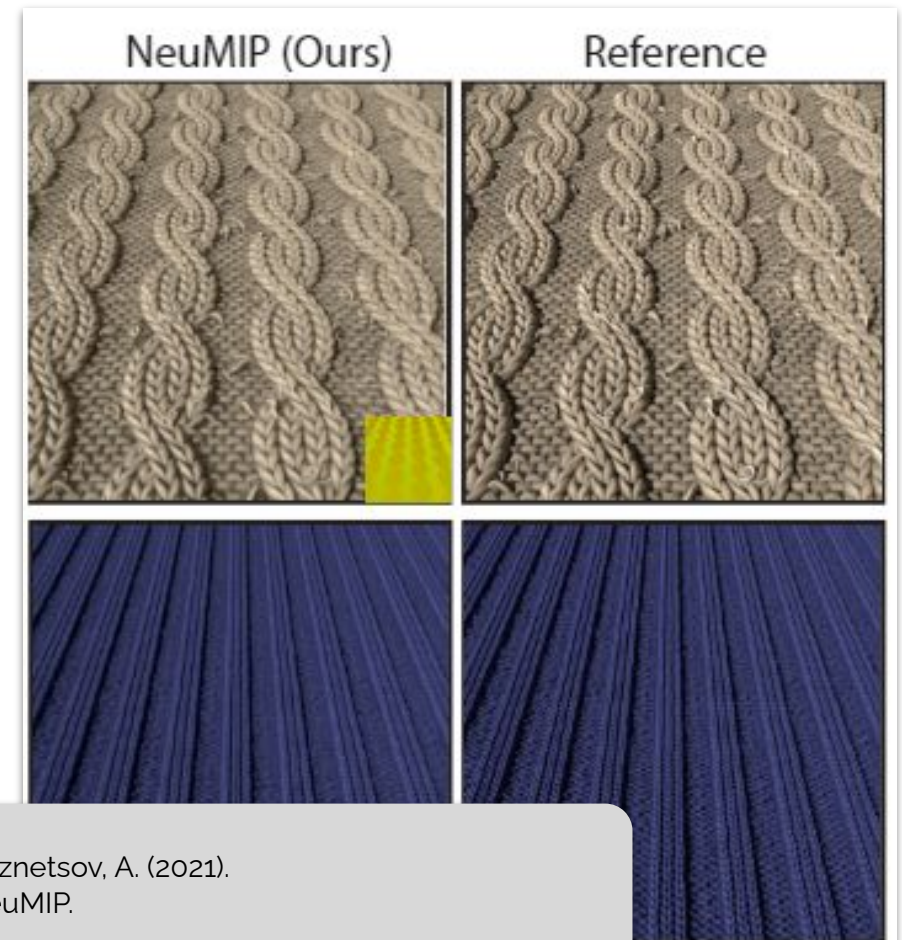
NEURAL MATERIALS

01 BTF: real-life captures

02 Neumip: Neural network material



Rainer, G. (2019).
High-Resolution BTF Capture for Delicate
Materials.



Kuznetsov, A. (2021).
NeuMIP.



PRODUCTION RENDERING

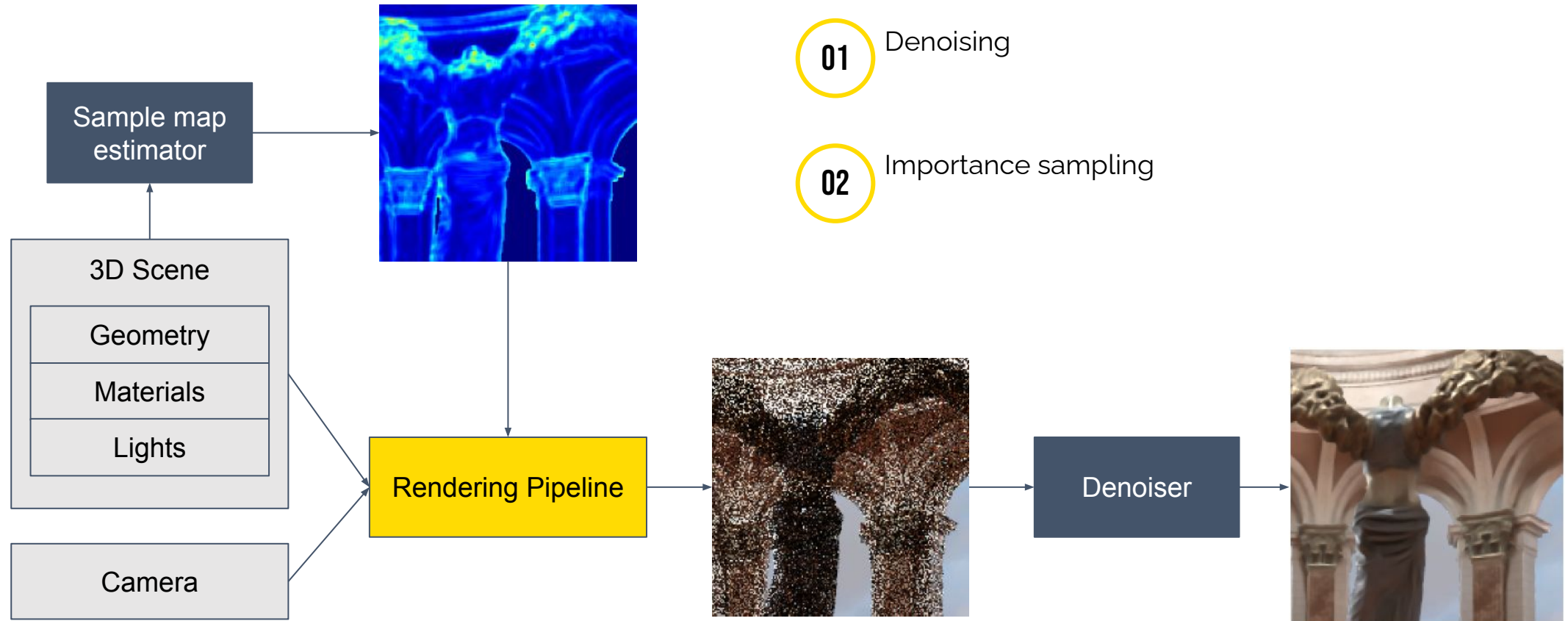


Noisy (16 spp)

Vogels, T., (2018).
Denoising with kernel prediction and
asymmetric loss functions.

Denoised

PRODUCTION RENDERING



01

Denoising

02

Importance sampling



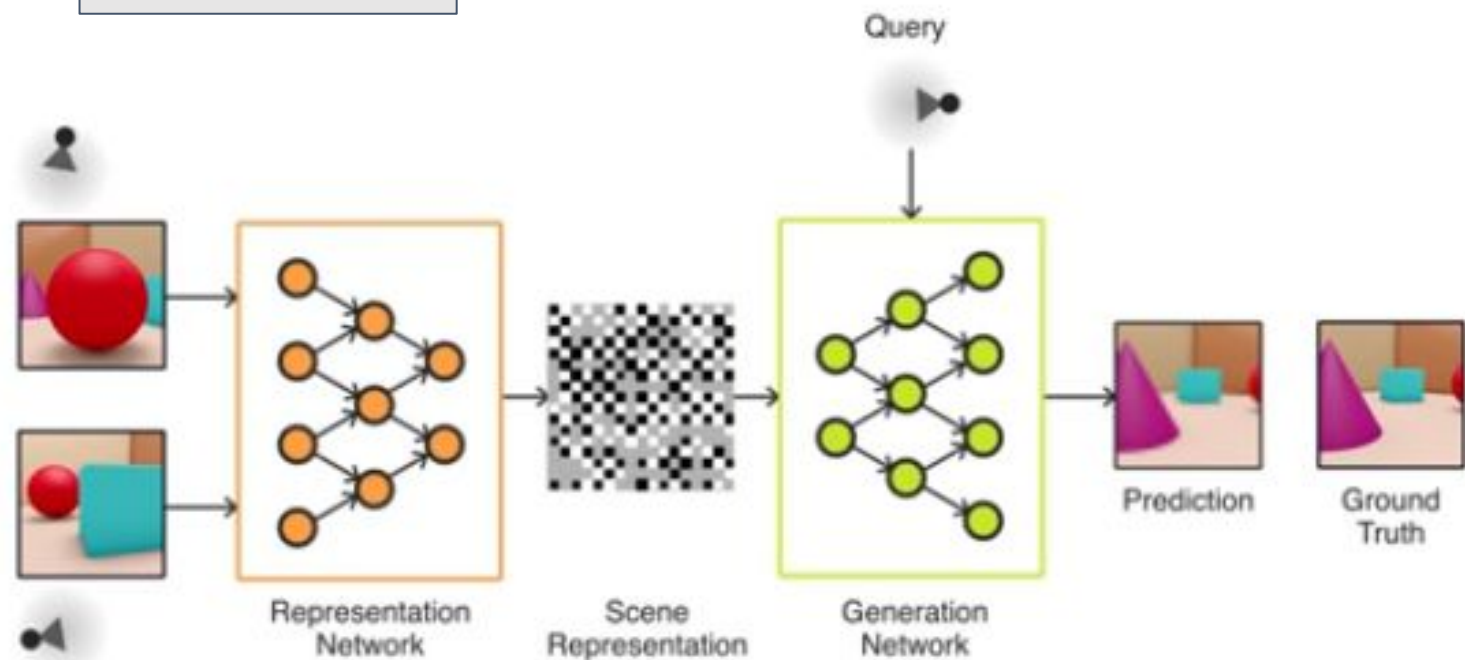
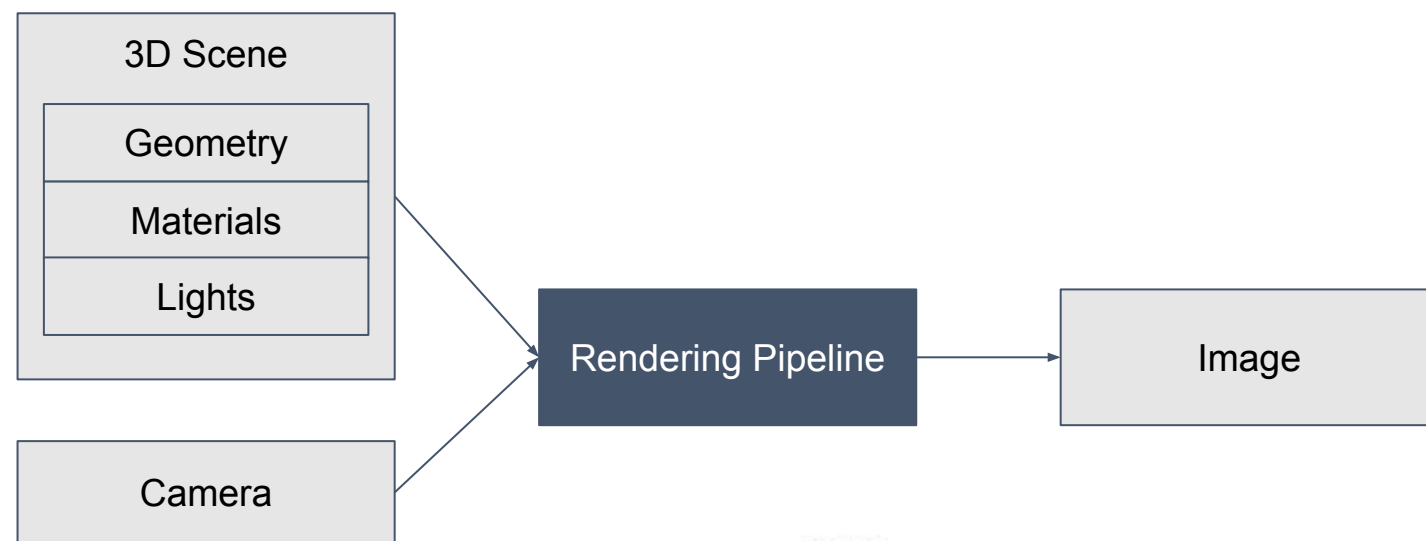
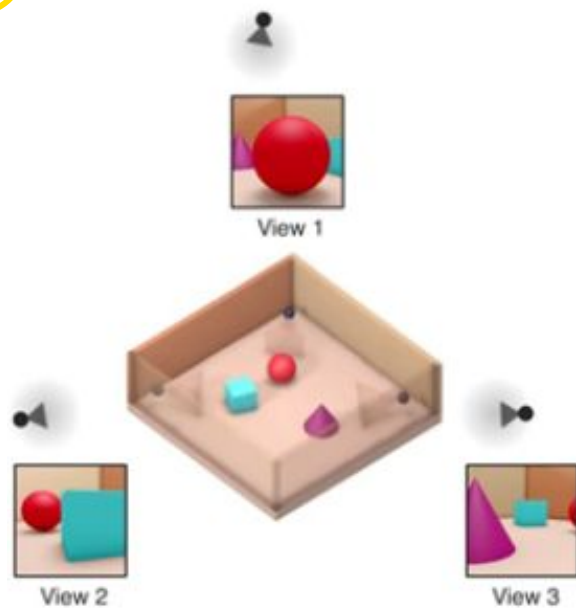
NEURAL RENDERING

01

Rendering pipeline replaced by neural network

02

Generative Query Networks
• from few observations



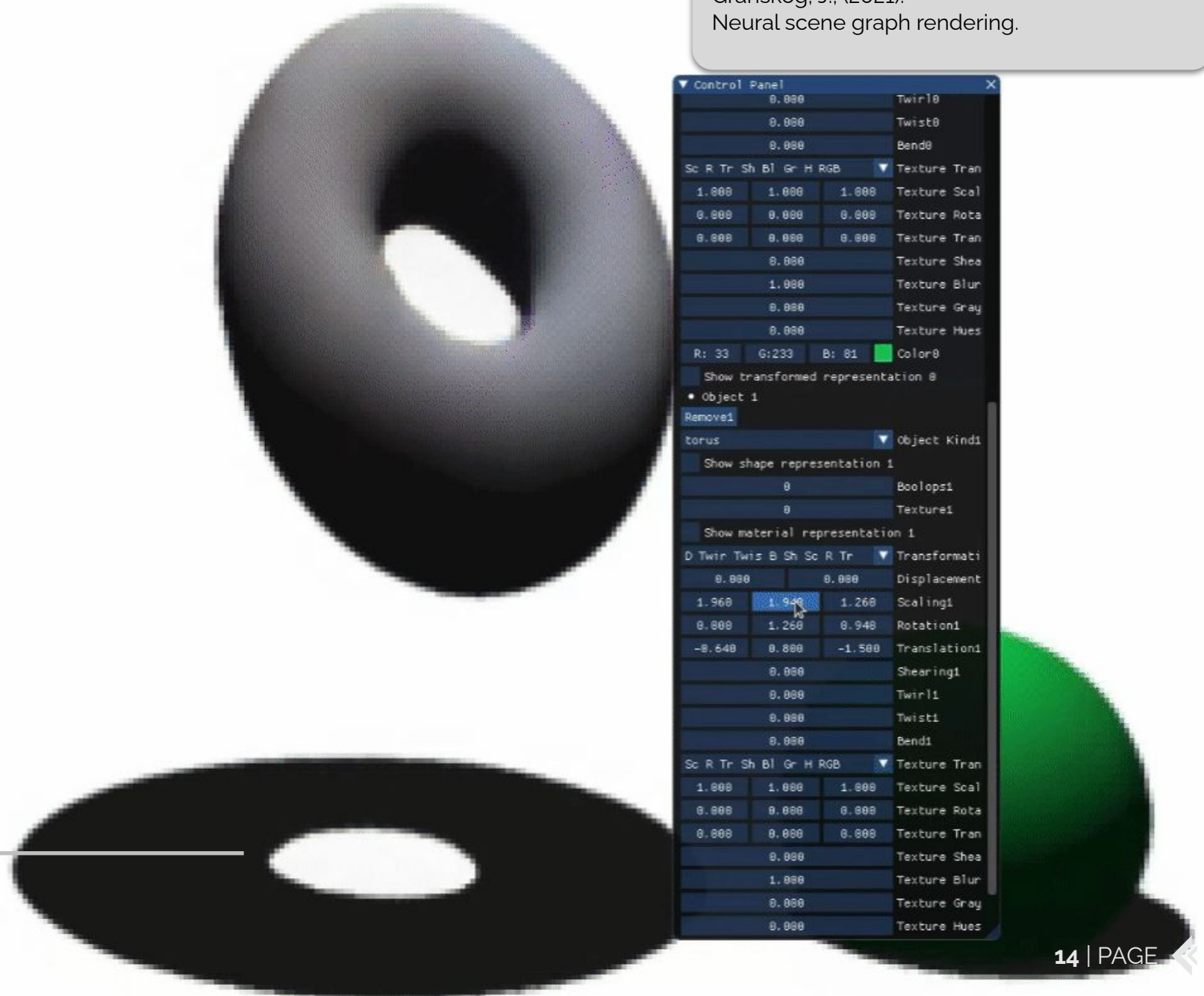
Eslami, S. M. A., (2018).
Neural scene representation and rendering.



NEURAL RENDERING

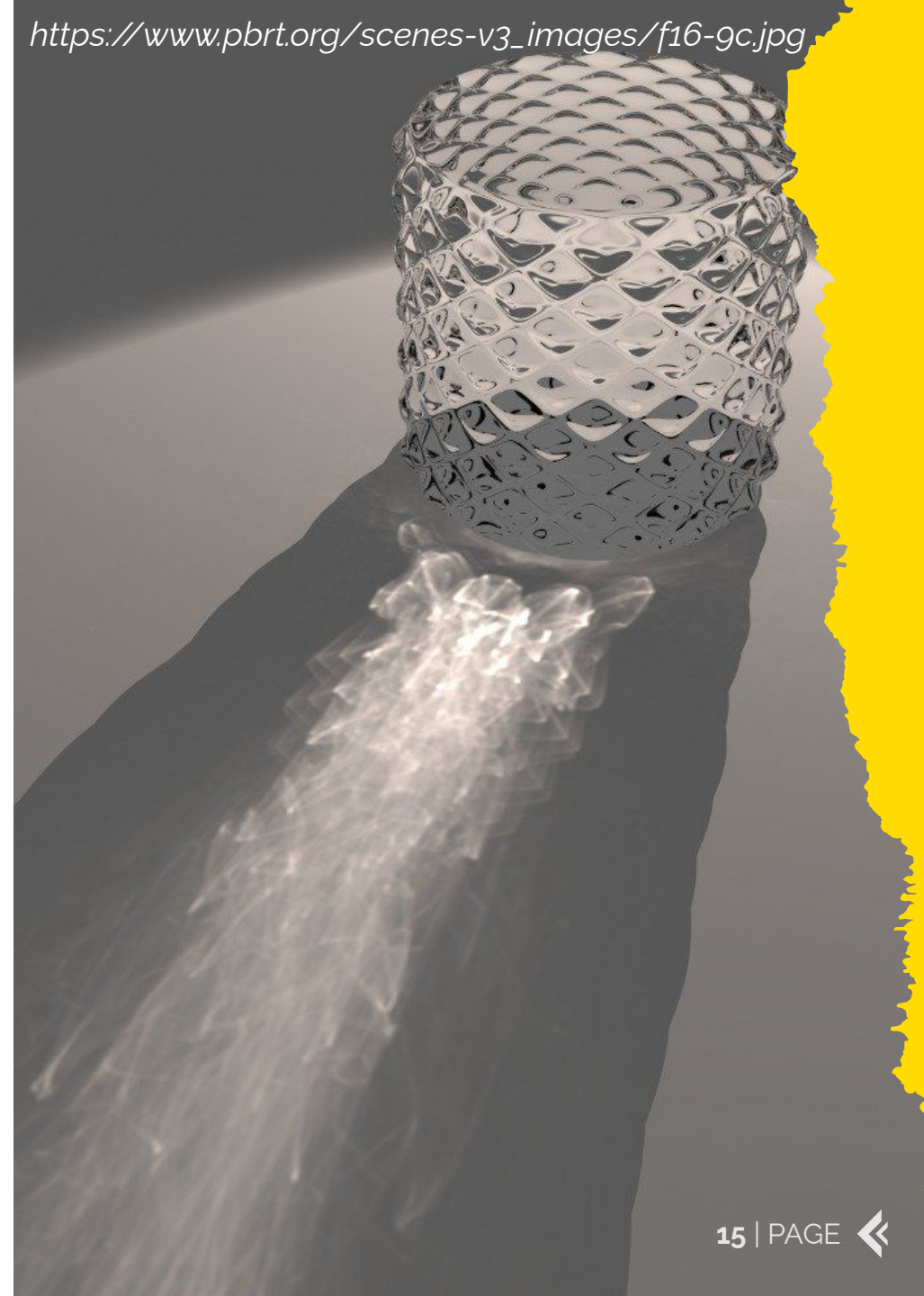
Granskog, J., (2021).
Neural scene graph rendering.

03. Neural scene graph rendering



RECAP

- 01 With Supersampling, we can render at lower resolution and achieve higher refresh rate
- 02 With Inverse Graphics we can create a 3D representation from few images/video
- 03 A Neural Network can represent 3D scenes
- 04 Speed up Production rendering by optimizing the amount of samples
- 05 Or simply replace everything with Neural networks





THANK YOU

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