



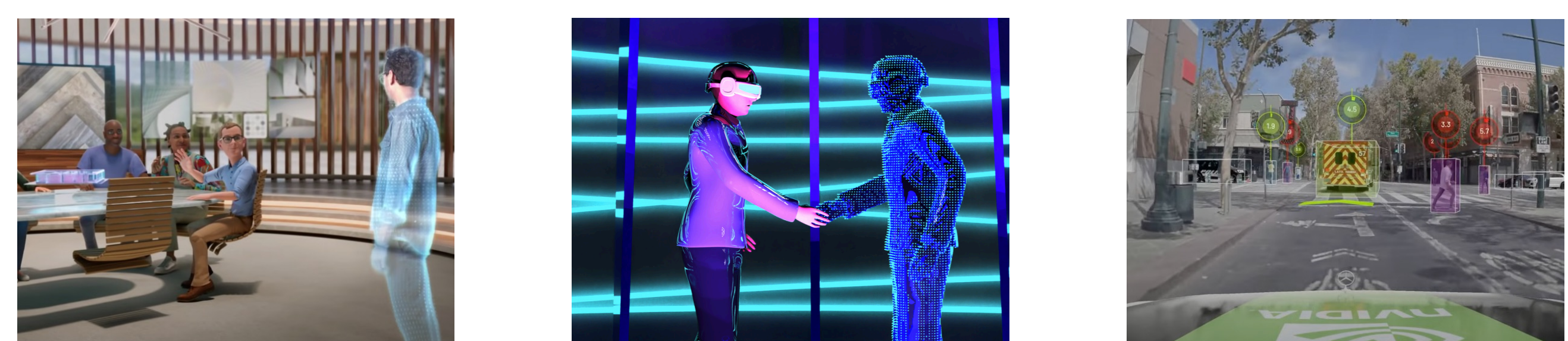
## NeRF: A Tool to Generate Novel Views

- Neural Radiance Fields (NeRF) can generate **arbitrary new views** of a specific scene **given sparsely sampled scene images**



**Inputs:** Sparsely sampled views **Outputs:** Images of any new view

## Real-Time NeRF Is Increasingly Demanded



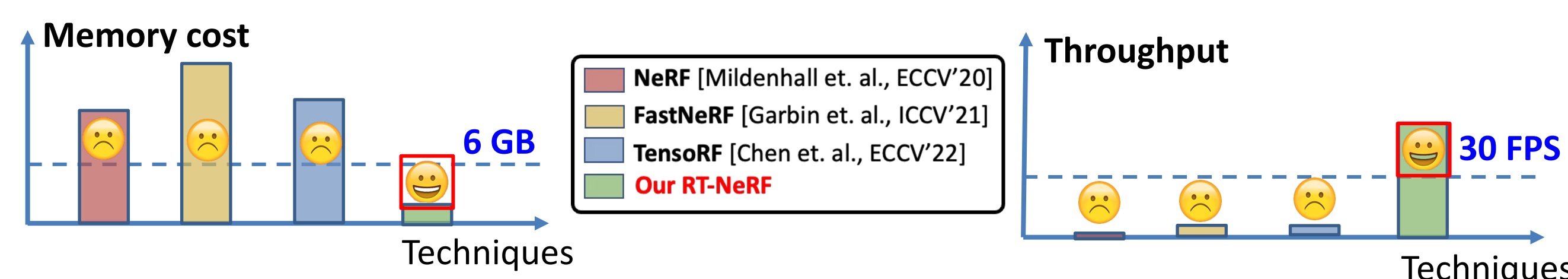
Virtual Meetings

Metaverse

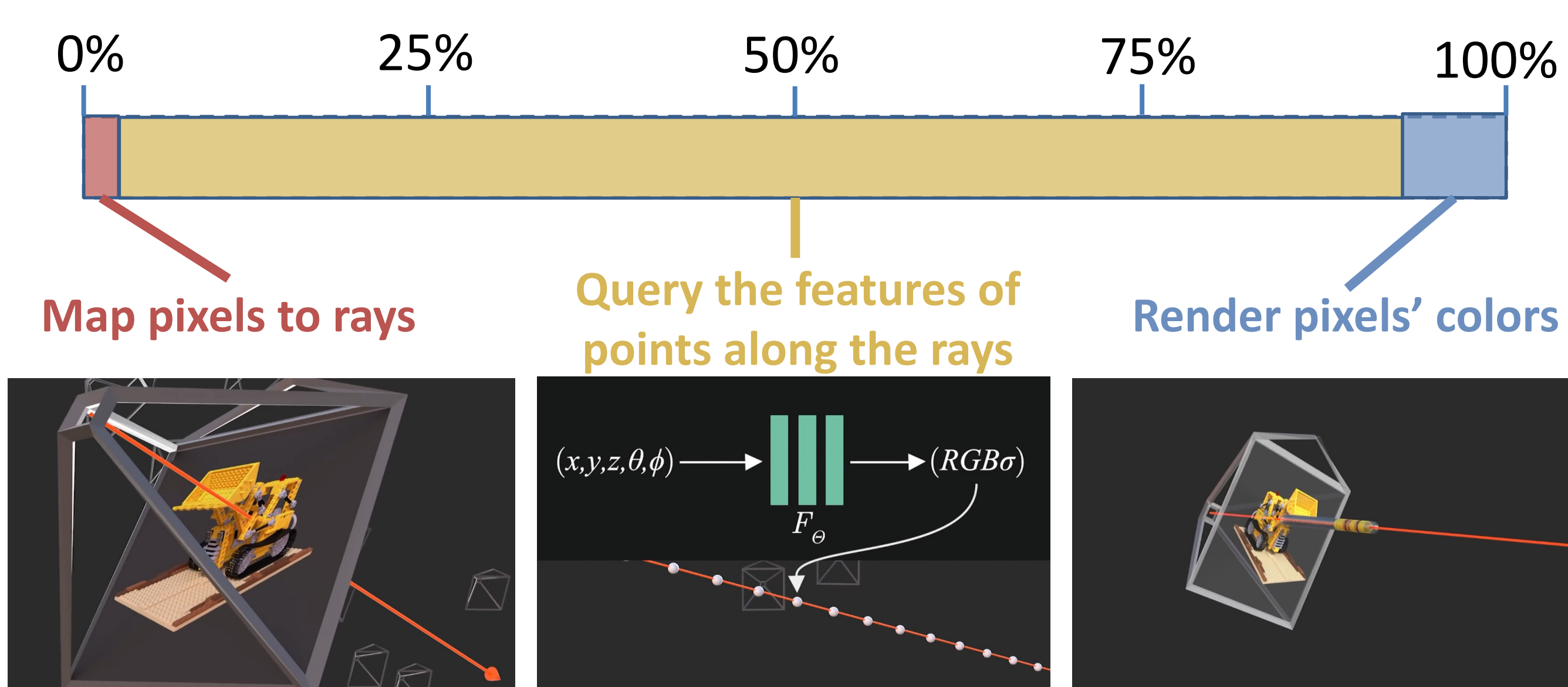
Autonomous Driving Simulation

## SOTA Efficient NeRF's Limitations

- Limitation 1:** Large memory requirement
- Limitation 2:** Low throughput



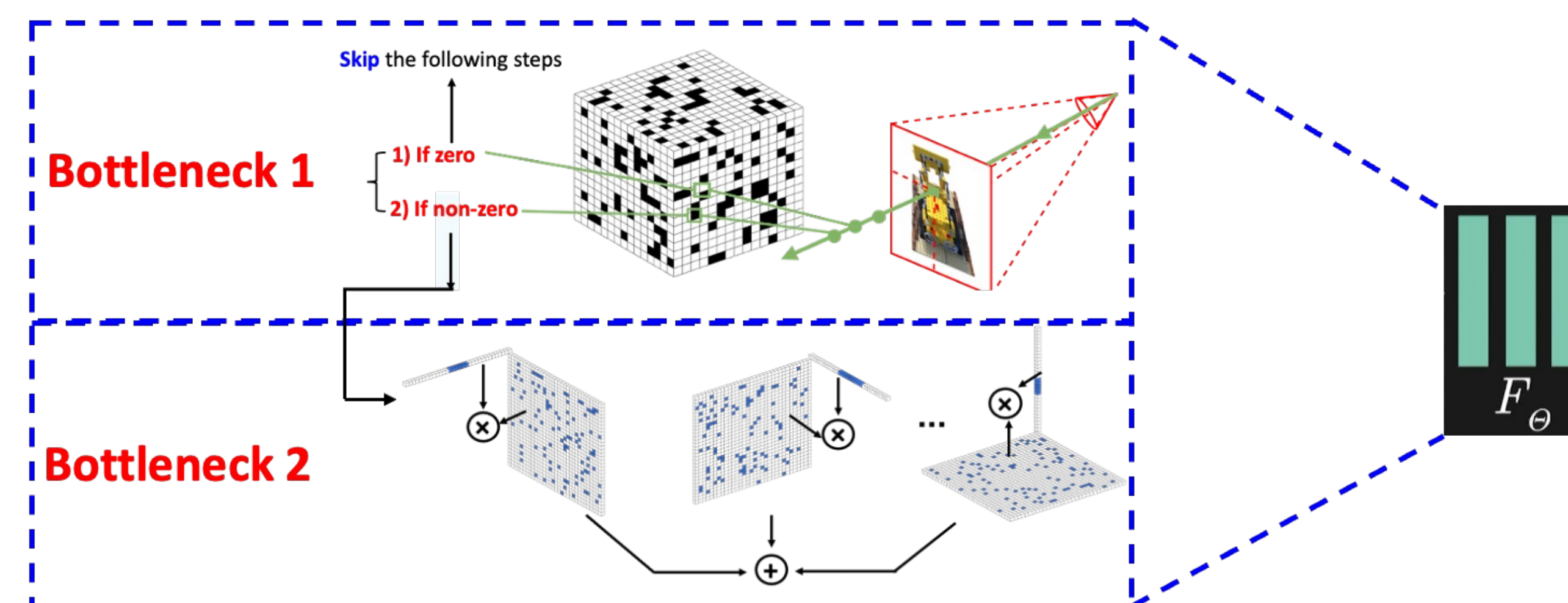
## Contribution 1: Analyze the Efficiency Bottleneck



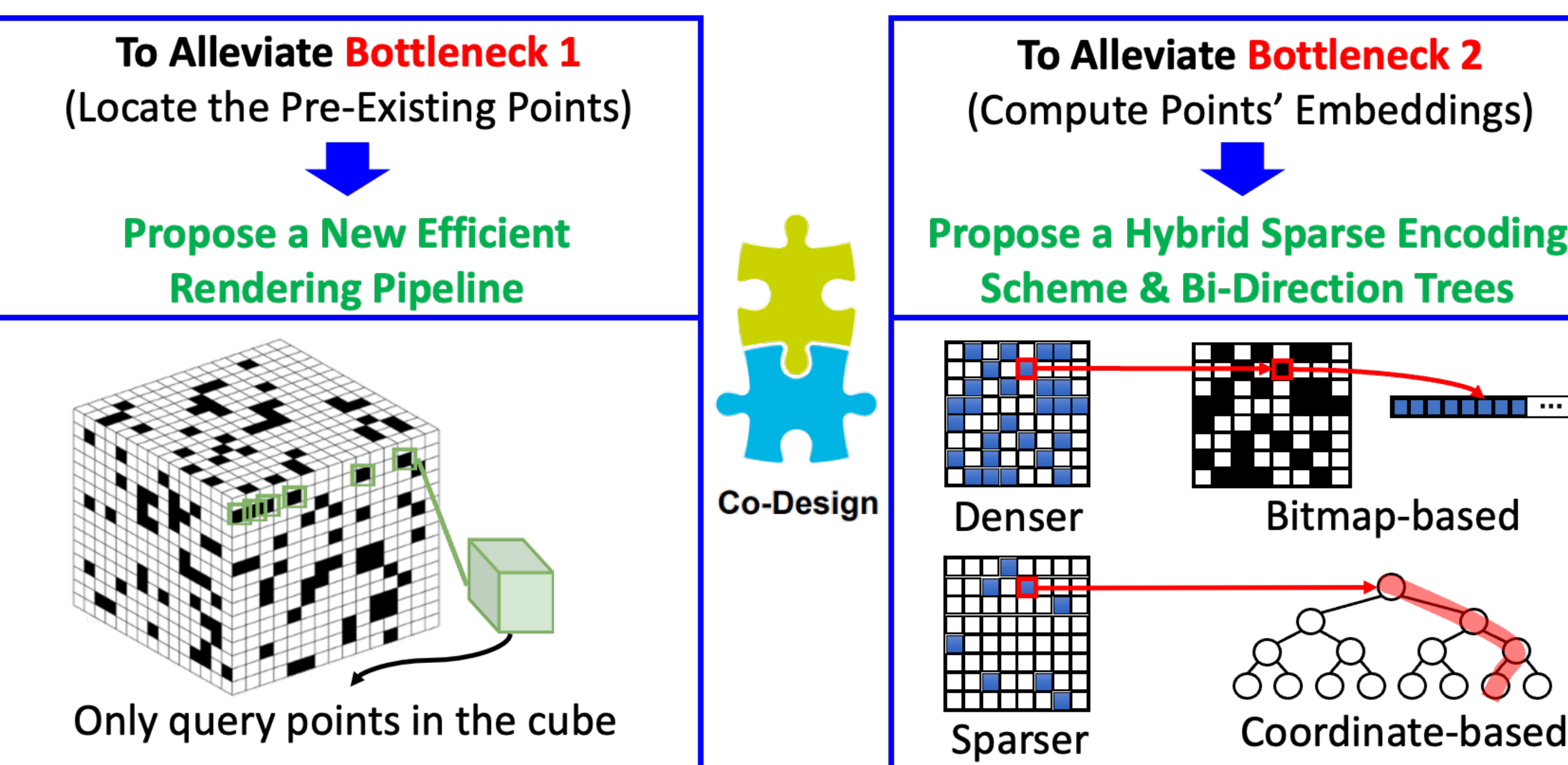
Source: [Mildenhall et. al., ECCV'20]

## Contribution 2: Identify Two Key Bottlenecks

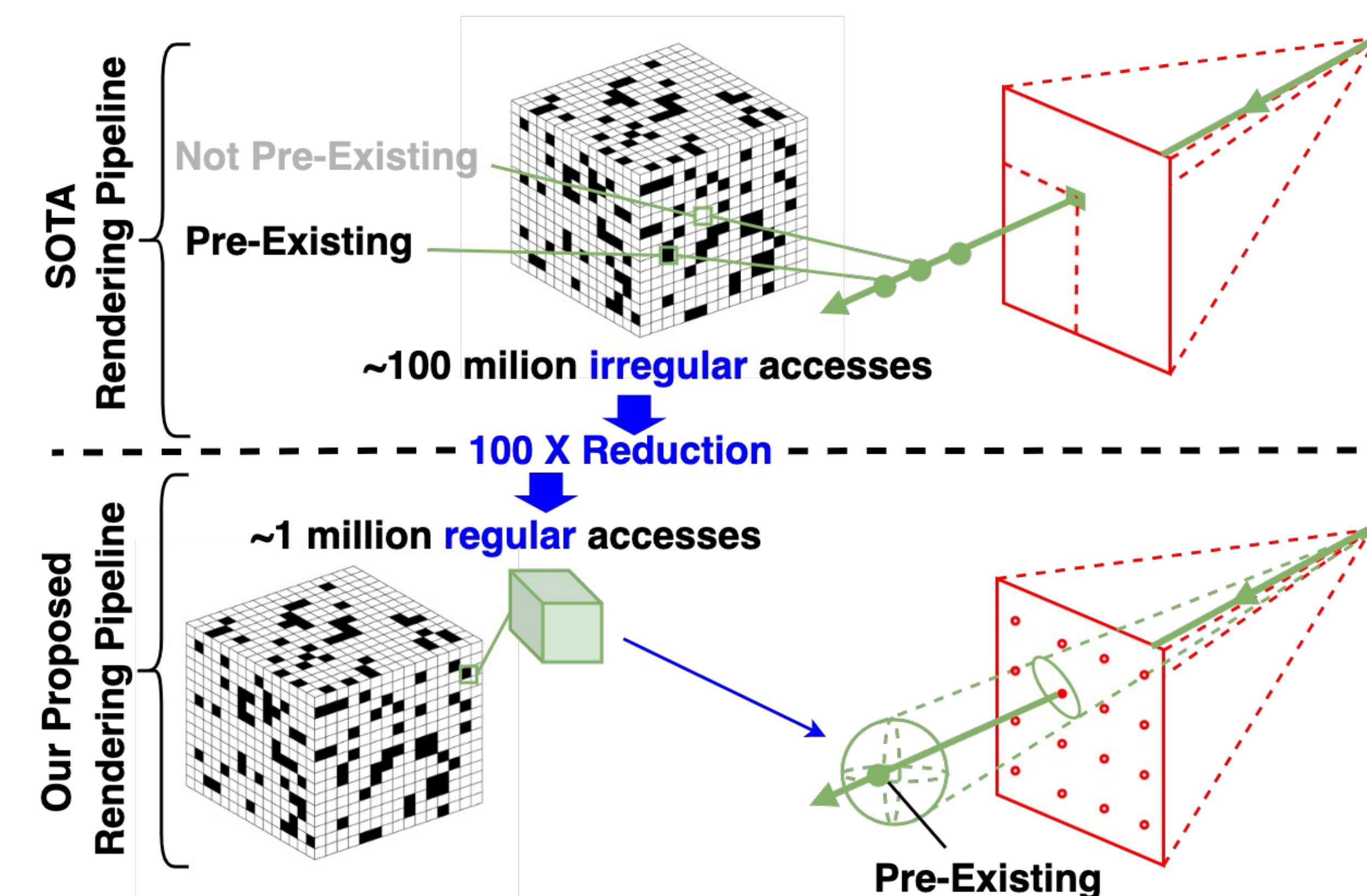
- Dominant step: Query the features of points along the rays
- Bottleneck 1** - Locate pre-existing points
- Bottleneck 2** - Compute points' embeddings



## Overview of the Proposed RT-NeRF



## Contribution 3: Efficient Rendering Pipeline



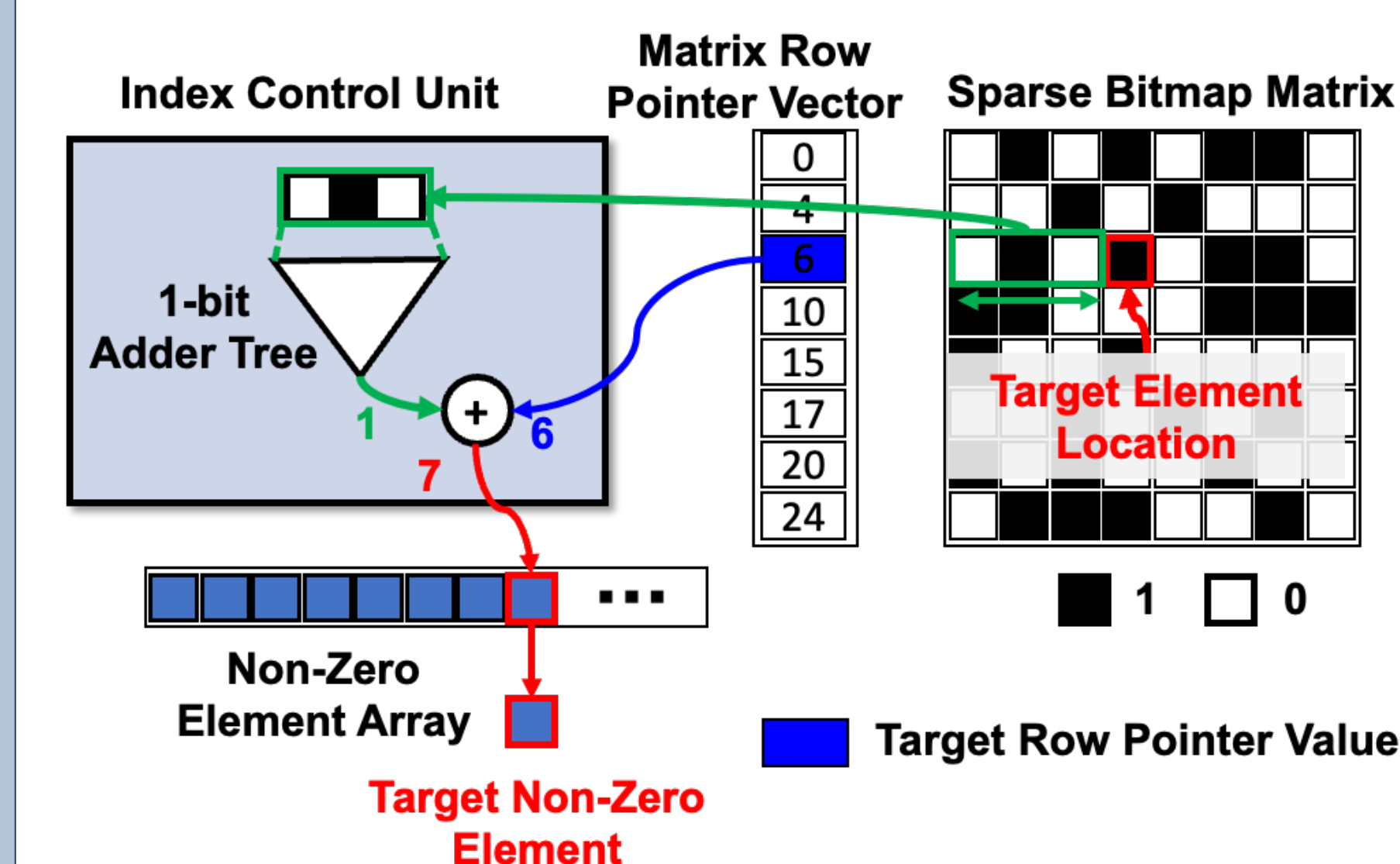
## Contribution 4: Hybrid Sparse Encoding

- For **dense** (< 80% sparsity) matrices
- For **sparse** ( $\geq 80\%$  sparsity) matrices

Encoding Scheme	Storage Size (↓)	Decoding Throughput (↑)	Resource Utilization (↑)
Bitmap-based	☆☆☆	☆☆	☆☆☆☆
Our proposed	☆☆☆☆	☆☆☆☆	☆☆☆☆

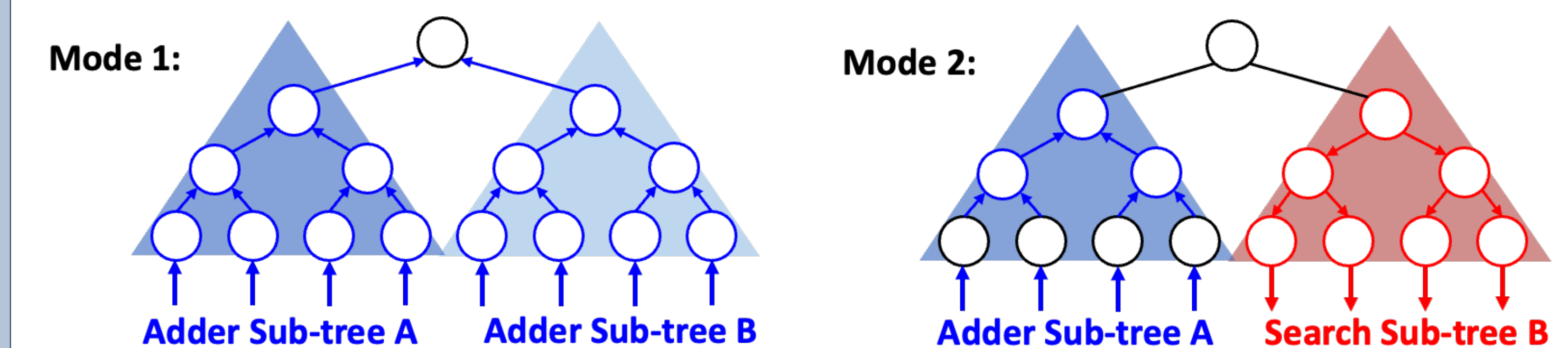
Encoding Scheme	Storage Size (↓)	Decoding Throughput (↑)	Resource Utilization (↑)
Coordinate-based	☆☆☆☆	☆☆☆☆	☆☆
Our proposed	☆☆☆☆	☆☆☆☆	☆☆☆☆

## Contribution 5: Improved Bitmap-Based Encoding Scheme to Boost Throughput

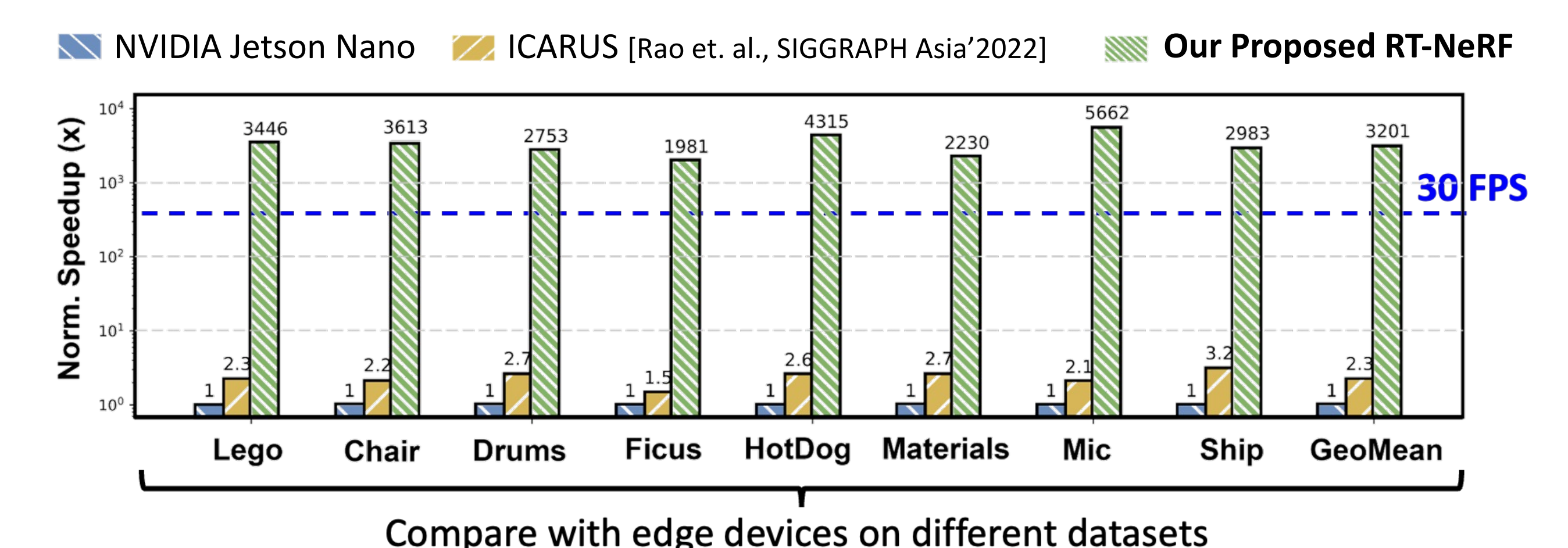


- Cycle 1:** Check the bitmap matrix element 1 or 0
- Cycle 2:** Sum up 1-bit bitmap vector and then add the row pointer value
- Cycle 3:** Fetch the target non-zero element

## Contribution 6: Bi-Direction Trees to Boost Utilization



## RT-NeRF's Speedup Over Baselines



Our RT-NeRF framework has delivered **the first real-time** neural rendering solution suited for edge applications