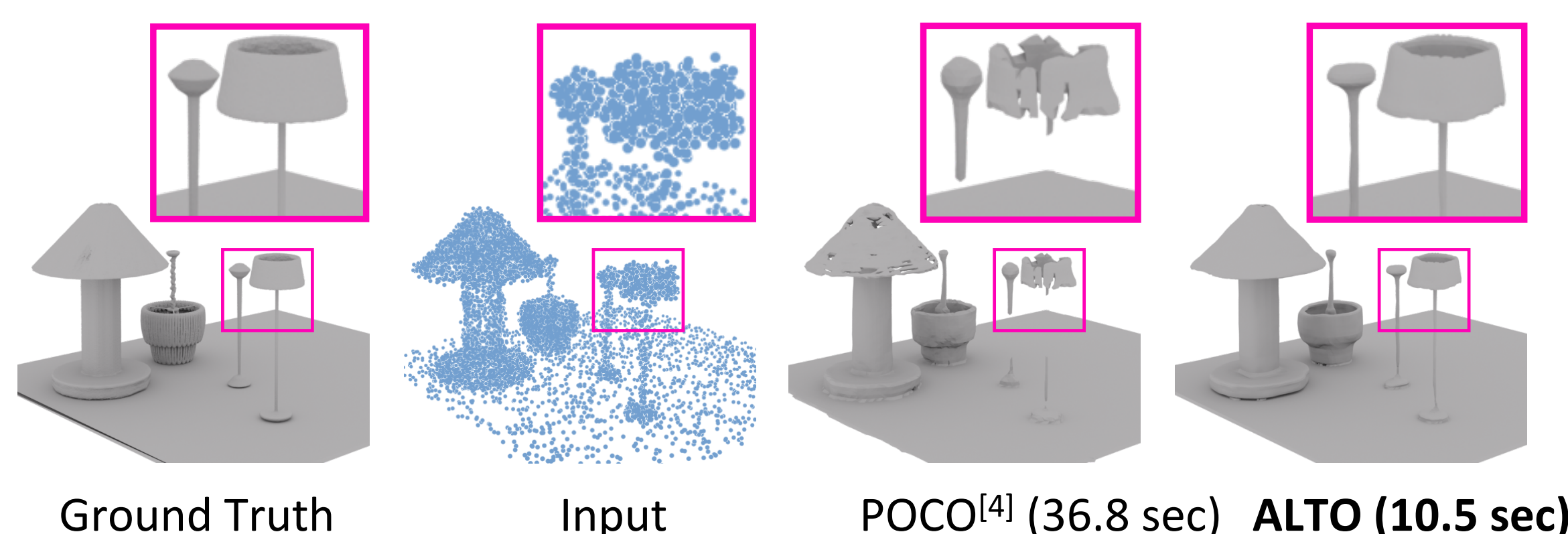


## Contributions

- We introduce an iterative technique to blend the strengths of different latent topologies for high-fidelity conditional neural fields generation.
- We propose an attention-based decoder that replaces naive linear interpolation of feature-grids or computationally expensive point-wise attention while keeping compute burden in check.
- We demonstrate performance and runtime improvements over the highest-quality previous method, as well as performance improvements over all other baselines.

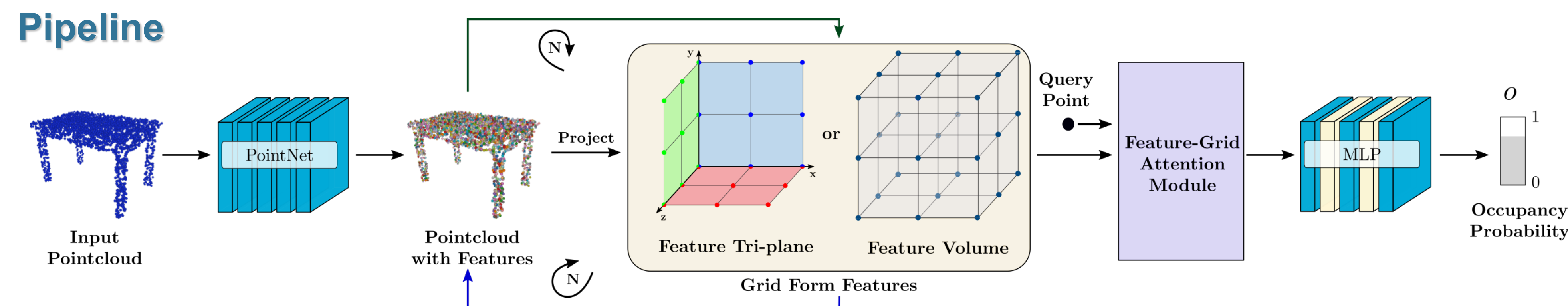


## References

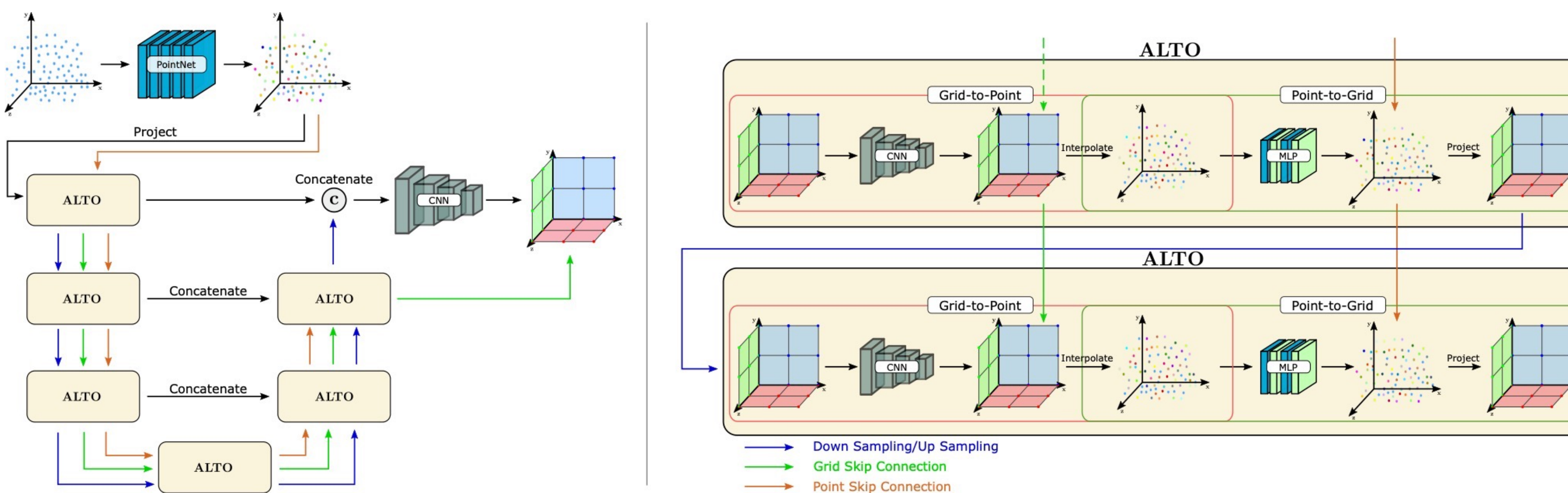
- [1] Chang, Angel X., et al. "Shapenet: An information-rich 3d model repository." arXiv preprint arXiv:1512.03012 (2015).  
 [2] Peng, Songyou, et al. "Convolutional occupancy networks." In ECCV, pp. 523-540, 2020.  
 [3] Dai, Angela, et al. "Scannet: Richly-annotated 3d reconstructions of indoor scenes." In CVPR, pp. 5828-5839, 2017.  
 [4] Boulch, Alexandre, and Renaud Marlet. "Poco: Point convolution for surface reconstruction." In CVPR, pp. 6302-6314, 2022.  
 [5] Kazhdan, Michael, and Hugues Hoppe. "Screened poisson surface reconstruction." ACM Transactions on Graphics (ToG) 32.3 (2013): 1-13.

## Method

### Pipeline

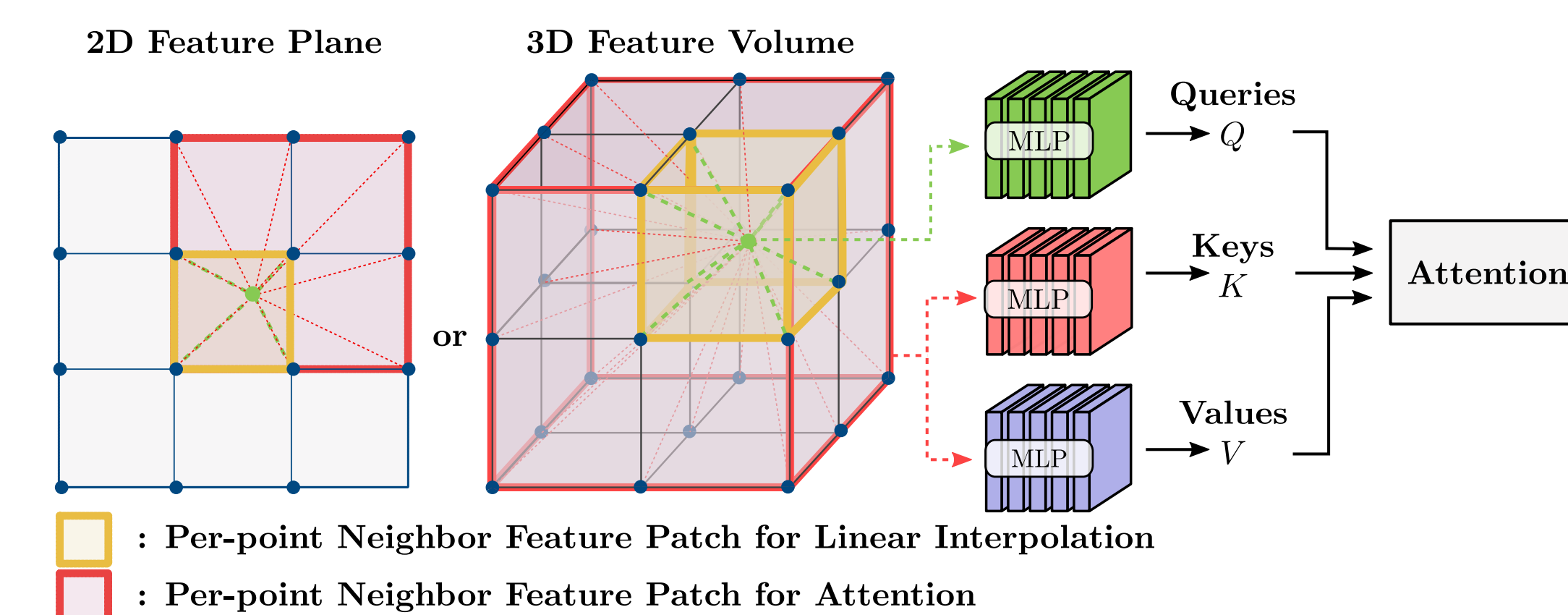


- Given input surface points, we obtain an implicit occupancy field with **iterative alternation between features in the forms of points and 2D or 3D grids**. Then we decode the occupancy values for query points with a learned attention-based interpolation from neighboring grids.



- As an example, we show the ALTO block instantiated by alternating between two latent topologies: point and triplanes via an "in-network" fashion, i.e. within each level of an hourglass framework U-Net.

### Feature-Grid Attention Module



- Attention-based decoder on neighboring grids (2D or 3D)



<https://visual.ee.ucla.edu/alto.htm/>

## Experimental Results

