

# IPFormer: Visual 3D Panoptic Scene Completion with Context-Adaptive Instance Proposals



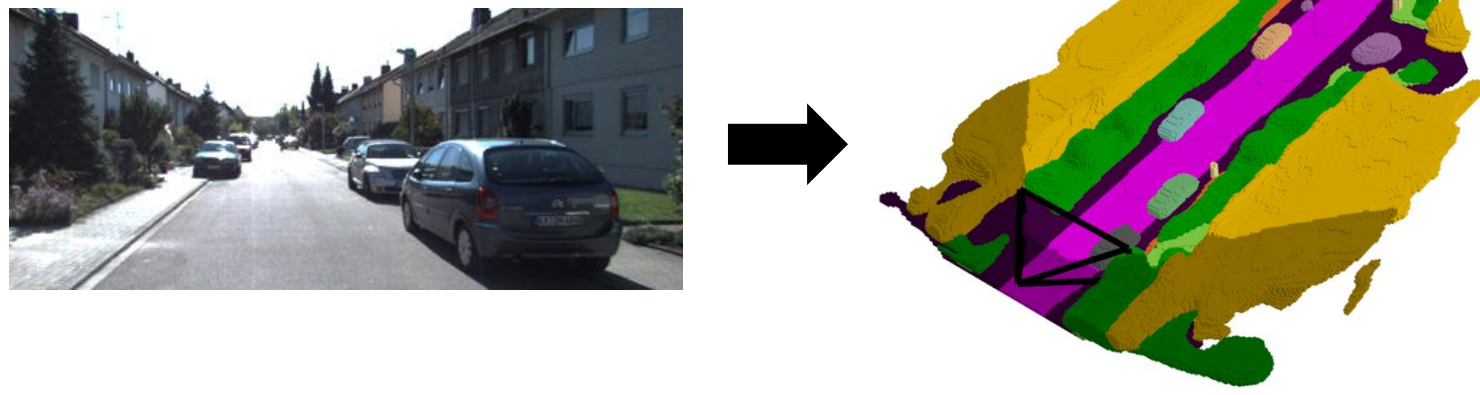
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## Task Description

Using camera images, infer the complete 3D structure of a scene as a voxel grid, including both visible and occluded regions. Every voxel carries

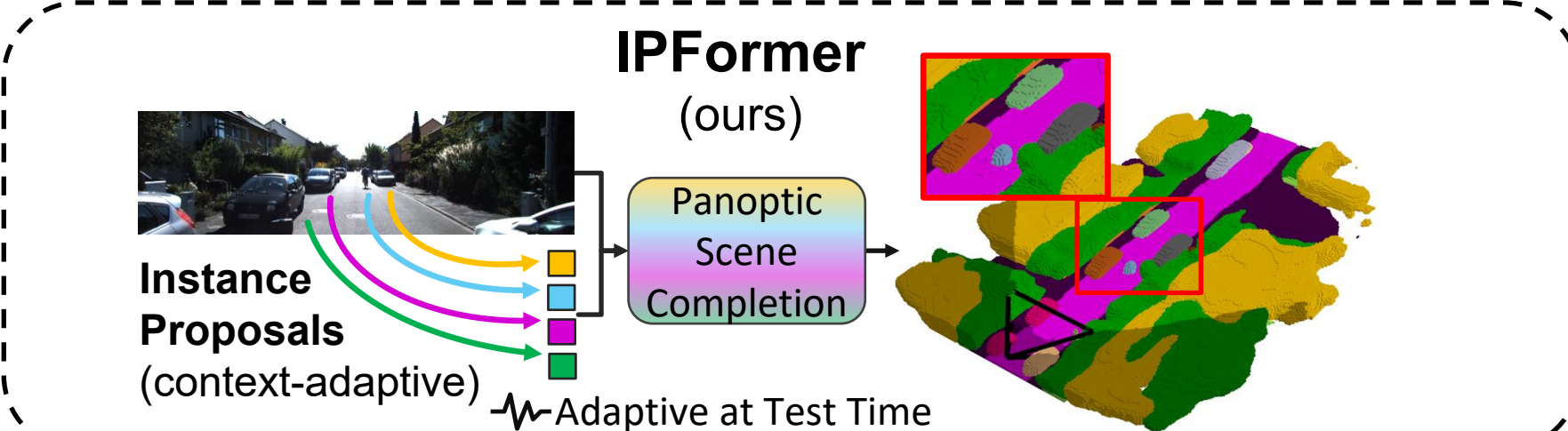
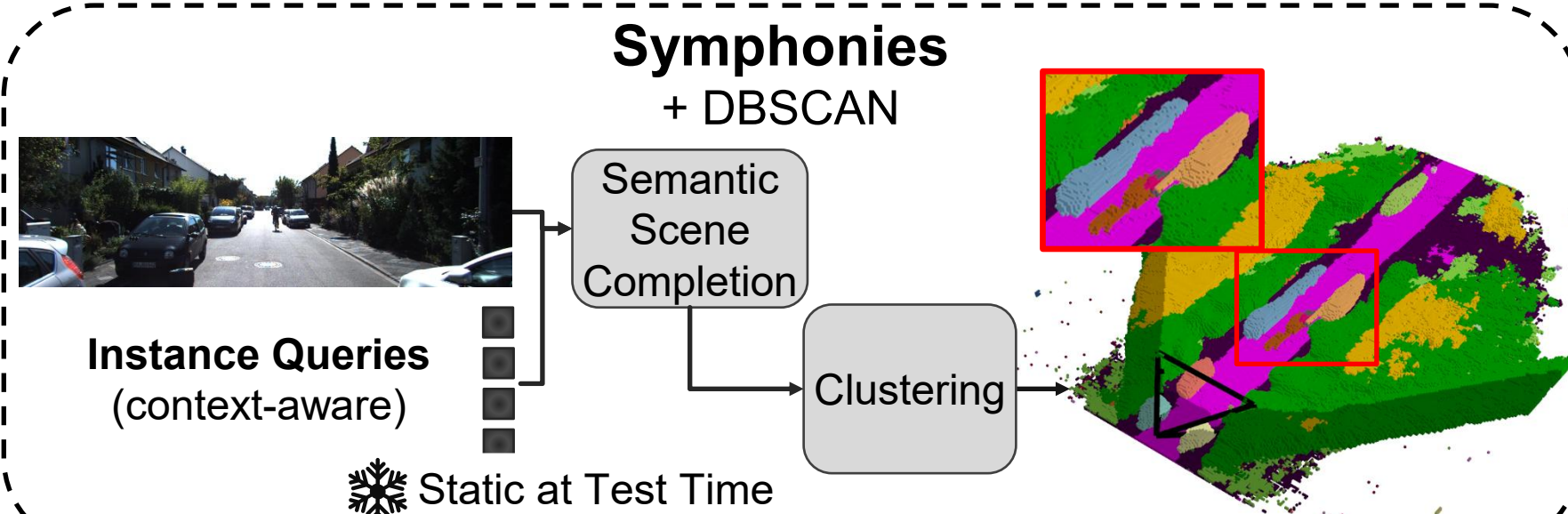
1. binary **occupancy**
2. a **semantic** label
3. an **instance** ID to group countable objects



## Challenges

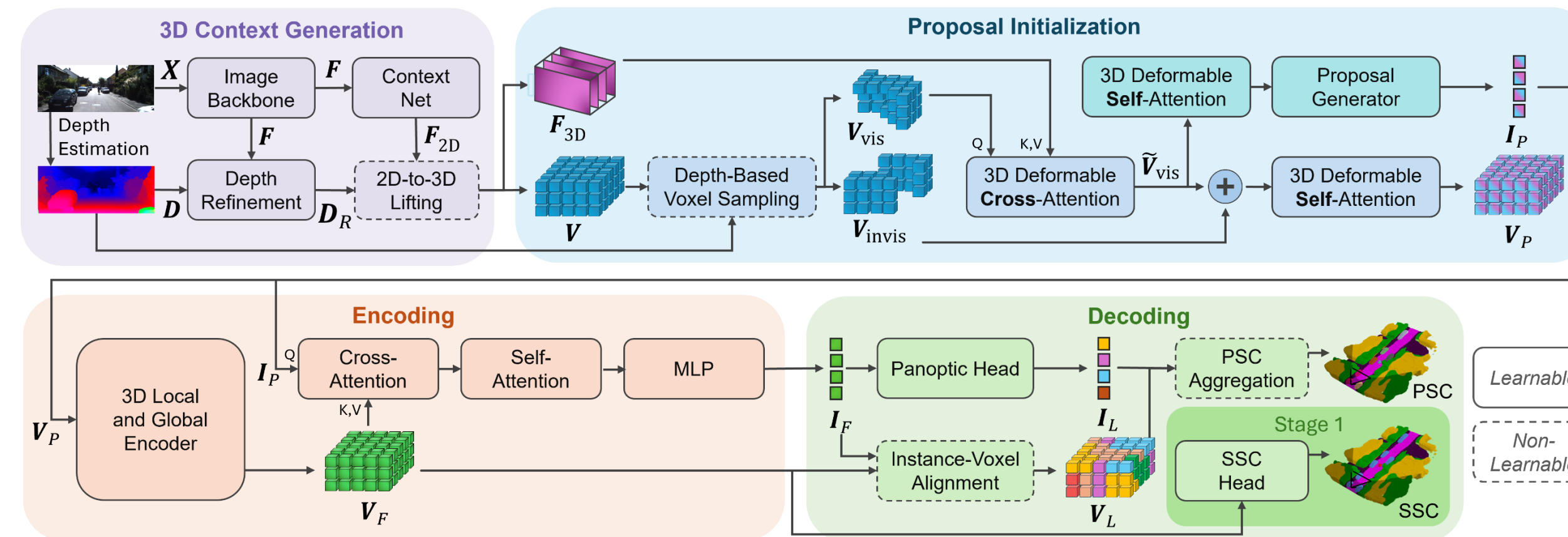
Previous methods

1. only infer **occupancy** and **semantics** in an end-to-end fashion (Semantic Scene Completion), but require subsequent, time-consuming Euclidean clustering to retrieve individual **instances**.
2. reconstruct objects using a fixed set of learned queries that are updated with image context during training, but remain static at test time and thus fail to dynamically adapt specifically to the observed scene.



## Our Approach

Our method (1) addresses Panoptic Scene Completion in an end-to-end fashion, and (2) initializes object queries as instance proposals that dynamically adapt specifically to the observed scene at train and test time.



Specifically, we

1. propose a dual-head architecture and a two-stage training scheme that effectively guides the latent space toward **occupancy** and **semantics** before **instance** registration.
2. introduce a visibility-based sampling strategy, which utilizes visible voxels and respective image context to adaptively initialize instance proposals.

## Quantitative Results

In-Domain Performance:

Method	All				PSC Metrics				SSC Metrics			
	PQ <sup>↑</sup>	PQ <sup>↑</sup>	SQ <sup>↑</sup>	RQ <sup>↑</sup>	PQ <sup>↑</sup>	Thing SQ <sup>↑</sup>	RQ <sup>↑</sup>	PQ <sup>↑</sup>	Stuff SQ <sup>↑</sup>	RQ <sup>↑</sup>	IoU <sup>↑</sup>	mIoU <sup>↑</sup>
MonoScene [4] + DBSCAN	10.12	3.43	15.15	5.33	0.51	7.36	0.87	5.56	20.81	8.57	36.80	11.31
Symphonies [21] + DBSCAN	11.69	3.75	26.09	5.95	1.07	27.65	1.76	5.70	24.95	8.99	41.92	15.02
OccFormer [63] + DBSCAN	11.25	4.32	24.19	6.69	0.68	21.47	1.15	6.96	26.16	10.73	36.43	13.51
CGFormer [59] + DBSCAN	14.39	6.16	48.14	9.48	2.20	44.46	3.47	9.03	50.82	13.86	45.98	16.89
IPFormer (ours)	14.45	6.30	41.95	9.75	2.09	42.67	3.33	9.35	41.43	14.43	40.90	15.33

Out-of-Domain Zero-Shot Generalization Performance:

	All				PSC Metrics				SSC Metrics			
	PQ <sup>↑</sup>	PQ <sup>↑</sup>	SQ <sup>↑</sup>	RQ <sup>↑</sup>	PQ <sup>↑</sup>	Thing SQ <sup>↑</sup>	RQ <sup>↑</sup>	PQ <sup>↑</sup>	Stuff SQ <sup>↑</sup>	RQ <sup>↑</sup>	IoU <sup>↑</sup>	mIoU <sup>↑</sup>
<b>SemanticKITTI</b>												
CGFormer [59] + DBSCAN	14.39	6.16	48.14	9.48	2.20	44.46	3.47	9.03	50.82	13.86	45.98	16.89
IPFormer (ours)	14.45	6.30	41.95	9.75	2.09	42.67	3.33	9.35	41.43	14.43	40.90	15.33
<b>KITTI-360</b>												
CGFormer [59] + DBSCAN	8.44	1.08	17.82	1.87	0.53	20.06	0.96	1.48	16.19	2.54	28.11	9.44
IPFormer (ours)	9.41	1.23	24.68	2.16	0.52	22.76	0.95	1.68	25.89	2.93	28.74	9.53
<b>Relative Gap ↓</b>												
CGFormer [59] + DBSCAN	41.37%	82.47%	62.98%	80.28%	75.91%	54.89%	72.34%	83.61%	68.15%	81.67%	38.88%	44.09%
IPFormer (ours)	34.88%	80.48%	41.19%	77.85%	75.12%	46.64%	71.53%	82.03%	37.52%	79.69%	29.73%	37.81%

## Qualitative Results

