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

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Task Description

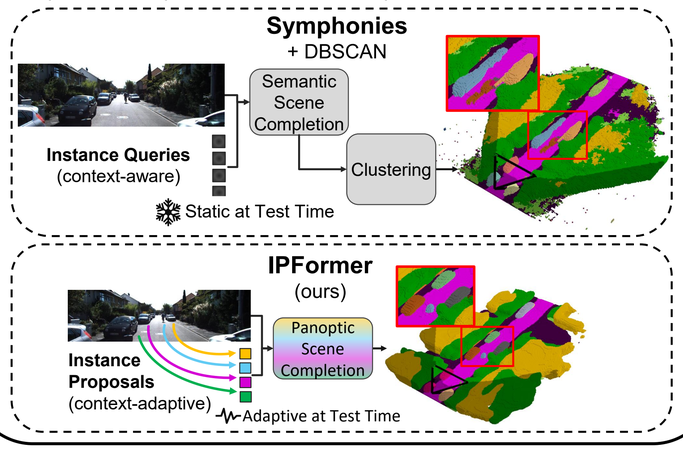
From a single image, infer the complete 3D structure of a scene as a voxel grid, including both visible and occluded regions. Every voxel carries

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

Challenges

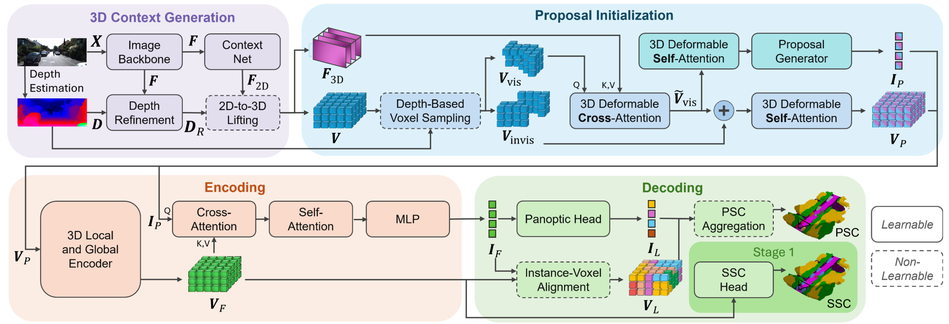
Previous methods

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	PSC Metrics: All PQ↑	All PQ↑	All SQ↑	All RQ↑	Thing PQ↑	Thing SQ↑	Thing RQ↑	Stuff PQ↑	Stuff SQ↑	Stuff RQ↑	SSC Metrics: IoU↑	mIoU↑	Runtime [s]↓
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	4.51
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	4.54
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	4.70
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**	4.70
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	**0.33**

Out-of-Domain Zero-Shot Generalization Performance:

	PSC Metrics: All PQ↑	All PQ↑	All SQ↑	All RQ↑	Thing PQ↑	Thing SQ↑	Thing RQ↑	Stuff PQ↑	Stuff SQ↑	Stuff RQ↑	SSC Metrics: IoU↑	mIoU↑
SemanticKITTI												
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
KITTI-360												
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**
Relative Gap ↓												
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