

Scene Graph Prediction with Limited Labels

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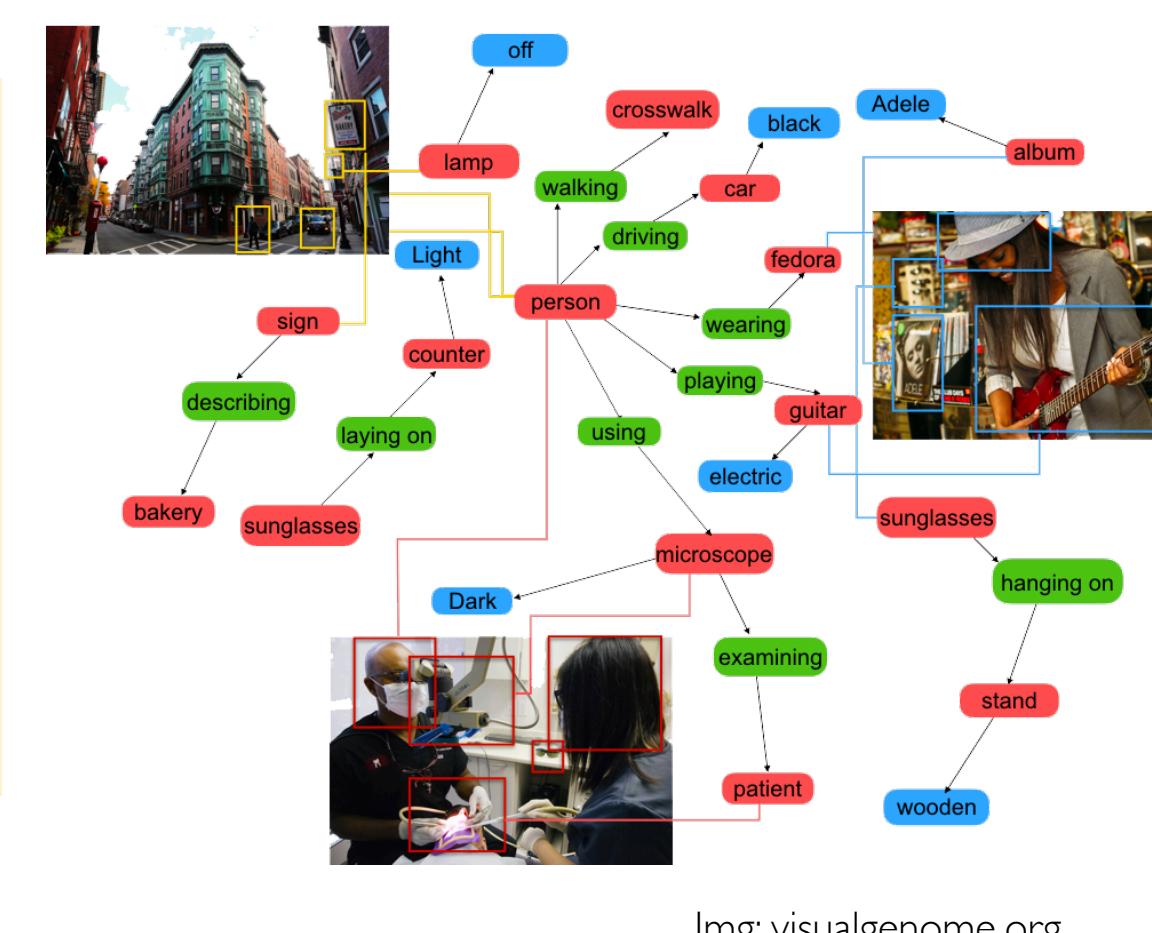
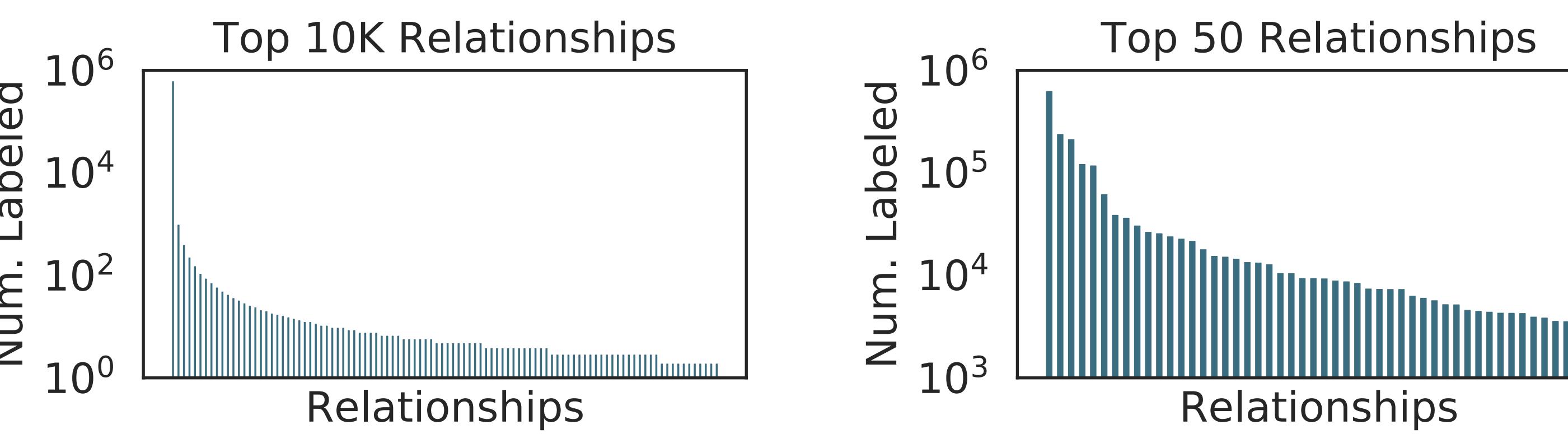


The Long Tail of Visual Relationships

Problem

Scene graph datasets are **incomplete** due to annotator error + cost.

State-of-the-art models only consider the top 50 relationships and ignore the rest!

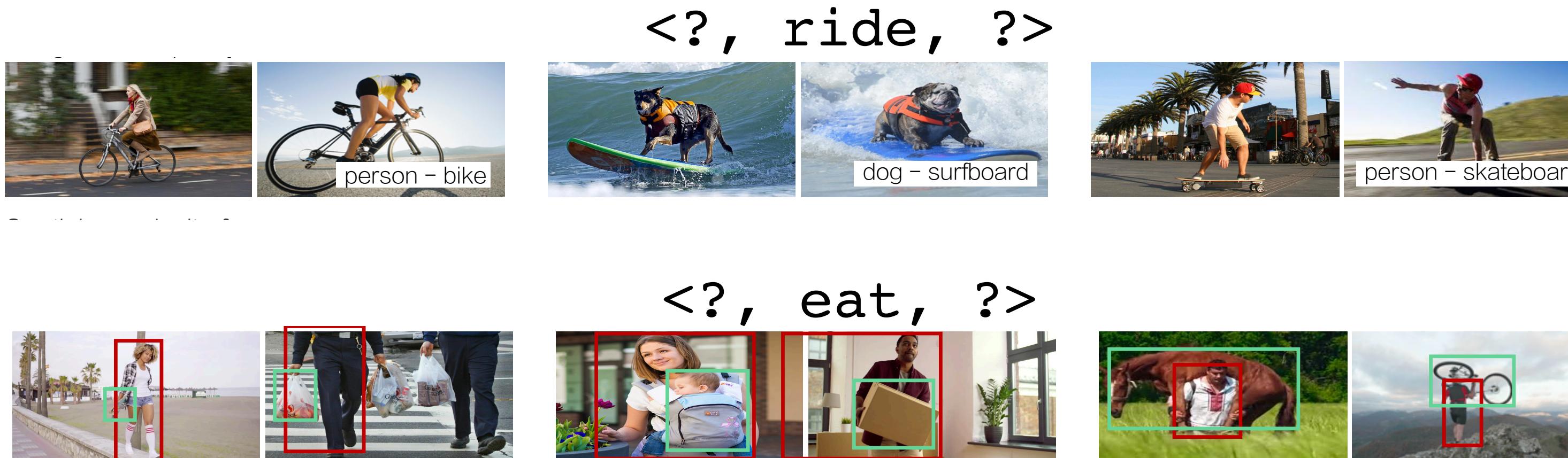


Leveraging Image-Agnostic Features

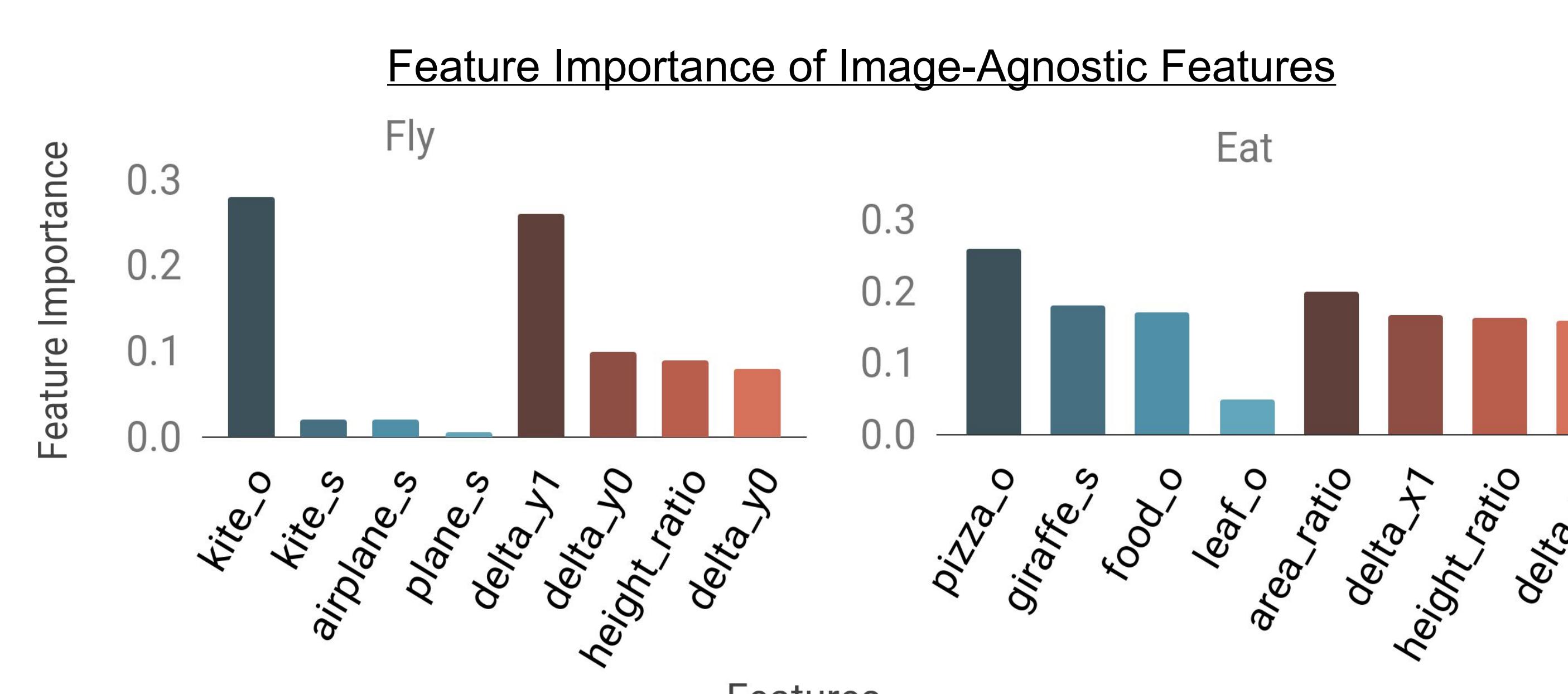
Inspiration: *Textual relation extraction relies on document-agnostic heuristic*

<Tokyo, capitol of, Japan>

Challenge: Visual relationships vary based on context within an image

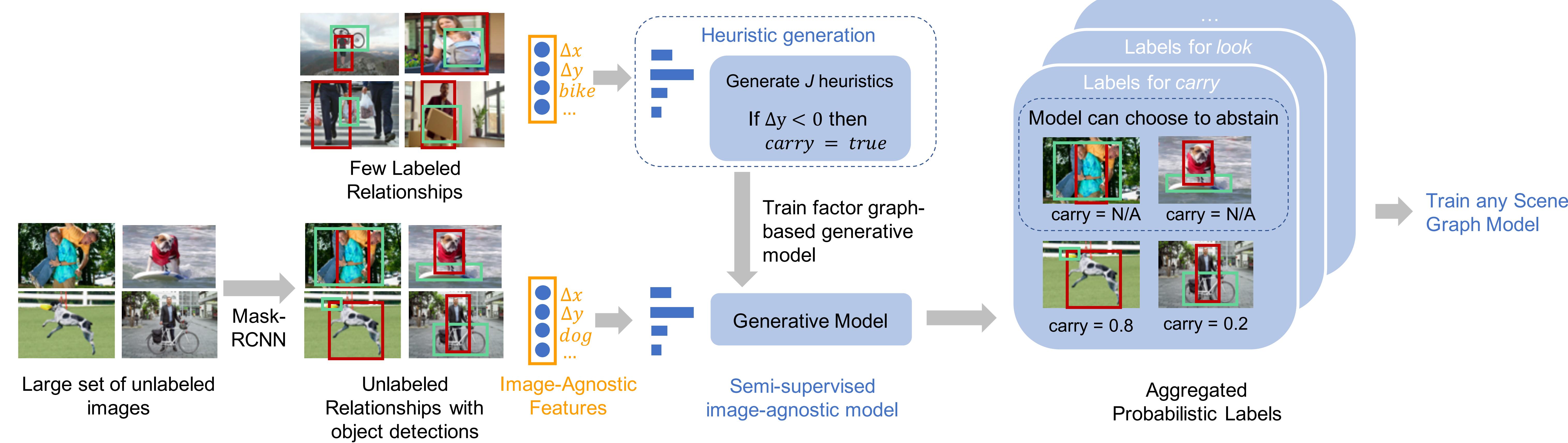


Idea: Leverage *image-agnostic* features in heuristics for label generation



Generating Labels for Structured Predictions

- I. Feature extraction: Extract spatial + categorical features based on object bounding boxes.
 2. Heuristic Generation: Using a limited set of labeled data, automatically generate noisy heuristics.
 3. Generative Model: Combine the noisy heuristics' outputs into probabilistic training labels.

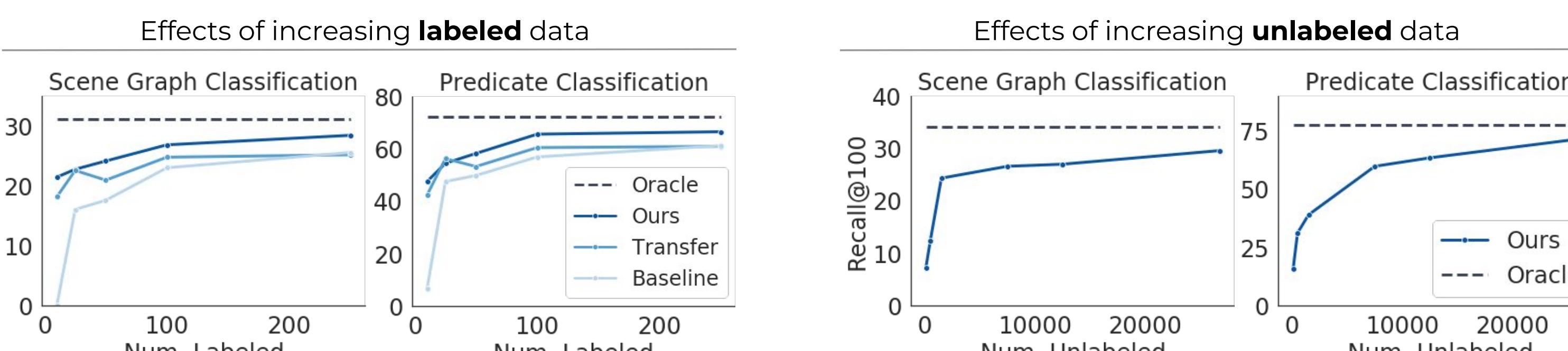


Experimental Results

Our approach outperforms naïve baselines (B, F+O) using only limited labels, semi-supervised methods (DT, LP) relying on image-agnostic features to learn patterns over labeled/unlabeled data, and transfer learning, which pretrains on the set of available relationships and fine-tunes on limited relationship labels.

Model	Scene Graph Detection			Scene Graph Classification			Predicate Classification		
	R@20	R@50	R@100	R@20	R@50	R@100	R@20	R@50	R@100
BASELINE [$n = 10$]	0.00	0.00	0.00	0.04	0.04	0.04	3.17	5.30	6.61
FREQ+OVERLAP	10.16	10.84	10.86	9.90	9.91	9.91	20.39	20.90	22.21
TRANSFER LEARNING	11.99	14.40	16.48	17.10	17.91	18.16	39.69	41.65	42.37
DECISION TREE	11.11	12.58	13.23	14.02	14.51	14.57	31.75	33.02	33.35
LABEL PROPAGATION	6.48	6.74	6.83	9.67	9.91	9.97	24.28	25.17	25.41
OURS (CATEG. + SPAT.)	17.67	18.69	19.28	20.91	21.34	21.44	45.49	47.04	47.53

Our approach improves with increasing labeled and unlabeled data.

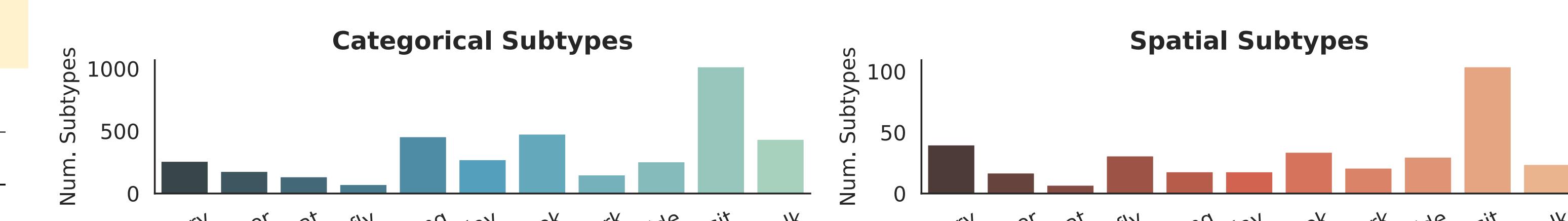


Challenges: Semantically similar phrasing / synonymous relationships



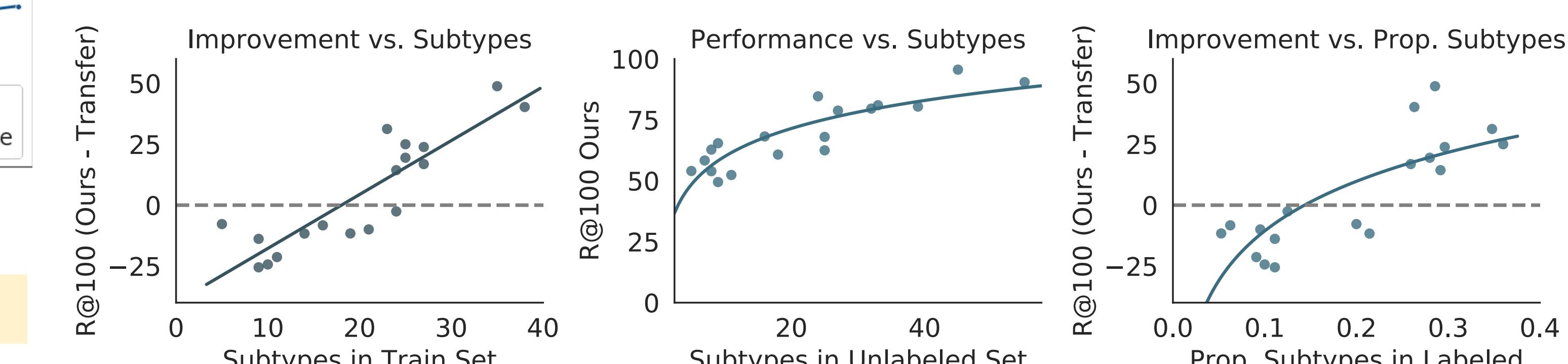
Studying Visual Relationship Complexity

Relationship subtypes capture the different ways that a visual relationship manifests in the dataset.



Categorical Subtype:
Count the number of object categories
for this relationship

Trends: With \uparrow relationship complexity (defined by subtypes), our weak supervision approach improves relative to transfer learning



Code: <https://github.com/vincentsschen/limited-label-scene-graphs>
Snorkel: <snorkel.org>