ICASSP 2024 KOREA

Boosting Zero-Shot HOI Detection with Vision-Language Transfer

Guwahati ndia "Stitute of Technolog"

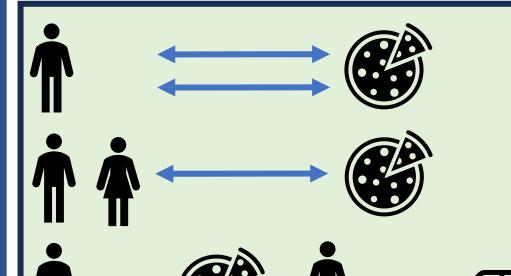
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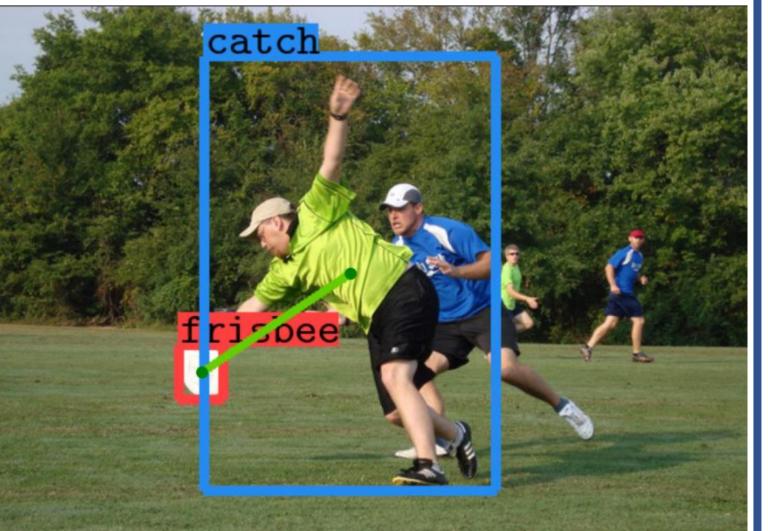
Human-Object Interaction Detection

Objectives

□ Localize interactive human-object pairs □ Recognize interactive object □ Recognize interaction class

Scenarios to handle





Dataset: HICO-DET

Goo HOI categories with human and object bounding-box annotations, and interaction labels

Total images = 47,776

Bo object categories

117 verb categories

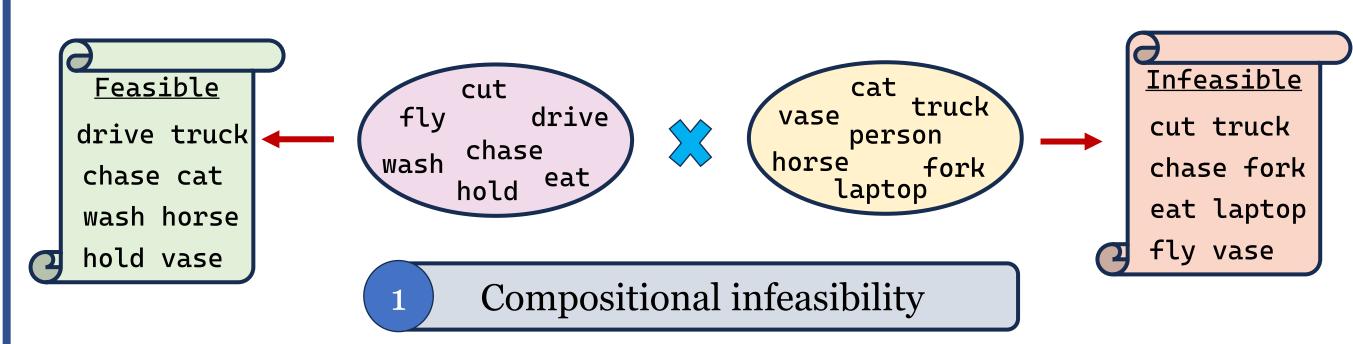
Splits

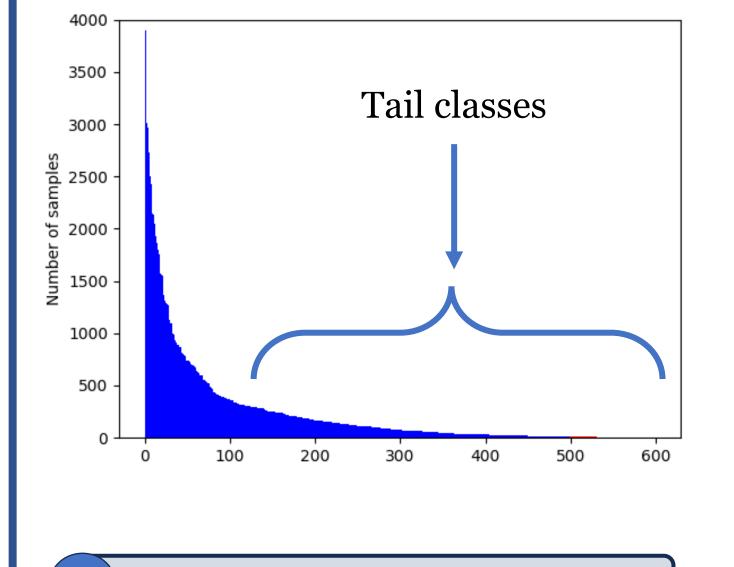
Experiments

Setting	Object	Verb	Interaction
UC	All	All	480 S / 120 U
RF-UC	All	All	480 S / 120 U
NF-UC	All	All	480 S / 120 U
UA	All	95 S / 22 U	500 S / 100 U
UO	68 S / 12 U	All	500 S / 100 U



Problems zero-shot learning can mitigate in HOID







ride elephant ride bicycle

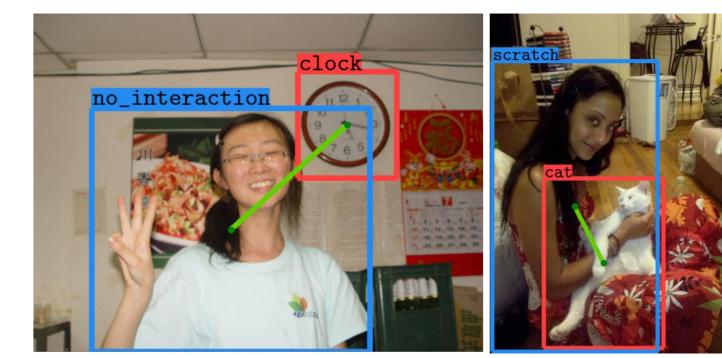


ride skateboard ride boat

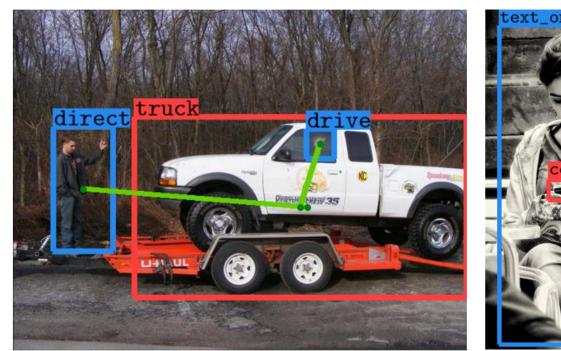
Zero-shot results

Method	Type	Full	Seen	Unseen
VCL baseline [6]	RF-UC	15.56	18.63	3.30
VCL-COCO [6]	RF-UC	16.58	18.84	7.55
VCL-HICODET [6]	RF-UC	21.43	24.28	10.06
ATL [3]	RF-UC	21.57	24.67	9.18
FCL baseline [5]	RF-UC	21.13	24.18	8.94
FCL [5]	RF-UC	22.01	24.23	13.16
THID [8]	RF-UC	22.96	24.32	15.53
Ours	RF-UC	24.51	26.90	16.50
VCL baseline [6]	NF-UC	11.23	12.77	5.06
VCL-COCO [6]	NF-UC	12.76	13.67	9.13
VCL-HICODET [6]	NF-UC	18.06	18.52	16.22
ATL [3]	NF-UC	18.67	18.78	18.25
FCL baseline [5]	NF-UC	18.07	19.22	13.47
FCL [5]	NF-UC	19.37	19.55	18.66
Ours	NF-UC	20.64	20.93	19.47
SHOI [1]	UC	6.26	-	5.62
FG [4]	UC	12.45	12.74	11.31
ConsNet [2]	UC	14.48	14.74	13.46
ZSHOI-AG [3]	UC	11.03	-	9.80
Ours	UC	23.79	24.90	19.36
ConsNet [6]	UA	14.35	14.72	12.50
Ours	UA	25.78	26.67	21.36
FG [4]	UO	13.84	14.36	11.22
ConsNet [2]	UO	14.48	14.67	13.51
ATL baseline [3]	UO	19.33	20.63	12.84
ATL-HICODET [3]	UO	19.36	20.96	11.35
ATL-COCO [3]	UO	20.47	21.54	15.11
FCL baseline [5]	UO	19.45	20.77	12.86
FCL [5]	UO	19.87	20.74	15.54
Ours	UO	21.70	23.26	13.94

Qualitative results



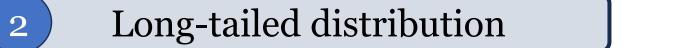
Detected human and object, but no interaction detected Detecting tail interactions



Multiple humans, same object, different interactions

Interaction with small object

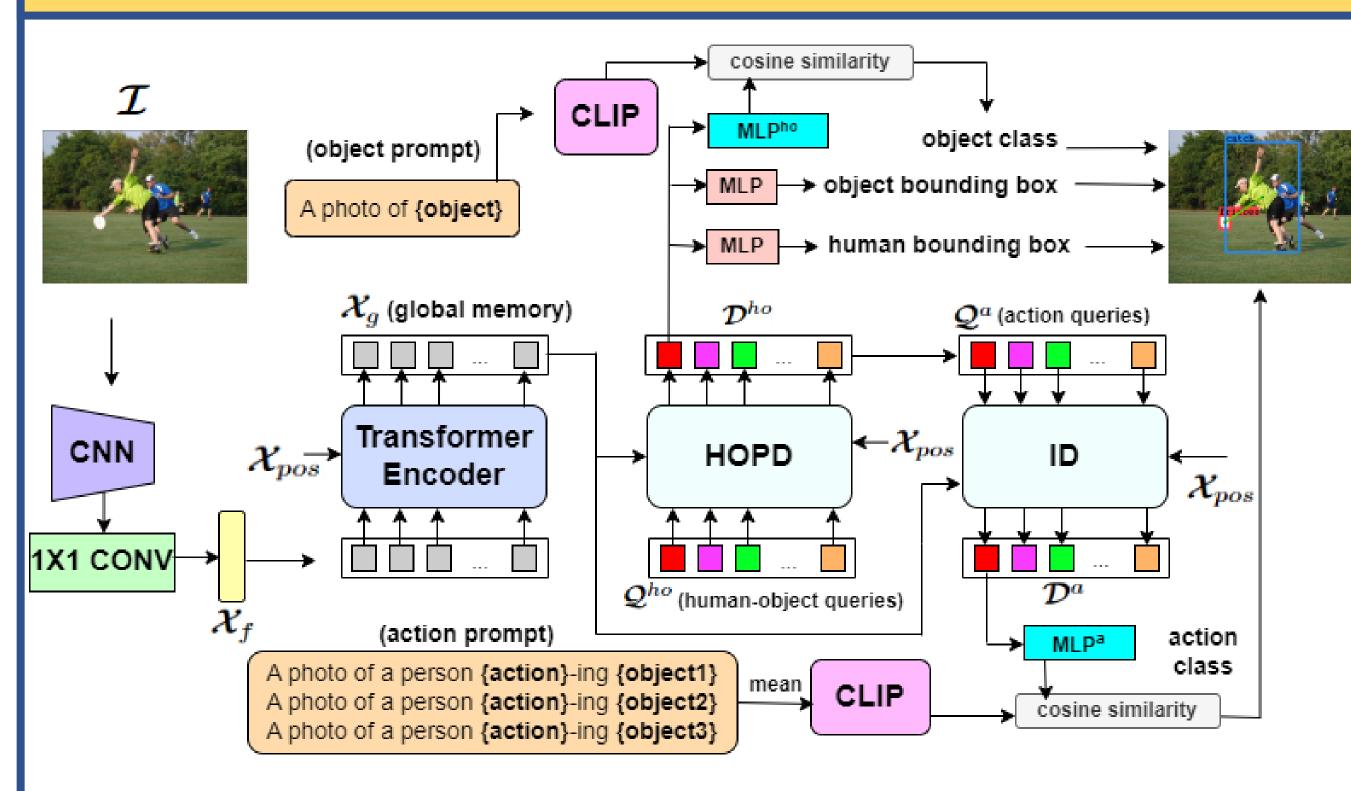
Comparative analysis



2



Proposed Framework

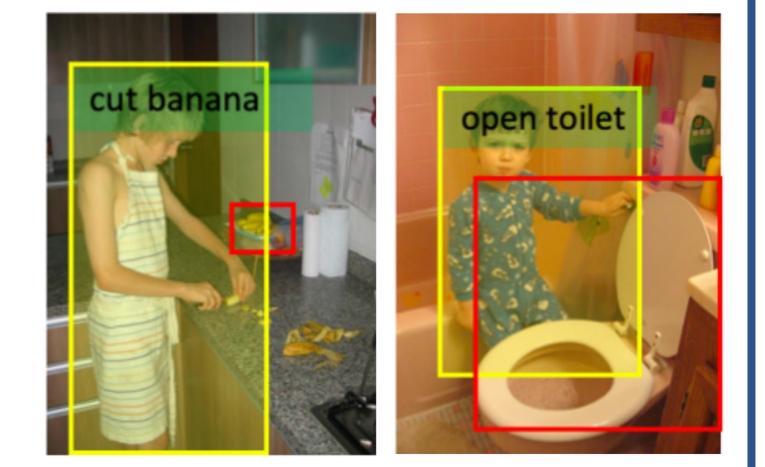


Visual features : Extracted from ResNet-50 backbone

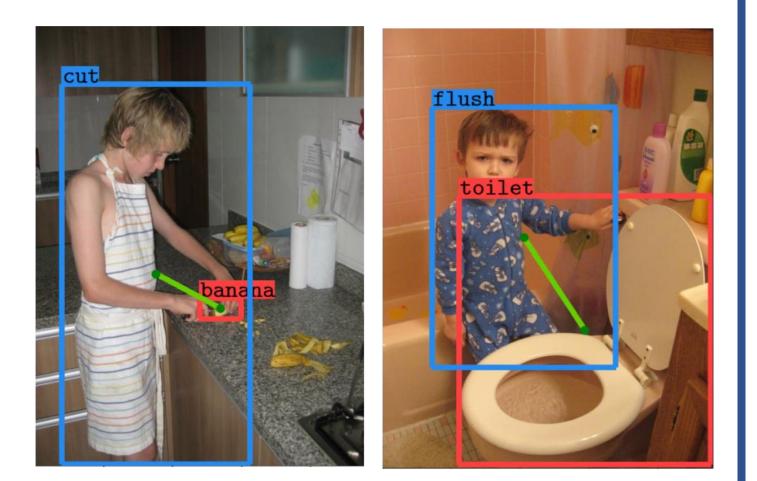
Semantic features for **object** : CLIP text encoder (objects are welldefined concepts)

Ablation on text embeddings

Text-Embedding	Туре	Full	Seen	Unseen
GAT	RF-UC	21.57	23.56	14.92
CLIP	RF-UC	24.51	26.90	16.50
GAT	NF-UC	18.25	18.37	17.76
CLIP	NF-UC	20.64	20.93	19.47
GAT	UC	21.78	23.25	15.91
CLIP	UC	23.79	24.90	19.36
GAT	UA	18.06	19.77	9.74
CLIP	UA	25.78	26.67	21.36
GAT	UO	19.71	21.21	12.16
CLIP	UO	21.70	23.26	13.94



Failure case of FCL



Our improved results

Takeaways

- **Semantic** features for **verbs** : CLIP text encoder (actions are abstract concepts).
- □ **Infeasible combinations** like "*feed chair*" are ignored while picking the different objects that can be combined with the action "*feed*" in order to form an interaction
- **Transformer encoder** captures global information via self-attention
- **Human-object pair** is anticipated using a pair decoder
- □ **Interaction features** are produced by the interaction decoder
- **Zero-shot** interaction classification is made possible by combing final interaction features with semantic features from CLIP text encoder

- □ Transformer's **attention** mechanism helps to utilize contextually important cues
- □ Joint visual-and-text modeling using **CLIP** helps in generalizing to unseen HOIs
- □ The query vectors in our DETR-based framework are vital in projecting an idea about "what" visual information about the human-object pairs to look for, with each vector element suggesting *"where"* to look for these pairs within the image. Since the final task is to detect human-object pairs, **unified query vectors** for human-object pairs are important **Improved prompting** to obtain semantic representations of HOI classes from action-object combinations aids in zero-shot HOI detection
- Despite the unavailability of certain actions and objects (such as in **UA and UO settings**), our method is better at detecting unseen interactions in such challenging settings