PIN:

Positional Insert Unlocks Object Localisation Abilities in VLMs

MICHAEL DORKENWALD, NIMROD BARAZANI, CEES G. M. SNOEK*, YUKI M. ASANO*

CVPR'24

Prompt 1: Provide a bounding box around the cat Prompt 2: Localise the cat in the image





Prompt 1: Provide a bounding box around the cat Prompt 2: Localise the cat in the image





P2: The cat in the image is sitting on the right side [...]

Prompt 1: Provide a bounding box around the cat Prompt 2: Localise the cat in the image





P2: The cat in the image is sitting on the right side [...] GPT-4V **P1**: To determine the size of the room.

P2: The cat in the image is localised in the image.



Prompt 1: Provide a bounding box around the cat Prompt 2: Localise the cat in the image





P2: The cat in the image is sitting on the right side [...] GPT-4V **P1**: To determine the size of the room.

P2: The cat in the image is localised in the image.

🌤 OpenFlamingo

P1: Cats are not fond of being confined in a small space.

P2: Yes, you can do that



Prompt 1: Provide a bounding box around the cat Prompt 2: Localise the cat in the image





P2: The cat in the image is sitting on the right side [...] GPT-4V **P1**: To determine the size of the room.

P2: The cat in the image is localised in the image.

🌤 OpenFlamingo

P1: Cats are not fond of being confined in a small space.

P2: Yes, you can do that



P1: Provide a bounding box around the cat's plant

P2: <empty string>



Prompt 1: Provide a bounding box around the cat Prompt 2: Localise the cat in the image







P2: The cat in the image is sitting on the right side [...] **P1**: To determine the size of the room.

P2: The cat in the image is localised in the image.

🌤 OpenFlamingo

P1: Cats are not fond of being confined in a small space.

P2: Yes, you can do that



P1: Provide a bounding box around the cat's plant

P2: <empty string>

🕂 BLIP-2



Our approach



frozen VLM, e.g. Flamingo

Positional Insert (PIN) module

Synthetic, unlabelled data



Dorkenwald, Barazani, Snoek, Asano. PIN: Positional Insert Unlocks Object Localisation Abilities in VLMs, CVPR'24.

Synthetic data generation



- Self-Supervision Signal: Location is known via pasting
- Avoid collapse: Pasting multiple objects

University of Amsterdam

Š

Zhao et al. X-Paste: Revisiting Scalable Copy-Paste for Instance Segmentation using CLIP and StableDiffusion. ICML 2023

Example generated data

UNIVERSITY OF AMSTERDAM

Š



- Non-realism is **not** an issue, as vision encoder is kept completely frozen
- Pasting objects from categories that do not overlap with test data
 - · Zero-shot evaluation

Dorkenwald, Barazani, Snoek, Asano. PINs: Positional Insert unlocks object localisation abilities in VLMs, CVPR'24.

Overview of VLMs



Feed the frozen VLM synthetic data



Provide spatial learning capacity via PIN



13

What is PIN? It's a PEFT method for VLMs





Results (zero-shot)





Results PVOC (zero-shot)





Predictions

Ground Truth

Few-shot learning doesn't work

	Mathad	F	$PVOC_{\leq 3 \text{ Objects}}$			$COCO_{\leq 3 \text{ Objects}}$		
	Method	mIoU	$mIoU_M$	$mIoU_L$	mIoU	$mIoU_M$	$mIoU_L$	
	Baselines							
	raw	0	0	0	0	0	0	
OpenFlamingo [4]	random	$0.22{\pm}0.04$	$0.10{\pm}0.02$	$0.33{\pm}0.06$	0.12 ± 0.04	$0.07 {\pm} 0.02$	0.22 ± 0.08	
	2 context	$0.19{\pm}0.11$	$0.08{\pm}0.05$	$0.30{\pm}0.18$	0.10±0.08	$0.06{\pm}0.04$	$0.18{\pm}0.16$	
	5 context	0.19 ± 0.09	$0.07 {\pm} 0.04$	$0.31{\pm}0.15$	0.10 ± 0.08	$0.06 {\pm} 0.04$	0.20 ± 0.16	
nin	10 context	0.20 ± 0.11	0.06 ± 0.03	$0.32{\pm}0.18$	0.09±0.07	$0.05{\pm}0.04$	$0.17{\pm}0.14$	
Flam	Prompt-learning							
en	CoOp on ϕ_V	0.28	0.11	0.43	0.22	0.10	0.39	
OpenF	VPT on F	0.32	0.14	0.50	0.25	0.12	0.46	
	VPT on ϕ_V	0.42	0.21	0.61	0.33	0.22	0.57	
	LoRA on ϕ_V	0.44	0.26	0.62	0.33	0.23	0.58	
	PIN (ours)	0.45	0.27	0.62	0.35	0.26	0.59	
2 [31]	Prompt-learning							
	VPT on F	0.37	0.16	0.56	0.29	0.15	0.53	
4	VPT on ϕ_V	0.31	0.13	0.47	0.26	0.11	0.46	
BL	PIN (ours)	0.44	0.24	0.63	0.34	0.22	0.60	

PIN enables localisation

Method		PVOC<3 Objects			COCO _{<3 Objects}		
		mIoU	$m\bar{IoU}_M$	$mIoU_L$	mIoU	$m\bar{IoU}_M$	$mIoU_L$
	Baselines						
	raw	0	0	0	0	0	0
	random	$0.22{\pm}0.04$	$0.10{\pm}0.02$	$0.33{\pm}0.06$	$0.12{\pm}0.04$	$0.07{\pm}0.02$	$0.22{\pm}0.08$
4	2 context	$0.19{\pm}0.11$	$0.08{\pm}0.05$	$0.30{\pm}0.18$	0.10 ± 0.08	$0.06{\pm}0.04$	$0.18{\pm}0.16$
go	5 context	$0.19{\pm}0.09$	$0.07{\pm}0.04$	$0.31{\pm}0.15$	$0.10 {\pm} 0.08$	$0.06{\pm}0.04$	0.20 ± 0.16
nin	10 context	$0.20{\pm}0.11$	$0.06{\pm}0.03$	$0.32{\pm}0.18$	0.09 ± 0.07	$0.05{\pm}0.04$	$0.17{\pm}0.14$
Flai	Prompt-learning						
en]	CoOp on ϕ_V	0.28	0.11	0.43	0.22	0.10	0.39
Op	VPT on F	0.32	0.14	0.50	0.25	0.12	0.46
	VPT on ϕ_V	0.42	0.21	0.61	0.33	0.22	0.57
	LoRA on ϕ_V	0.44	0.26	0.62	0.33	0.23	0.58
	PIN (ours)	0.45	0.27	0.62	0.35	0.26	0.59
1	Prompt-learning						
2 [3	VPT on F	0.37	0.16	0.56	0.29	0.15	0.53
Ľ.	VPT on ϕ_V	0.31	0.13	0.47	0.26	0.11	0.46
BLJ	PIN (ours)	0.44	0.24	0.63	0.34	0.22	0.60

PIN outperforms other PEFT methods

	Mathad	$PVOC_{\leq 3 \text{ Objects}}$ $COCO_{\leq 3 \text{ Object}}$			ts		
	Method	mIoU	$mIoU_M$	$mIoU_L$	mIoU	$mIoU_M$	$mIoU_L$
	Baselines						
	raw	0	0	0	0	0	0
	random	$0.22{\pm}0.04$	0.10 ± 0.02	0.33 ± 0.06	$0.12{\pm}0.04$	$0.07{\pm}0.02$	$0.22{\pm}0.08$
4	2 context	$0.19{\pm}0.11$	0.08 ± 0.05	$0.30{\pm}0.18$	$0.10 {\pm} 0.08$	0.06 ± 0.04	0.18 ± 0.16
00	5 context	0.19 ± 0.09	$0.07{\pm}0.04$	$0.31{\pm}0.15$	0.10 ± 0.08	0.06 ± 0.04	0.20 ± 0.16
OpenFlaming	10 context	$0.20{\pm}0.11$	$0.06{\pm}0.03$	$0.32{\pm}0.18$	$0.09{\pm}0.07$	$0.05{\pm}0.04$	$0.17{\pm}0.14$
	Prompt-learningCoOp on ϕ_V VPT on FVPT on ϕ_V LoRA on ϕ_V \widehat{T} PIN (ours)	0.28 0.32 0.42 0.44 0.45	0.11 0.14 0.21 0.26 0.27	0.43 0.50 0.61 0.62 0.62	0.22 0.25 0.33 0.33 0.35	0.10 0.12 0.22 0.23 0.26	0.39 0.46 0.57 0.58 0.59
BLIP-2 [31]	Prompt-learningVPT on F VPT on ϕ_V \mathfrak{T} PIN (ours)	0.37 0.31 0.44	0.16 0.13 0.24	0.56 0.47 0.63	0.29 0.26 0.34	0.15 0.11 0.22	0.53 0.46 0.60

PIN works on other VLMs too

	Mathad	$PVOC_{\leq 3 \text{ Objects}}$			$COCO_{\leq 3 \text{ Objects}}$		
	Wiethou	mIoU	$m \overline{IoU}_M$	$mIoU_L$	mIoU	$mIoU_M$	$mIoU_L$
	Baselines						
	raw	0	0	0	0	0	0
	random	$0.22{\pm}0.04$	0.10 ± 0.02	0.33 ± 0.06	$0.12{\pm}0.04$	$0.07{\pm}0.02$	$0.22{\pm}0.08$
4	2 context	$0.19{\pm}0.11$	0.08 ± 0.05	0.30 ± 0.18	$0.10 {\pm} 0.08$	$0.06 {\pm} 0.04$	$0.18{\pm}0.16$
0	5 context	0.19 ± 0.09	$0.07{\pm}0.04$	$0.31{\pm}0.15$	0.10 ± 0.08	$0.06 {\pm} 0.04$	$0.20{\pm}0.16$
ning	10 context	$0.20{\pm}0.11$	$0.06{\pm}0.03$	$0.32{\pm}0.18$	$0.09{\pm}0.07$	$0.05{\pm}0.04$	$0.17{\pm}0.14$
Flar	Prompt-learning						
en	CoOp on ϕ_V	0.28	0.11	0.43	0.22	0.10	0.39
Op	VPT on F	0.32	0.14	0.50	0.25	0.12	0.46
	VPT on ϕ_V	0.42	0.21	0.61	0.33	0.22	0.57
	LoRA on ϕ_V	0.44	0.26	0.62	0.33	0.23	0.58
	B PIN (ours)	0.45	0.27	0.62	0.35	0.26	0.59
Ξ	Prompt-learning						
2 0	VPT on F	0.37	0.16	0.56	0.29	0.15	0.53
Ľ.	VPT on ϕ_V	0.31	0.13	0.47	0.26	0.11	0.46
BLJ	PIN (ours)	0.44	0.24	0.63	0.34	0.22	0.60

UNIVERSITY OF ANISTENDAN

With slight modification, can work on RefCOCO.



"Left black shirt"



"Old lady in between the players"



"A guy in red on left"



"Guy in orange"



"Right player"



"Top left apron strings"



"Pizza squares left"



"Pizza right front piece

in middle"



"A man black"

"A right person"





Thank you!