

CO2: Consistent Contrast for Unsupervised Visual Representation Learning

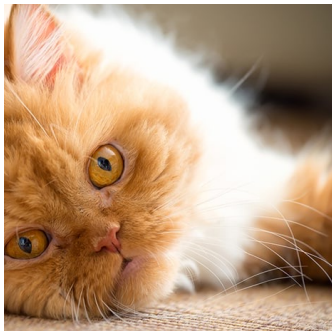
ICLR 2021

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Johns Hopkins University, Shanghai Jiao Tong University

Instance Discrimination

query



positive key



negative keys

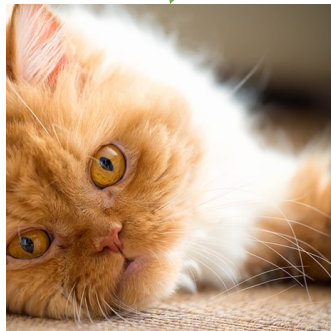
Instance Discrimination



query



...from the same instance



positive key



negative keys

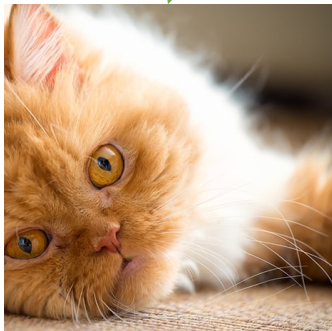
Instance Discrimination



query



...from the same instance



positive key



randomly sampled...

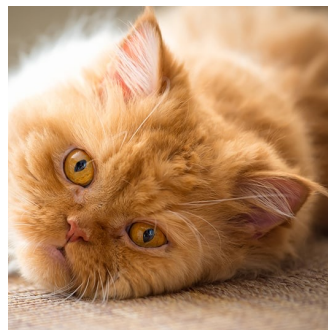
negative keys

Cross-Entropy Loss



...from the same instance

query



1

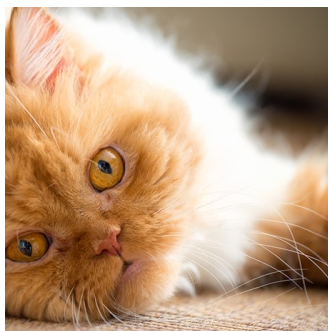
0

0

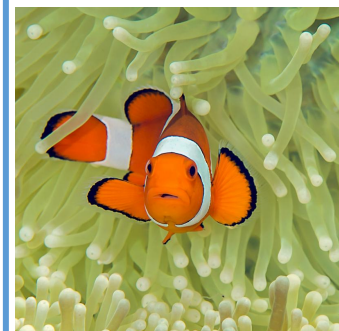
0

0

0



positive key



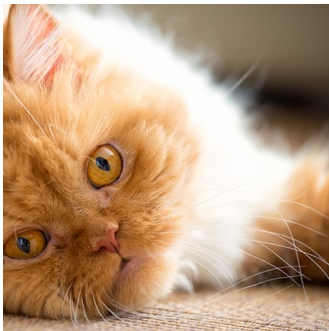
negative keys (randomly sampled)

Cross-Entropy Loss

query



1



0



0



0



0



0



Cross-Entropy Loss

query



self-**supervised**:
artificially created task
artificial **labels**

1



0



0



0



0



0

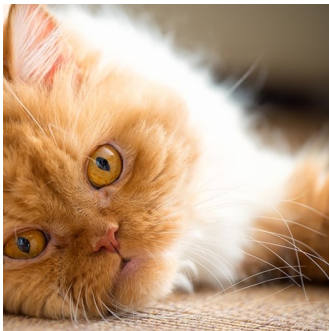


Negative?

query



1



0



0



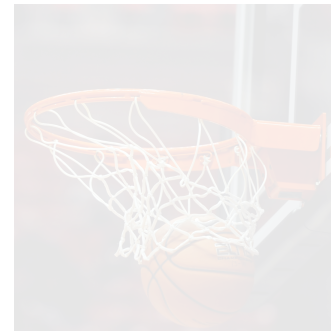
0



0



0



semantically similar!

Negative?

query



but we have no labels for this
query-negative similarity...

1



0



0



0



0

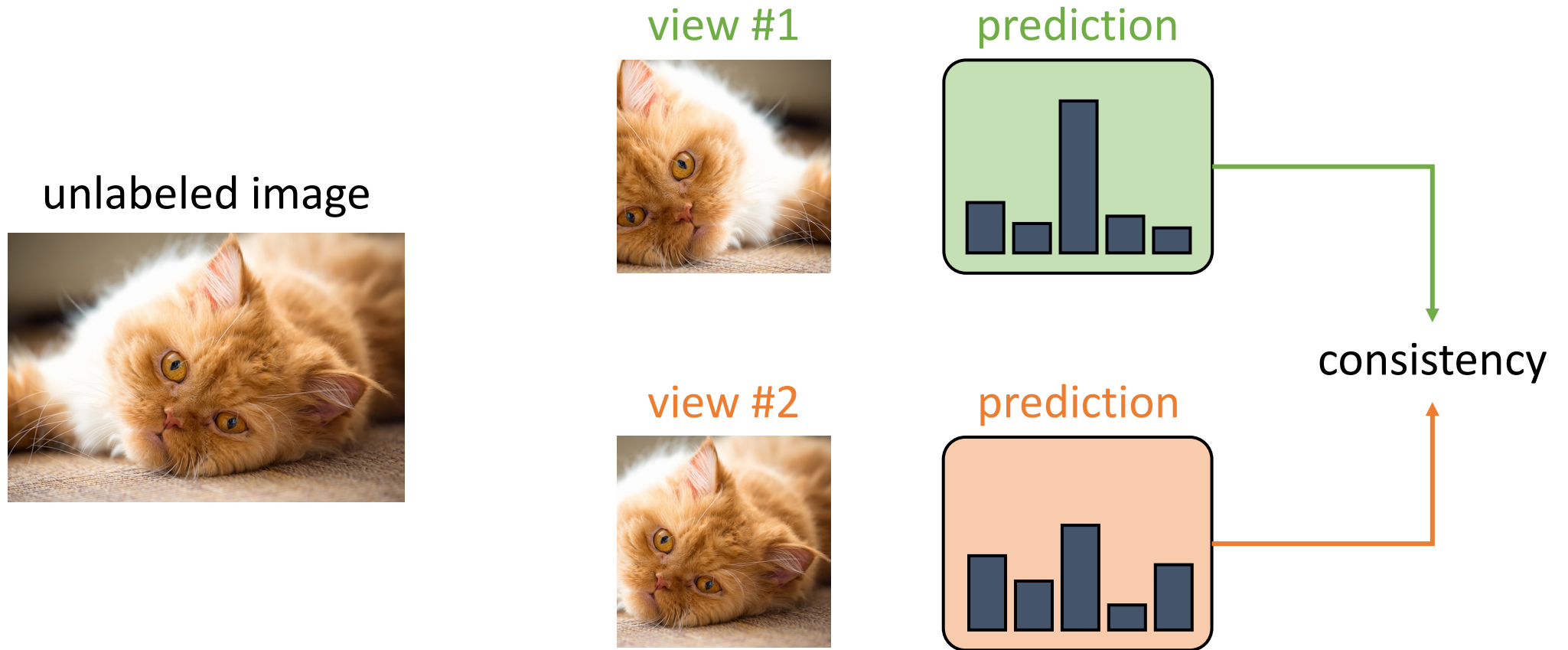


0



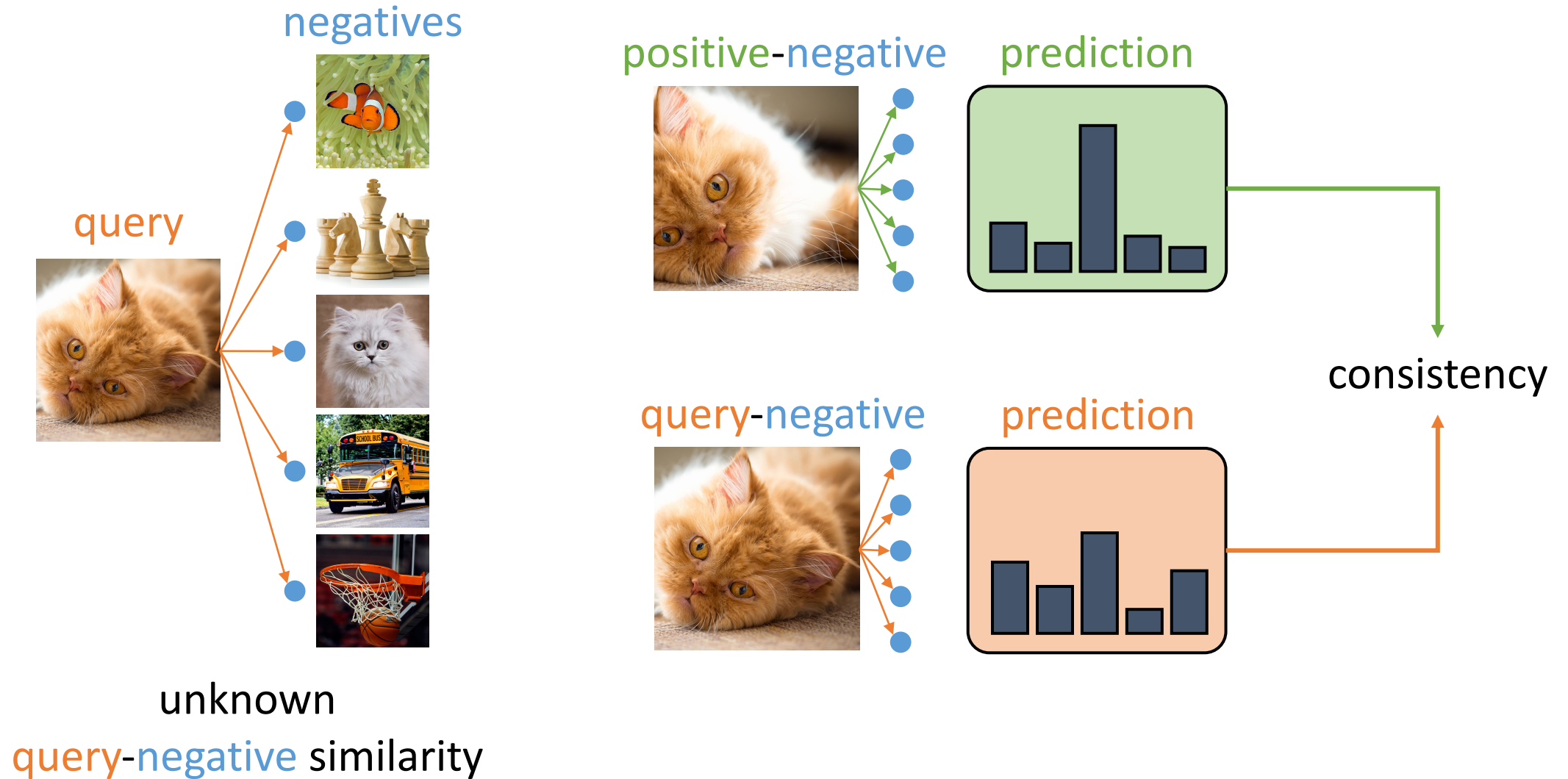
semantically similar!

Inspiration: Consistency Regularization

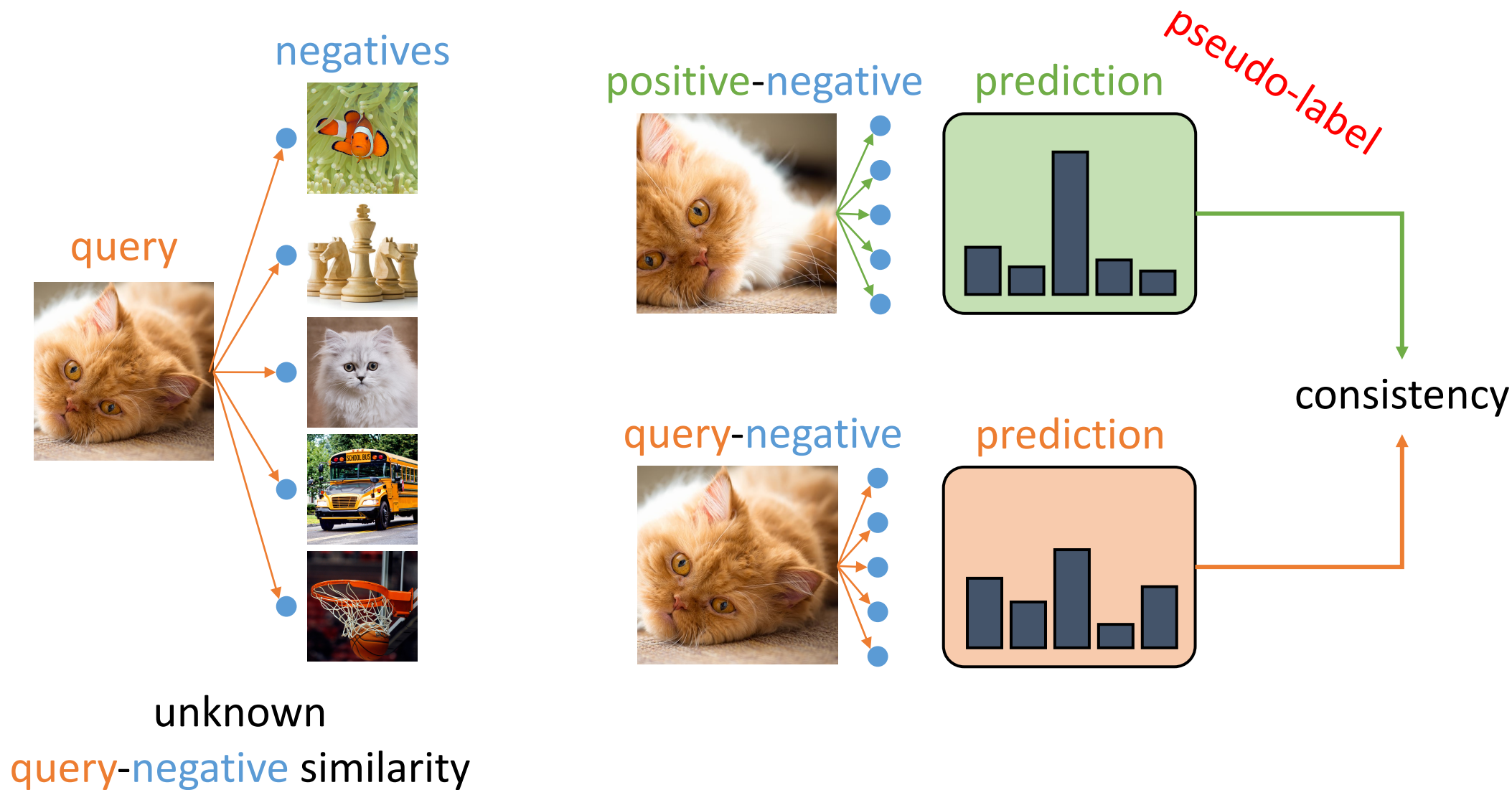


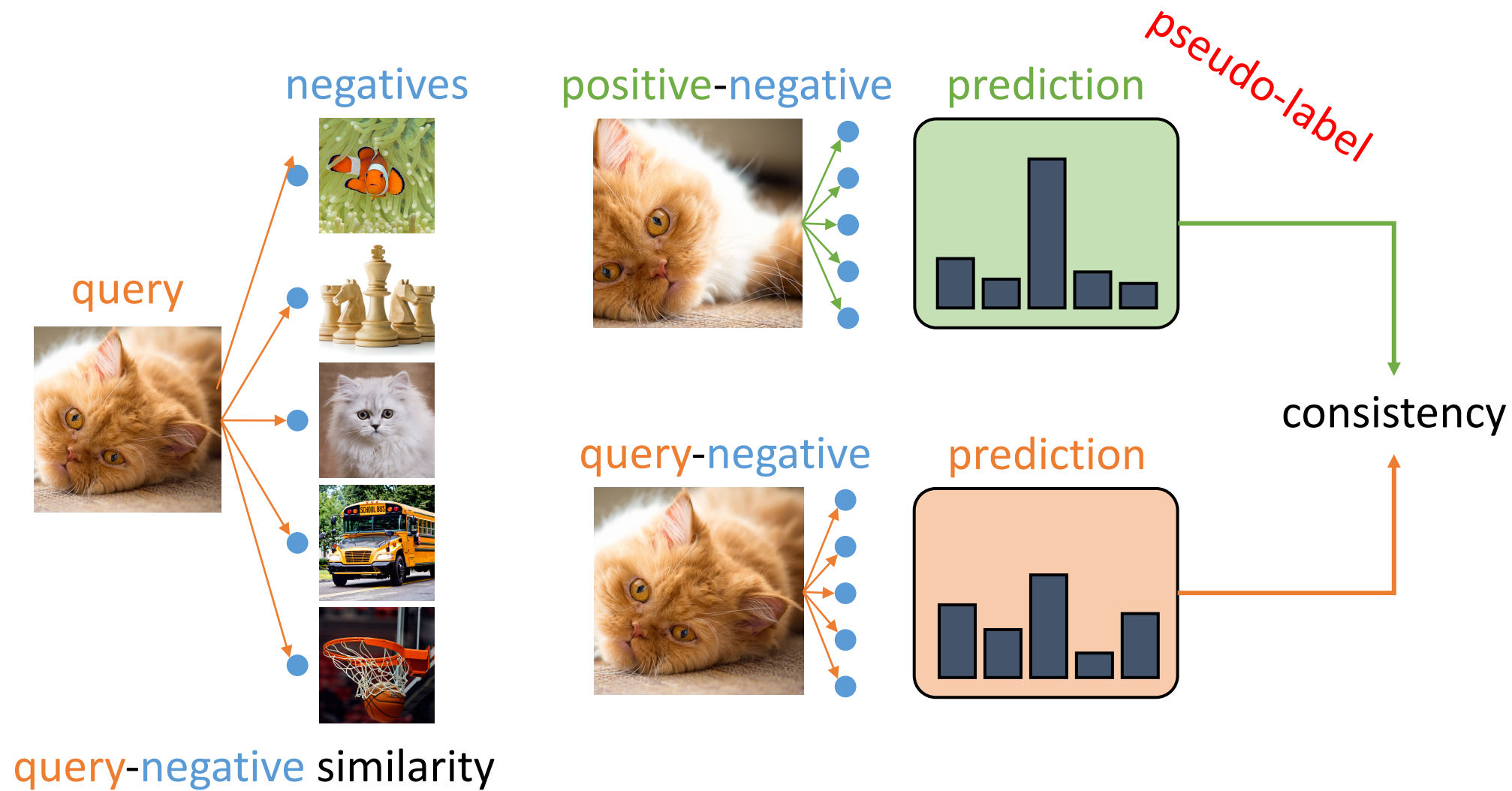
...effective in **semi**-supervised learning

Our Method: Consistent Contrast (CO2)

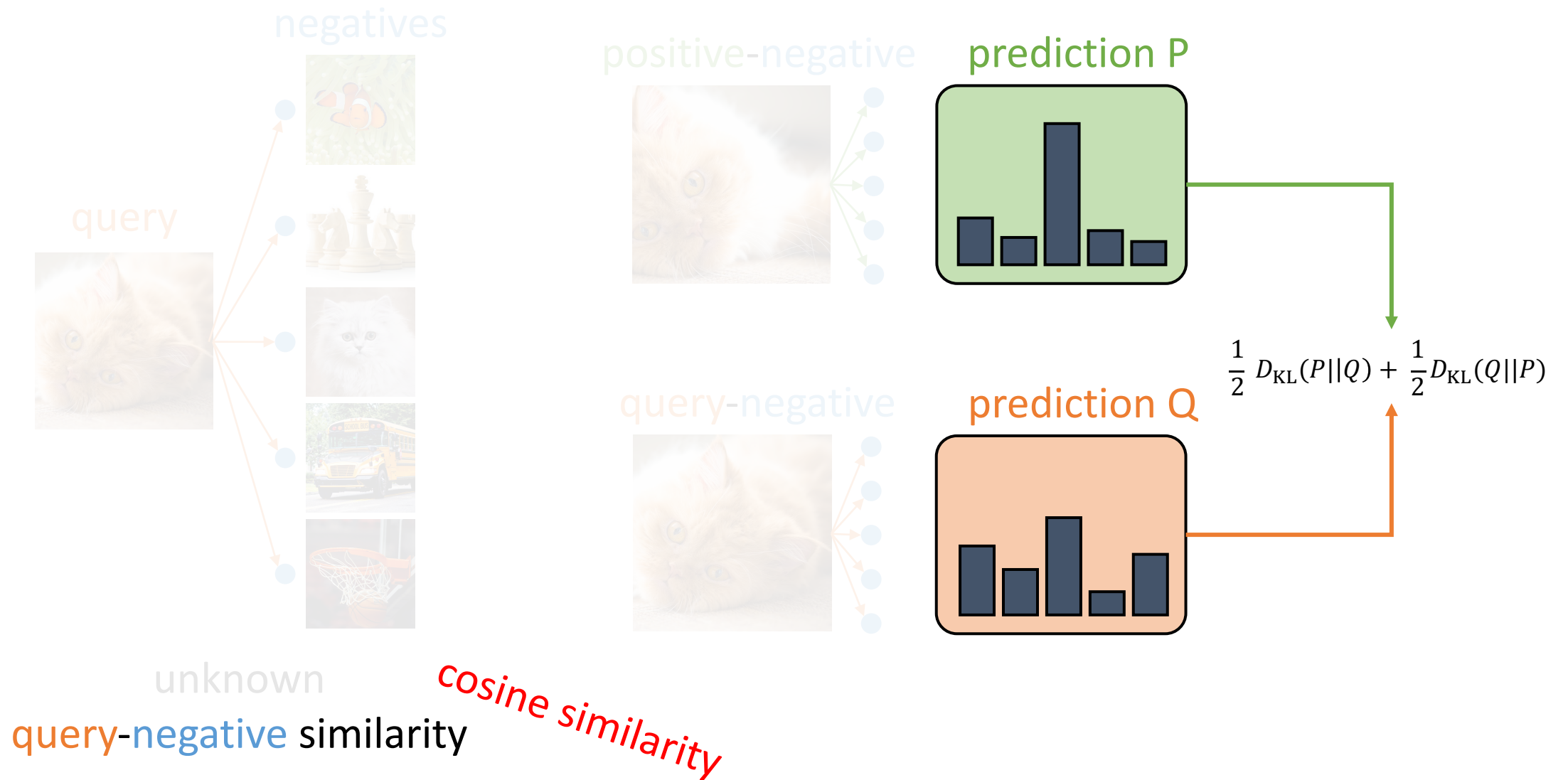


Our Method: Consistent Contrast (CO2)

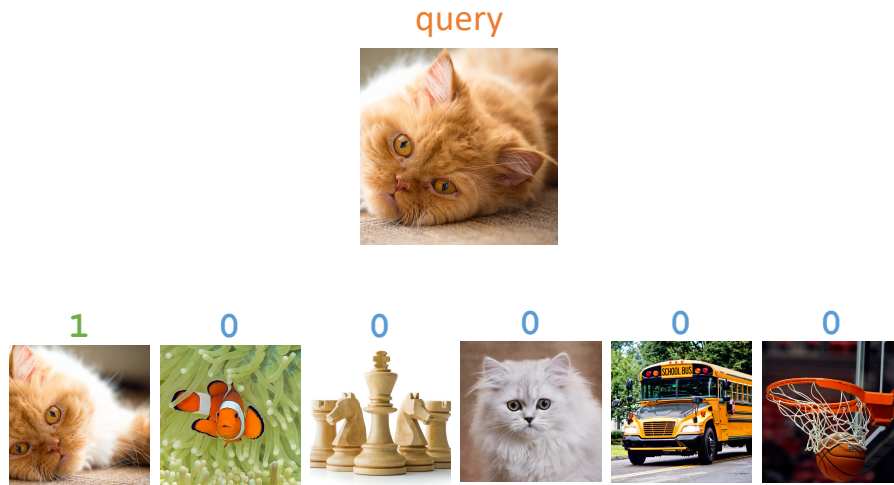




Details

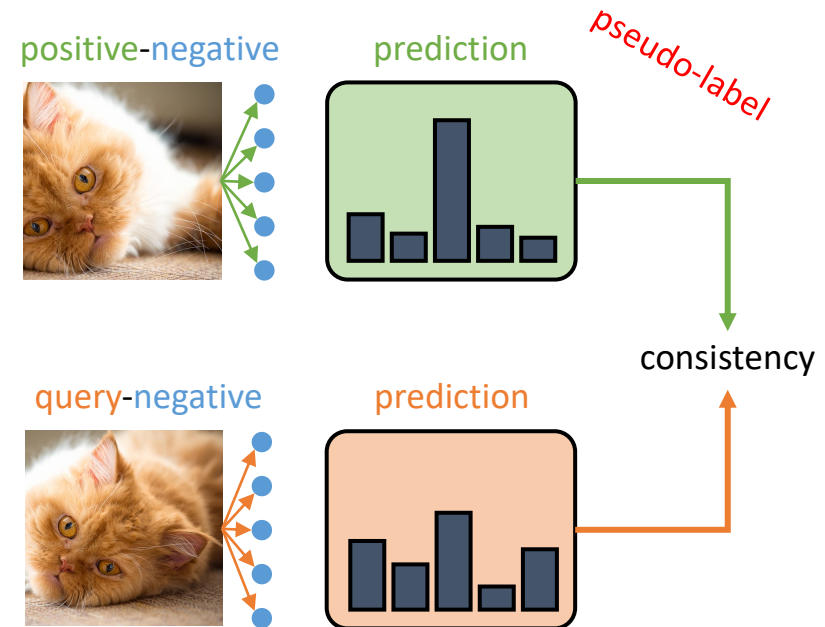


Comparison



Instance Discrimination

- Artificial one-hot label
- Ignoring query-negative similarity
- Self-supervised



Consistent Contrast

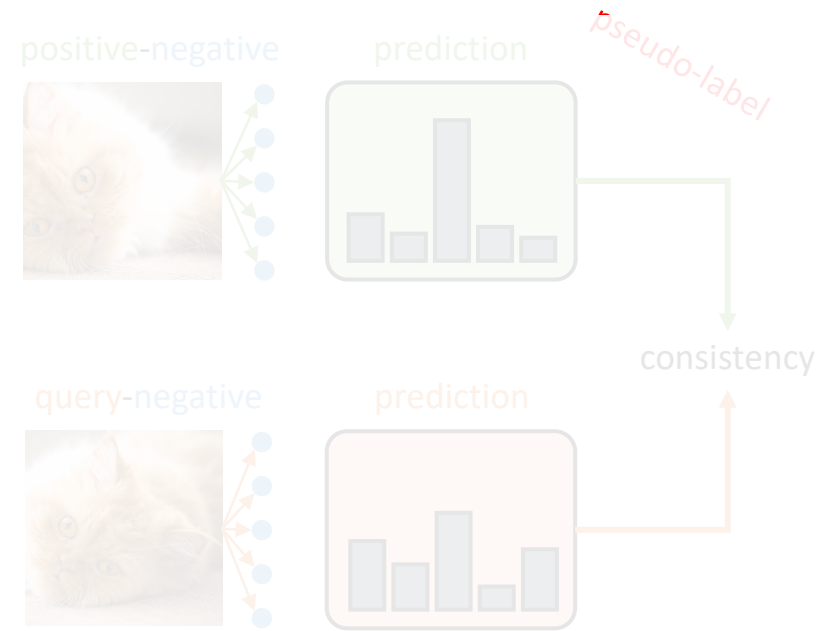
- Soft pseudo-label
- Learning query-negative similarity by consistency
- Self-semi-supervised

Combination



Instance Discrimination

- Artificially created one-hot label
- Ignoring query-negative similarity
- Self-supervised



$\alpha \cdot$ Consistent Contrast

- Soft pseudo-label
- Learning query-negative similarity by consistency
- Self-semi-supervised

Results: Linear Classifier

Pretext Task	Arch.	Head	#epochs	Top-1 Acc. (%)	ResNet50 ImageNet-1K
MoCo	R50	Linear	200	60.6	
MoCo + CO2	R50	Linear	200	63.5 (+2.9)	
MoCo v2	R50	MLP	200	67.5	
MoCo v2 + CO2	R50	MLP	200	68.0 (+0.5)	

Pretext Task	Arch.	Head	#epochs	Top-1 Acc. (%)	ResNet18 ImageNet-100
MoCo	R18	MLP	200	63.1	
MoCo + CO2	R18	MLP	200	69.7 (+6.6)	
SimCLR	R18	MLP	200	68.9	
SimCLR + CO2	R18	MLP	200	72.3 (+3.4)	

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ResNet50
ImageNet-1K

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ResNet18
ImageNet-100

Results: Transfer Learning

	Image Classification	Object Detection	Semantic Segmentation
Pretext Task	mAP	AP ₅₀	mIoU
MoCo	79.7	81.6	72.6
MoCo + CO2	82.6 (+2.9)	81.9 (+0.3)	73.3 (+0.7)
MoCo v2	85.0	82.4	74.2
MoCo v2 + CO2	85.2 (+0.2)	82.7 (+0.3)	74.7 (+0.5)

ResNet50

ImageNet-1K



VOC 2007

Results: Semi-Supervised Learning

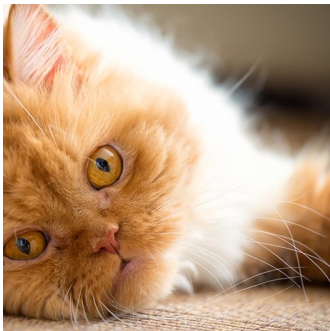
Pretext Task	1% Labels (Top-5)	10% Labels (Top-5)
MoCo	62.4	84.1
MoCo + CO2	66.2 (+3.8)	85.2 (+1.1)
MoCo v2	69.5	85.1
MoCo v2 + CO2	70.6 (+1.1)	85.4 (+0.3)

ResNet50
ImageNet-1K

More broadly...

- Rethinking the use of artificial labels...
 - Do they have to be known and clean?
 - Or they can also be noisy, partially-available...

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0?



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