

Relational Embedding for Few-Shot Classification

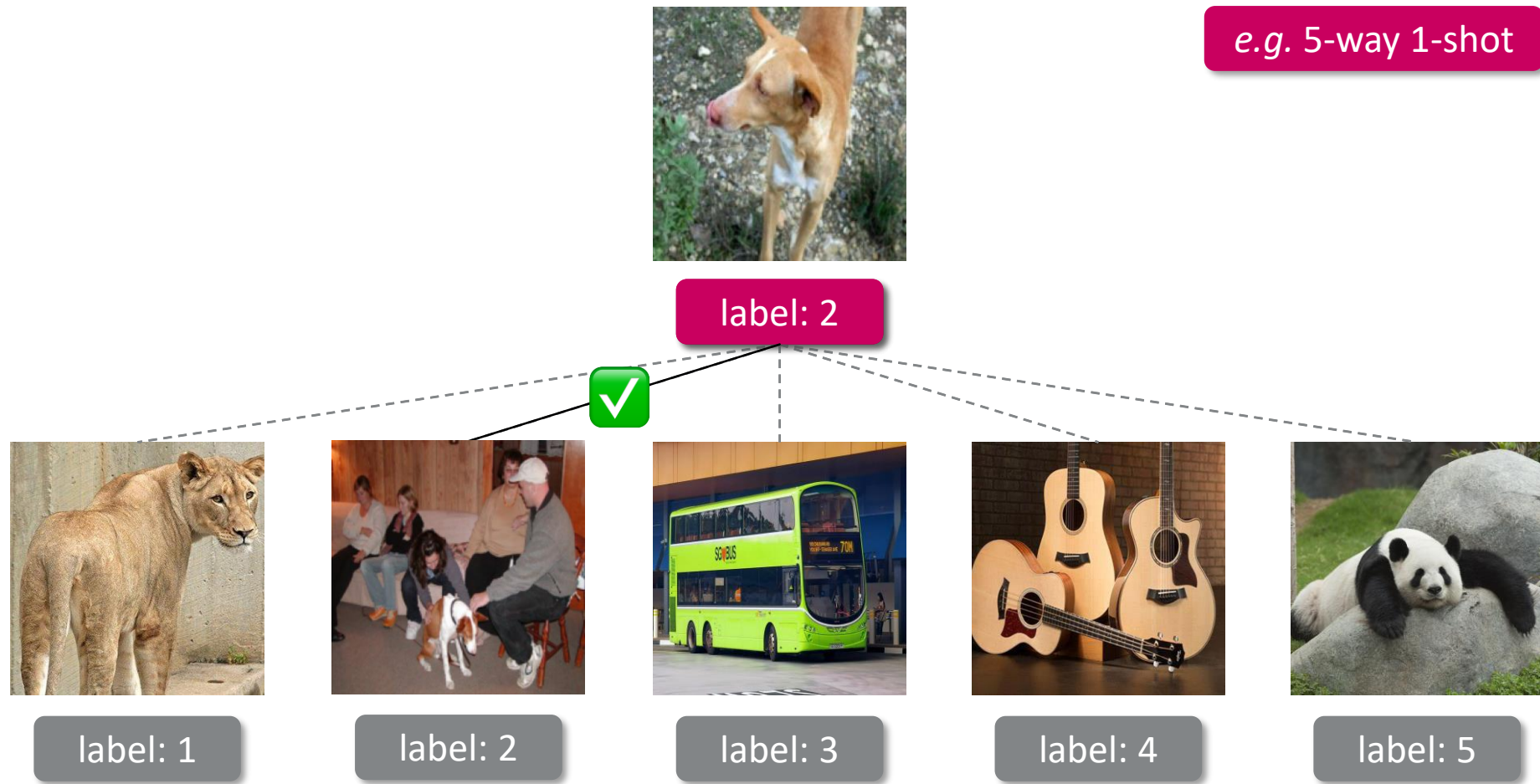
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Few-shot image classification

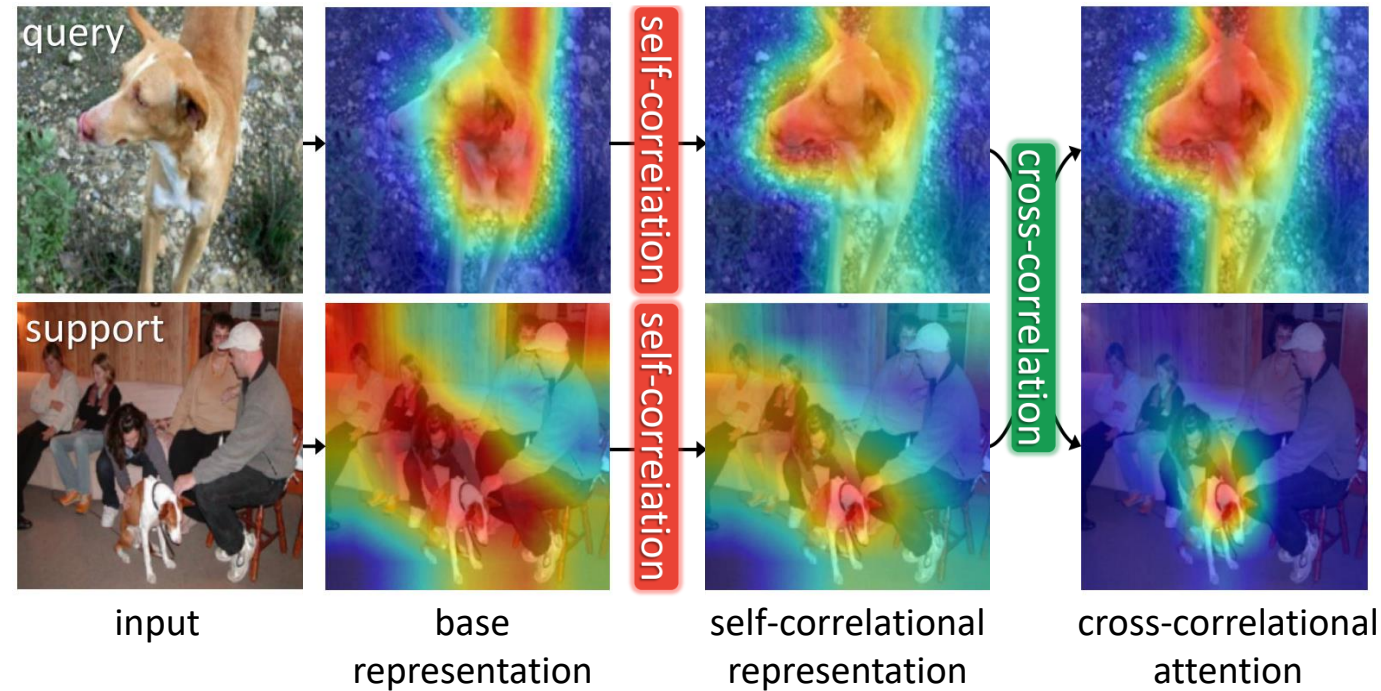


Few-shot image classification setup: train/test classes are disjoint

train classes	test classes (target classes)
bald eagle boxer bath towel ...	shopping cart pretzel pineapple ...
  	  

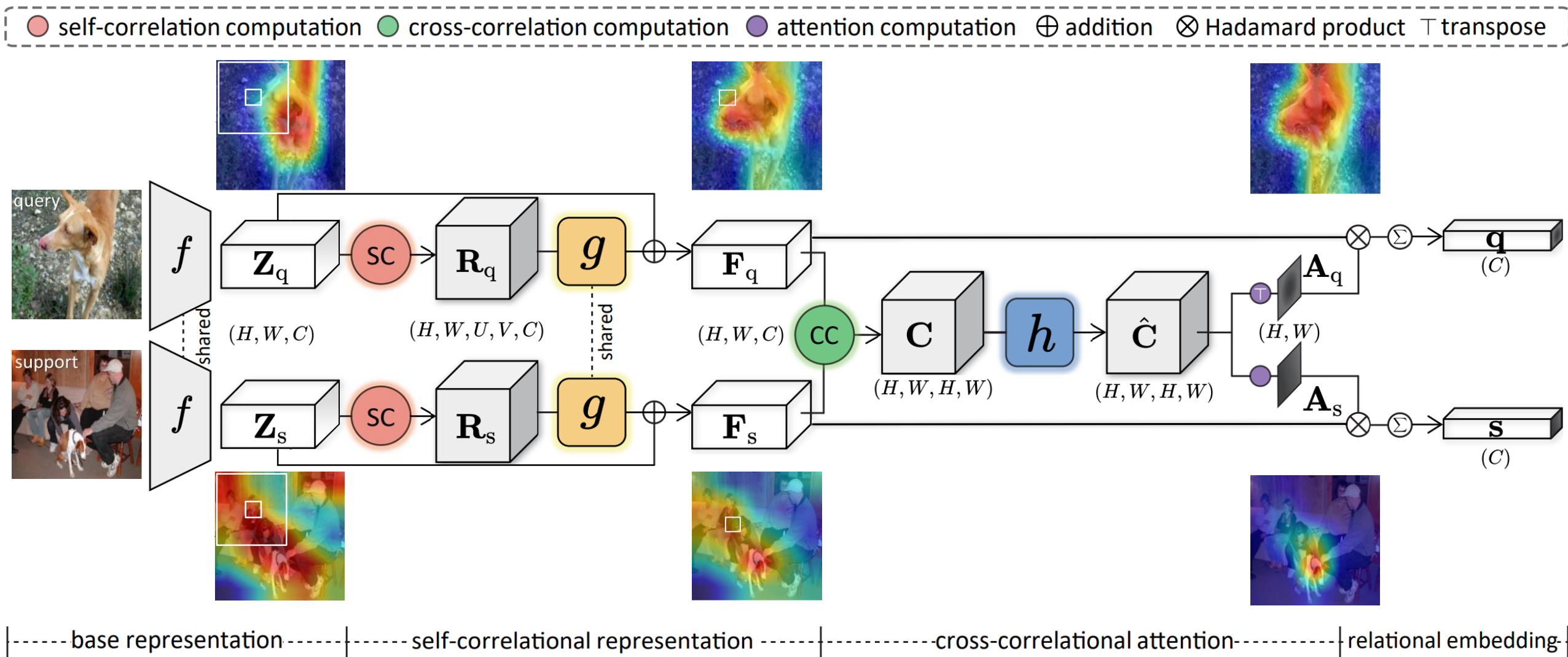
- Classifying an image into target classes that are *unseen during training*.

Relational Embedding Network (RENet)



- Relational embedding for a metric-based classifier
 - Learning *what to observe* via self-correlational representation (SCR) module
 - Learning *where to attend* via cross-correlational attention (CCA) module

Overall architecture



Comparison with state of the arts

method	venue	5-way 1-shot	5-way 5-shot
CTM (Li <i>et al.</i>)	ICCV'19	64.12 ± 0.82	80.51 ± 0.13
CAN (Hou <i>et al.</i>)	Neurips'19	63.85 ± 0.48	79.44 ± 0.34
FEAT (Ye <i>et al.</i>)	CVPR'20	66.78 ± 0.20	82.05 ± 0.14
DeepEMD (Zhang <i>et al.</i>)	CVPR'20	65.91 ± 0.82	82.41 ± 0.56
RENet (ours)	ICCV'21	67.60 ± 0.44	82.58 ± 0.30

minilImageNet

method	venue	5-way 1-shot	5-way 5-shot
CTM (Li <i>et al.</i>)	ICCV'19	68.41 ± 0.39	84.28 ± 1.73
CAN (Hou <i>et al.</i>)	Neurips'19	69.89 ± 0.51	84.23 ± 0.37
FEAT (Ye <i>et al.</i>)	CVPR'20	70.80 ± 0.23	84.79 ± 0.16
DeepEMD (Zhang <i>et al.</i>)	CVPR'20	71.16 ± 0.87	86.03 ± 0.58
RENet (ours)	ICCV'21	71.61 ± 0.51	85.28 ± 0.35

tieredImageNet

method	venue	5-way 1-shot	5-way 5-shot
NegMargin (Liu <i>et al.</i>)	ECCV'20	72.66 ± 0.85	89.40 ± 0.43
S2M2 (Mangla <i>et al.</i>)	WACV'20	72.92 ± 0.83	86.55 ± 0.51
FEAT (Ye <i>et al.</i>)	CVPR'20	73.27 ± 0.22	85.77 ± 0.14
DeepEMD (Zhang <i>et al.</i>)	CVPR'20	75.65 ± 0.83	88.69 ± 0.50
RENet (ours)	ICCV'21	79.49 ± 0.44	91.11 ± 0.24

CUB-200-2011

method	venue	5-way 1-shot	5-way 5-shot
MetaOptNet (Lee <i>et al.</i>)	CVPR'19	72.6 ± 0.7	84.3 ± 0.5
Boosting (Gidaris <i>et al.</i>)	ICCV'19	73.6 ± 0.3	86.0 ± 0.2
S2M2 (Mangla <i>et al.</i>)	WACV'20	62.77 ± 0.23	75.75 ± 0.13
RFS (Tian <i>et al.</i>)	ECCV'20	71.5 ± 0.8	86.0 ± 0.5
RENet (ours)	ICCV'21	74.51 ± 0.46	86.60 ± 0.32

CIFAR-FS

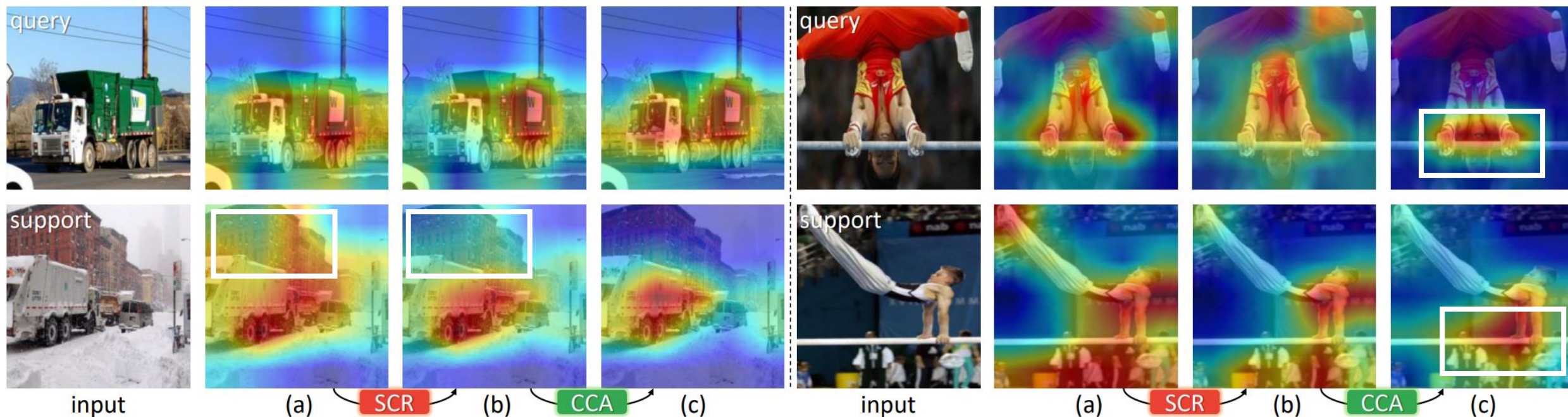
RENet achieves **~3.8%p improvement** from the recent state-of-the-arts models

Effects of SCR and CCA

SCR	CCA	<i>mini-</i> ImageNet	CUB
<i>×</i>	<i>×</i>	65.33	77.54
✓	<i>×</i>	66.66 (+1.33)	78.69 (+1.15)
<i>×</i>	✓	65.90 (+0.57)	78.49 (+0.95)
✓	✓	67.60 (+2.27)	79.49 (+1.95)

- The two proposed modules consistently improve the baseline model.

Qualitative results




(a): Channel activation of base representation.

(b): Channel activation of self-correlational representation.

(c): Cross-correlational attention maps.

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Paper	Code 	Project homepage	Author's homepage
arxiv.org/abs/2108.09666	github.com/dahyun-kang/renet	cvlab.postech.ac.kr/research/RENet	dahyun-kang.github.io