

# Attention Diversification for Domain Generalization

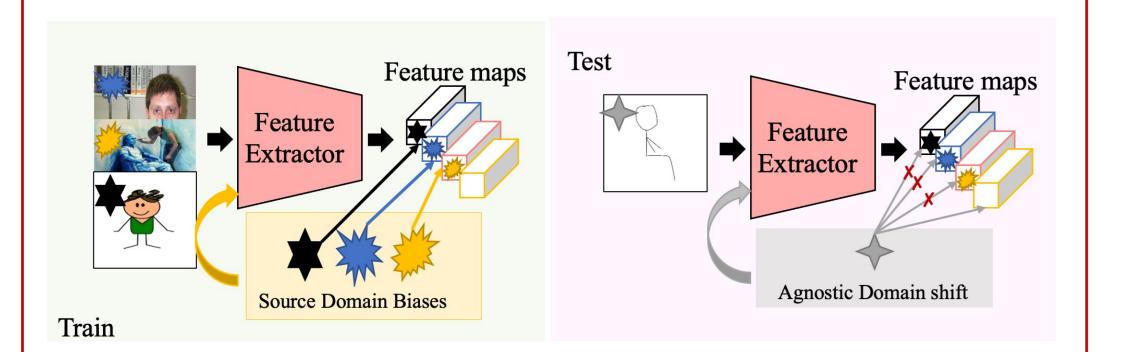


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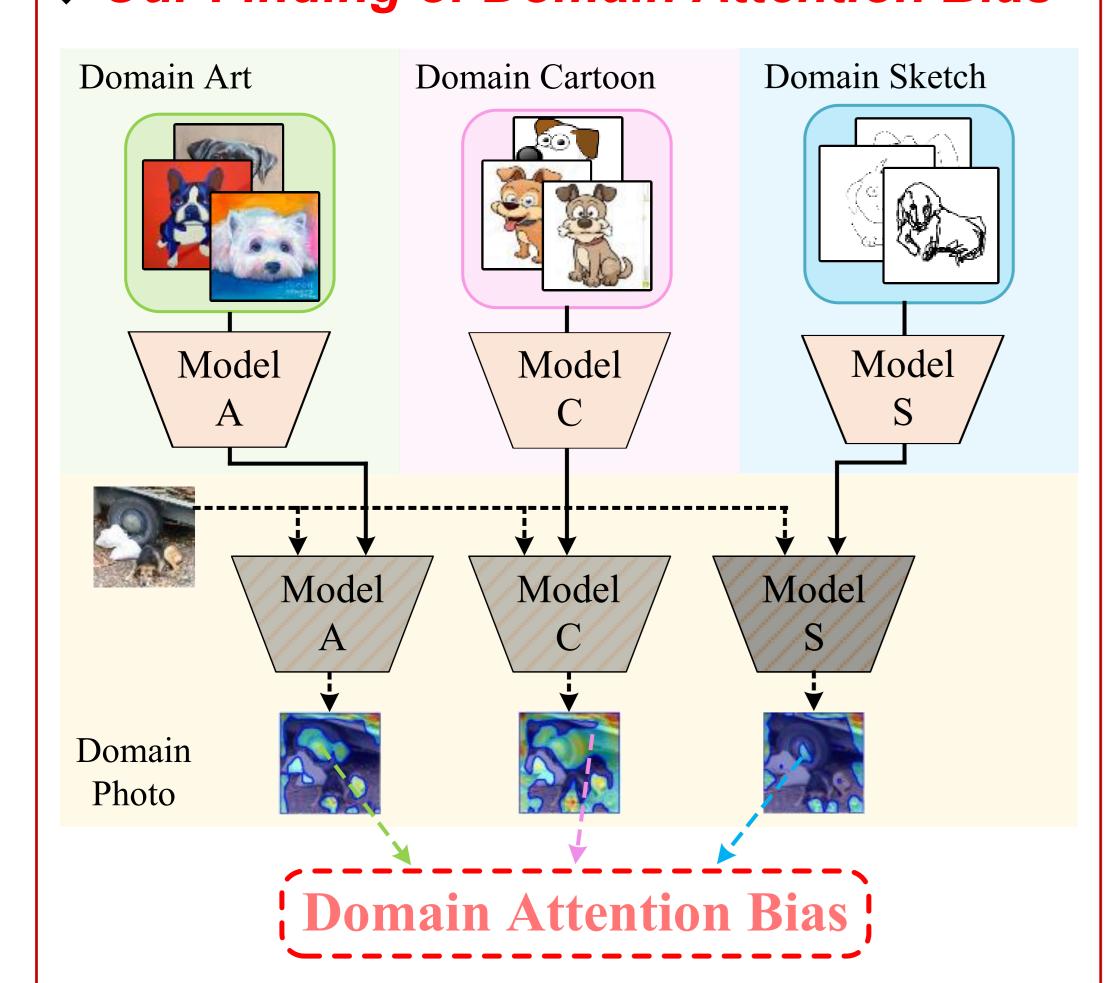
#### Motivation

Background

Deep models trained on seen domains perform poorly on unseen domains for the seen/unseen domain shifts.



◆ Our Finding of Domain Attention Bias



We find that models trained on different domains have different attention bias

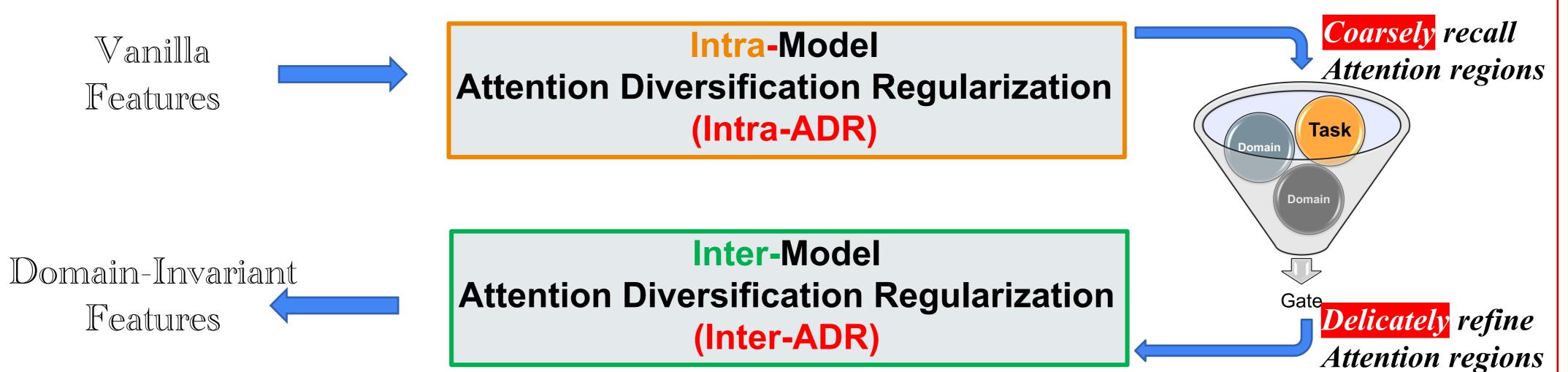
Our Revisiting of DG from Maximum Entropy Principle

Maximum Entropy Principle: when estimating the probability distribution, we should select that distribution with the largest uncertainty under given constraints.

Our Insight: when testing the unseen domains, each task-related attention is equally-useful (the maximum entropy)

## Methodology

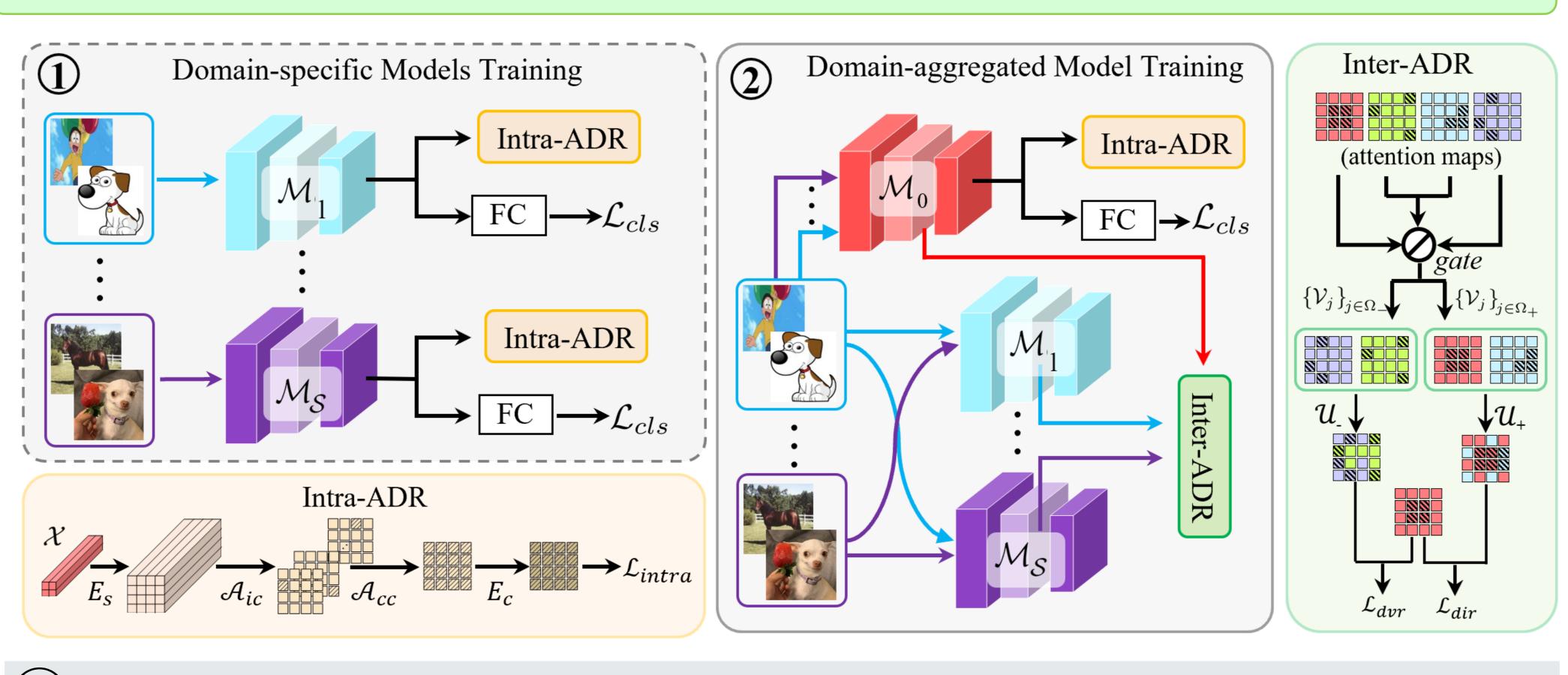
Workflow of Our Attention Diversification Framework



Training Scheme of Our Attention Diversification Framework

Intra-ADR forces different channels to concern on different regions and activates all the regions.

Inter-ADR uses a paradigm of "simulate, divide and assemble".



- 1) <u>Domain-specific Models Training</u> → Warmup step equipped with Intra-ADR and for "simulate" part in Inter-ADR.
- (2) <u>Domain-aggregated Model Training</u> → Main step equipped with Intraand Inter-ADR for coarse-to-fine Attention Diversification.



### Experiments

**♦** Results

Methods	References	Art	Cartoor	n Photo	Sketch	$ \mathbf{Avg.} $	Art	Cartoon	Photo	Sketch	$ \mathbf{A}\mathbf{v}\mathbf{g} $
Wiethous			I	ResNet-1a	8		F	$ResNet ext{-}50$	)		
Baseline	-	79.0	74.3	94.9	71.4	79.9	86.2	78.7	97.6	70.6	83.2
MetaReg [1]	NeurIPS'18	83.7	77.2	95.5	70.3	81.7	87.2	79.2	97.6	70.3	83.6
MASF [12]	NeurIPS'19	80.2	77.1	94.9	71.6	81.0	82.8	80.4	95.0	72.2	82.6
Epi-FCR [26]	ICCV'19	82.1	77.0	93.9	73.0	81.5	-	-	-	-	_
JiGen [4]	CVPR'19	79.4	75.2	96.0	71.3	80.5	-	-	-	-	_
DMG [5]	ECCV'20	76.9	80.4	93.4	75.2	81.5	82.6	78.1	94.5	78.3	83.4
$\mathrm{RSC}$ [21]	ECCV'20	84.4	80.3	95.9	80.8	85.1	87.8	82.1	97.9	83.3	87.9
MixStyle [78]	ICLR'21	84.1	78.8	96.1	75.9	83.7	-	-	-	-	_
SelfReg [24]	ICCV'21	82.3	78.4	$\overline{96.2}$	77.5	83.6	87.9	79.4	96.8	78.3	85.6
DAML [50]	CVPR'21	83.0	74.1	95.6	78.1	82.7	-	-	-	-	_
SagNet [42]	CVPR'21	83.6	77.7	95.5	76.3	83.3	81.1	75.4	95.7	77.2	82.3
FACT [65]	CVPR'21	85.4	78.4	95.2	79.2	84.5	89.6	81.7	96.8	84.4	88.1
Intra-ADR	Ours	82.4	79.4	95.3	82.3	84.9	87.7	81.2	97.1	83.8	87.5
$I^2$ -ADR	Ours	82.9	80.8	95.0	83.5	85.6	88.5	83.2	95.2	85.8	88.2
MixStyle + Intra-ADR	Ours	86.0	$\overline{80.3}$	96.0	84.4	86.7	88.6	$\overline{83.2}$	98.0	85.2	88.7
$MixStyle + I^2-ADR$	Ours	85.3	81.2	95.4	$\overline{86.1}$	$ \overline{87.0} $	$\overline{87.7}$	$\overline{84.5}$	$\overline{98.2}$	85.6	89.2

Office Ho	me										
Methods	Ar (	Cl	$\mathbf{Pr}$	$\mathbf{R}\mathbf{w}$	$\overline{ \mathbf{Avg.} }$	Methods	Ar	Cl	$\mathbf{Pr}$	$ \mathbf{Rw} $	$\mathbf{Av}$
	ResNet-18					$\overline{Re}$	sNet- $s$	50			
Baseline	57.8 52	$\overline{2.7}$	73.5	74.8	$\overline{  64.7  }$	Baseline	61.3	52.4	75.8	76.6	66.
RSC [21]	58.4 4'	7.9	71.6	74.5	63.1	MLDG [28]	61.5	53.2	75.0	77.5	66
MixStyle [78]	58.7 53	3.4	74.2	75.9	65.5	RSC [21]	50.7	51.4	74.8	75.1	65.
SagNet[42]	$ 60.2 \ 4 $	5.4	70.4	$\overline{73.4}$	62.3	SelfReg [24]	63.6	53.1	76.9	78.1	67.
FACT [65]	60.3 54	4.9	74.5	76.6	66.6	$\operatorname{SagNet} [42]$	63.4	54.8	75.8	78.3	68.
Intra-ADR	$ 64.5 \ 54$	4.0	73.9	74.7	66.8	Intra-ADR	67.3	54.1	78.8	78.8	69.

Intra-ADR	64.5 54.0	$73.9 \ 74.7$	66.8 In	itra-ADR		67.3	54.1	78.8	78.8	69.8
$I^2$ -ADR	66.4 53.3	74.9 75.3	$  67.5   I^2$	-ADR		70.3	55.1	80.7	79.2	71.4
MixStyle + Intra-ADR	$\sqrt{65.9}$ $55.3$	$\overline{74.3}$ 75.1	67.7 M	IixStyle +	Intra-ADR	$\overline{69.5}$	55.9	$\overline{80.6}$	80.4	$\overline{71.4}$
$MixStyle + I^2-ADR$	$66.8\ \overline{56.8}$	<b>75.3</b> 75.7	$\overline{68.7}$ M	IixStyle +	$I^2$ -ADR	71.1	<b>56.9</b>	81.8	$\overline{80.5}$	<b>72.5</b>
DomainNet										
Domainivet										
Methods	Reference	s   Clipart	Infograph	Painting	Quickdray	v R	Real	Ske	etch	Avg.

Dabonno		0	11.0	10.2	10.0	<u> </u>	00.1	00				
$\operatorname{MetaReg} [1]$	NeurIPS'18	53.7	21.1	<b>45.3</b>	$\overline{10.6}$	$\overline{58.5}$	42.3	38.6				
DMG [5]	ECCV'20	60.1	18.8	44.5	$\bf 14.2$	54.7	41.7	39.0				
Intra-ADR	Ours	$ 57.3\pm0.1 $	$14.9 \pm 0.3$	$42.8 {\pm} 0.2$	$12.2 {\pm} 0.4$	$52.9 \pm 0.5$	$46.0 \pm 0.2$	37.7				
$I^2$ -ADR	Ours	$57.3 \pm 0.3$	$15.2 \pm 0.3$	$44.1 {\pm} 0.1$	$12.1 {\pm} 0.4$	$53.9 \pm 0.6$	$46.7 {\pm} 0.2$	38.2				
MixStyle + Intra-ADR	Ours	$57.4 \pm 0.2$	$15.3 \pm 0.1$	$43.3 {\pm} 0.2$	$12.3 {\pm} 0.4$	$53.5 {\pm} 0.3$	$46.5 {\pm} 0.2$	38.1				
$MixStyle + I^2-ADR$	Ours	$57.4 \pm 0.4$	$15.7 {\pm} 0.2$	$44.7 \pm 0.1$	$12.3 {\pm} 0.4$	$54.4 \pm 0.2$	$\textbf{47.4} {\pm} \textbf{0.1}$	38.7				
ResNet-50												
Baseline	_	62.2	19.9	45.5	13.8	57.5	44.4	40.5				
$\operatorname{MetaReg}\left[1\right]$	NeurIPS'18	59.8	25.6	<b>50.2</b>	11.5	64.6	50.1	43.6				
MLDG [28]	AAAI'18	$59.1 \pm 0.2$	$19.1 \pm 0.3$	$45.8 {\pm} 0.7$	$13.4 {\pm} 0.3$	$59.6 {\pm} 0.2$	$50.2 {\pm} 0.4$	41.2				
C-DANN [31]	ECCV'18	$54.6 \pm 0.4$	$17.3 \pm 0.1$	$43.7 {\pm} 0.9$	$12.1 {\pm} 0.7$	$56.2 {\pm} 0.4$	$45.9 {\pm} 0.5$	38.3				
RSC [21]	ECCV'20	$55.0 \pm 1.2$	$18.3 {\pm} 0.5$	$44.4 {\pm} 0.6$	$12.2 {\pm} 0.2$	$55.7 \pm 0.7$	$47.8 {\pm} 0.9$	38.9				
DMG[5]	ECCV'20	65.2	22.2	50.0	15.7	59.6	49.0	43.6				
SagNet [42]	CVPR'21	$57.7 \pm 0.3$	$19.0 \pm 0.2$	$45\overline{.3\pm0}.3$	$12.7 {\pm} 0.5$	$58.1 {\pm} 0.5$	$48.8 {\pm} 0.2$	40.3				
SelfReg [24]	ICCV'21	$60.7 \pm 0.1$	$21.6 \pm 0.1$	$49.4 {\pm} 0.2$	$12.7 {\pm} 0.1$	$60.7 \pm 0.1$	$51.7 \pm 0.1$	$\mid 42.8$				
Intra-ADR	Ours	$63.6 \pm 0.1$	$20.0 \pm 0.1$	$49.4 \pm 0.1$	$14.8 \pm 0.3$	$60.0 \pm 0.4$	$54.4 \pm 0.1$	43.7				

#### Ablation Study

lethod	$ \mathcal{L}_{intra} $	$ \mathcal{L}_{dir} $	$\mathcal{L}_{dvr}$	$ \mathbf{Art}$	Cartoon	Photo	Sketch	$ \mathbf{Avg.} $	Method	$ E_s $	$ E_c $	$ \mathbf{Art} $	Cartoon	Photo	Sketch	Avg.
	✓	-	-	82.4	79.4	<b>95.3</b>	82.3	84.9		_	_	81.3	77.3	94.7	78.8	83.0
-ADR	-	✓	$\checkmark$	82.3	80.0	95.1	82.6	85.0	Intra-ADR	_	1	80.0	77.2	96.0	80.9	83.5
	✓	✓	-	82.7	80.5	95.0	83.2	85.4			_	81.9	79.3	95.5	$\frac{30.3}{79.3}$	84.0
	✓	-	$\checkmark$	82.5	80.2	95.1	82.9	85.2		\ \ /	_	$\frac{81.3}{82.4}$		$\frac{95.3}{95.3}$		l ——
	✓	<b>\</b>	✓	82.9	80.8	$\overline{95.0}$	83.5	85.6		<b>  √</b>	✓	02.4	$\bf 79.4$	95.5	<b>82.3</b>	84.9

#### **♦** Attention Visualization

