# Fine-Grained Auto Augmentation for multi-label Classification

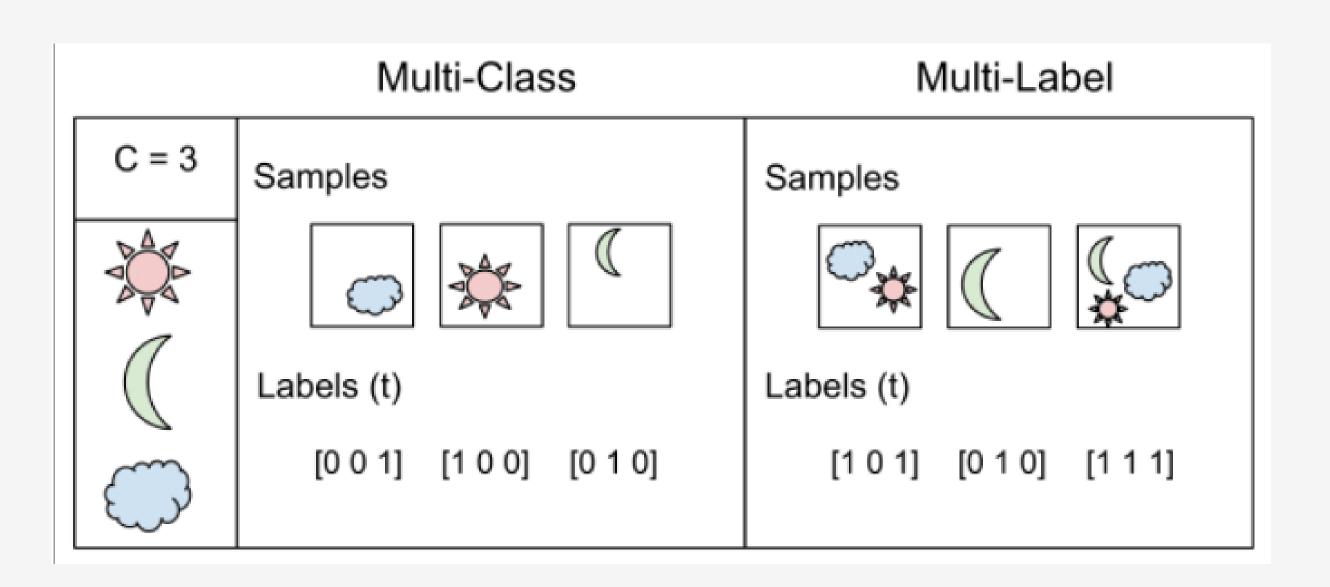
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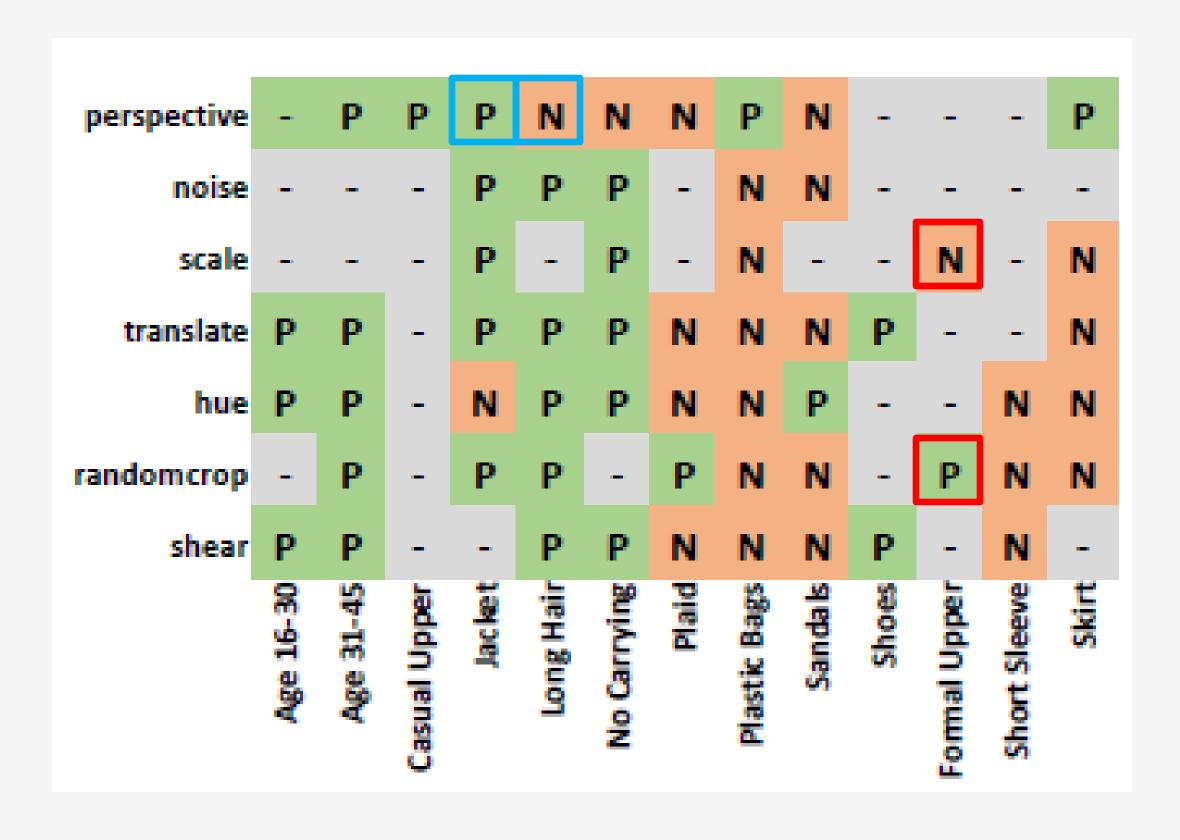
신우정

#### Multi-Class vs Multi-Label

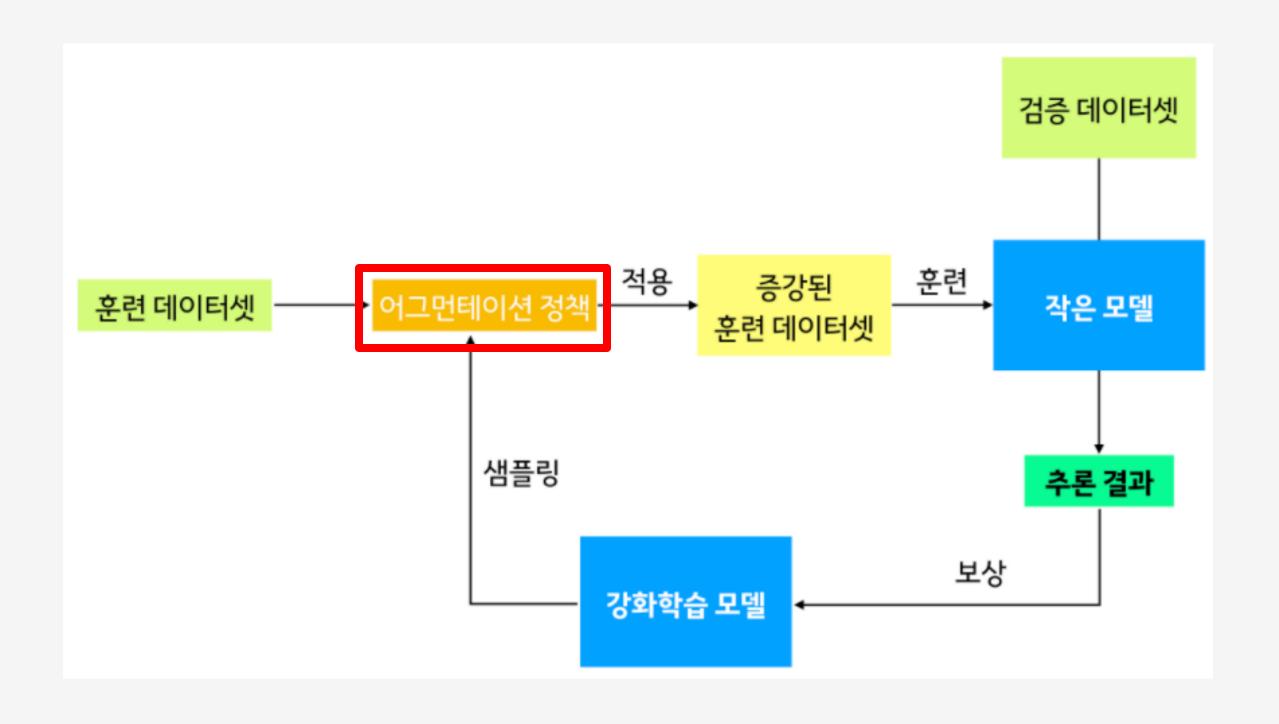


Softmax + cross entropy Sigmoid + cross entropy

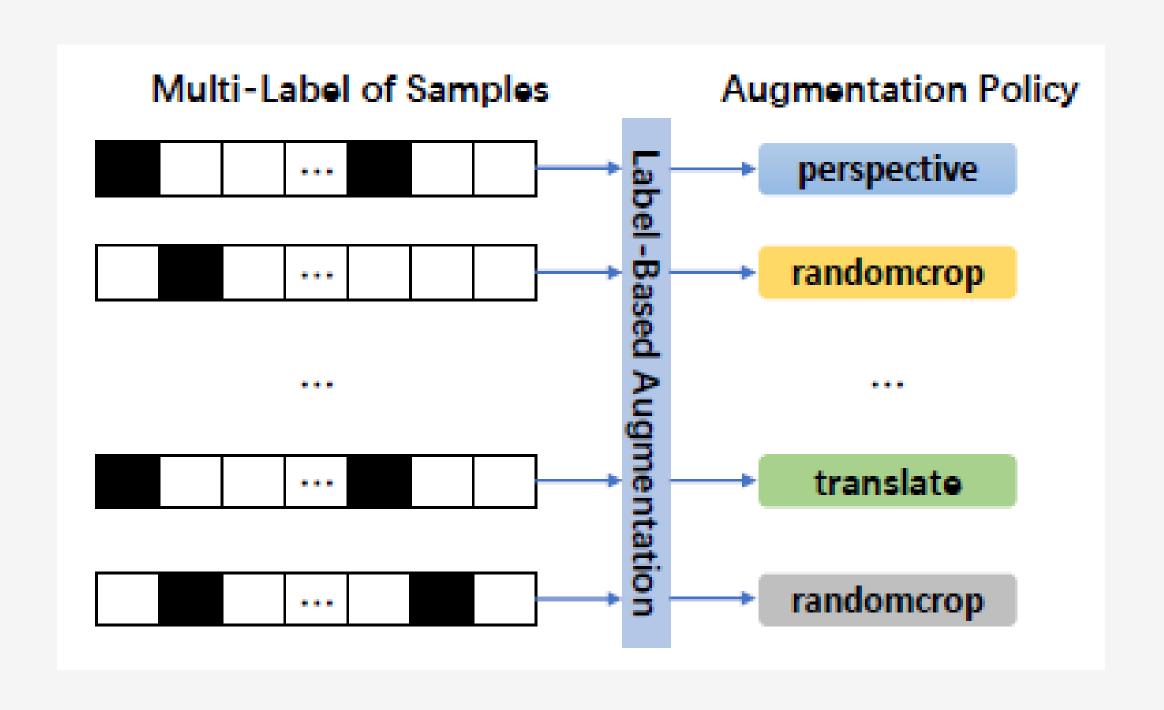
#### Grain/Drop of augmentation



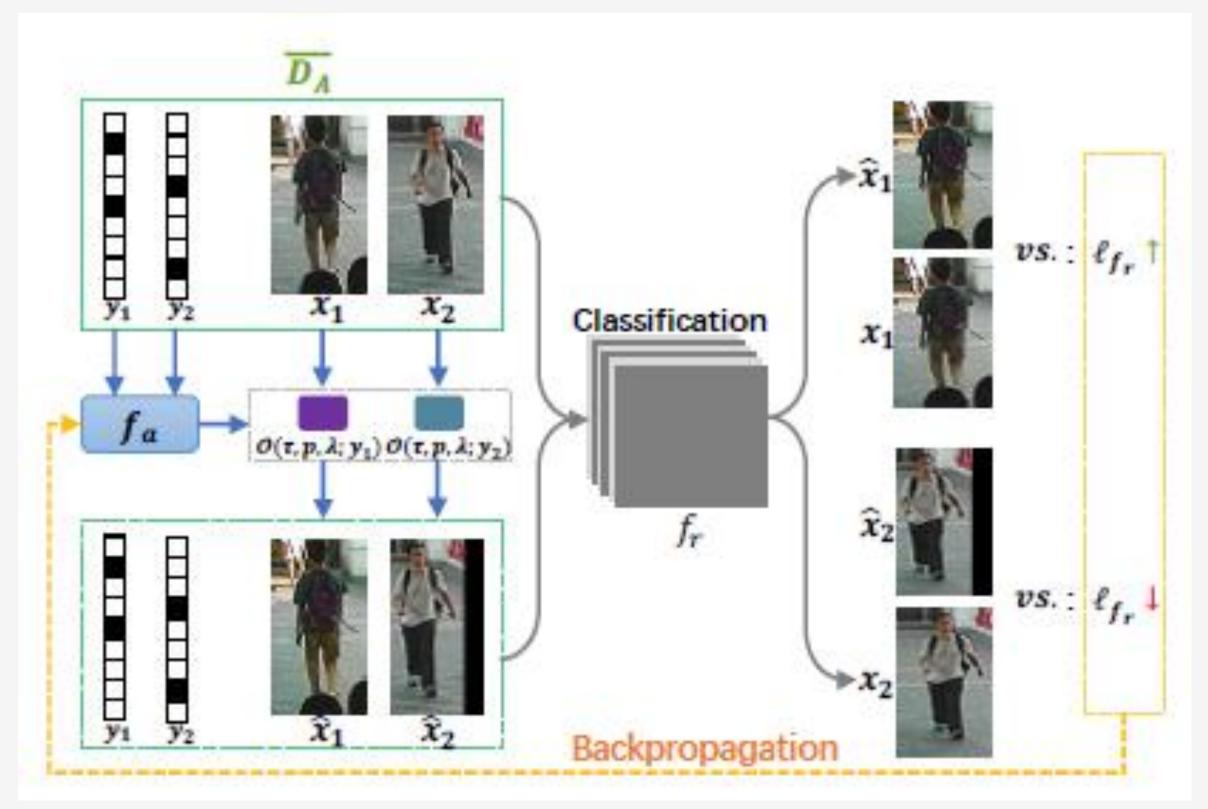
#### AA(AutoAugmentation) & FAA(Fast AutoAugmentaion)



#### Idea of LB-Aug (Label-Based AutoAugmentaion)



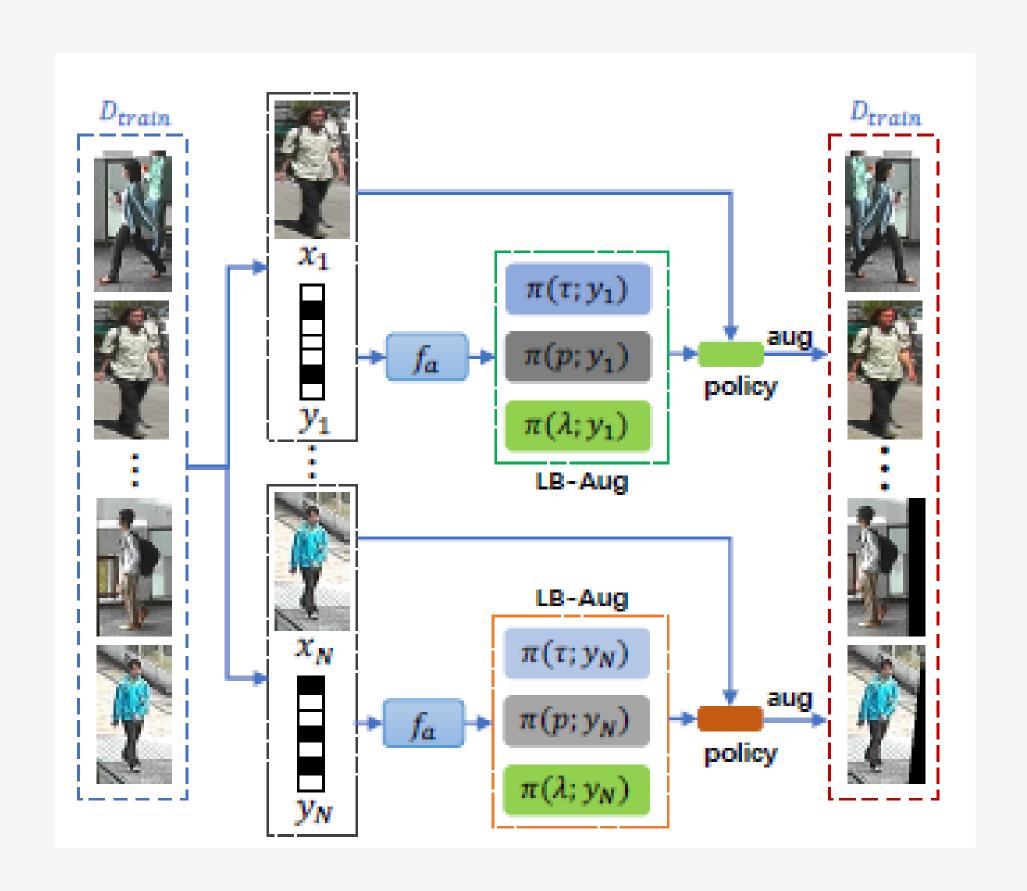
#### Augmentation model Training



 $f_a$ : LB Aug model

 $f_r$ : multi label classifier

#### LB-Aug model Training



#### **Experiment setup**

#### 1. Data

- Peta
- MS-COCO
- Chardes

- 3. Model
  - ResNet50
  - ResNet101
  - Inception-V3
  - Inception-I3D
  - S3D

- 2. Augmentation operations
- ShearX, ShearY, TranslateX, TranslateY, Rotate, AutoContrast, Invert, Equalize, Solarize, Posterize, Contrast, Color, Brightness,

Sharpness, Cutout and Sample Pairing)

#### Performance comparison on Peta

Method	ResNet50		ResNet101			Inception-V3			
	mA	Accu	F1	mA	Accu	F1	mA	Accu	F1
Baseline <sup>†</sup>	84.9	78.1	85.5	85.4	78.9	85.8	86.0	79.6	86.5
Baseline+Random <sup>‡</sup>	85.2	78.5	85.2	85.6	79.0	86.0	86.2	79.7	86.5
FAA 5	85.5	78.9	85.7	85.8	79.3	86.4	86.4	79.8	86.8
$LB-Aug_E^*$	86.6	79.9	86.8	86.6	80.1	86.8	87.2	80.8	87.3
$LB-Aug_H^*$	86.8	80.1	87.0	87.0	80.4	87.1	87.4	80.9	87.6

<sup>† :</sup> counterpart without additional augmentation from search space.

<sup>&</sup>lt;sup>‡</sup>: counterpart with random policies from search space.

<sup>\* :</sup> counterpart with fixed calling probabilities 0.5 and magnitudes 1.

<sup>\* :</sup> counterpart with learnable calling probabilities and magnitudes.

### Performance comparison on MS-COCO

Method	R	ResNet50		ResNet101			Inception-V3		
Wiction	mAP	CF1	OF1	mAP	CF1	OF1	mAP	CF1	OF1
Baseline	74.7	69.4	73.8	77.1	71.3	76.0	77.6	74.1	76.4
Baseline+Random	75.1	70.6	74.1	77.4	72.8	76.7	78.2	74.4	76.7
FAA	76.0	71.1	75.3	78.7	73.5	76.4	79.1	74.7	76.9
$LB-Aug_E$	77.1	72.3	76.2	79.7	74.5	77.5	80.2	75.9	77.8
$LB-Aug_H$	77.4	72.6	76.5	80.1	74.8	77.8	80.4	76.2	78.0

### Performance comparison on Chardes

Backbone	Baseline	Bas	eline+Random	FAA	LB-Aug	
		$p^{\dagger} = 0.25$	p = 0.75	p = 1	IAA	LD-Aug
I3D	36.3	35.5	34.8	34.1	36.4	37.6
S3D	36.8	36.2	35.7	35.0	37.0	38.0

#### Comparing LB-Aug against other SOTA methods on Peta

Methods	Backbone	Pretrain	mA	Accu	F1
JRL [21]	Alex Net	ImageNet	85.7	_	85.4
DeepMar [22]	CaffeNet	ImageNet	82.9	75.1	83.4
DeepMar <sup>†</sup> [22]	Inception-V3	ImageNet	81.5	_	85.7
VeSPA [23]	GoogleNet	ImageNet	83.5	77.7	85.5
WPAL [24]	GoogleNet	ImageNet	85.5	77.0	84.9
PGDM [25]	CaffeNet	ImageNet	83.0	78.1	85.8
ALM [18]	BN-Inception		86.3	79.5	86.9
GRL [20]	Inception-V3	ImageNet	86.7	_	86.5
FAA [5]	Inception-V3	ImageNet	86.4	79.8	86.8
$LB-Aug_H$	BN-Inception	ImageNet	86.7	80.1	87.2
$LB-Aug_H$	Inception-V3	ImageNet	87.4	80.9	87.6

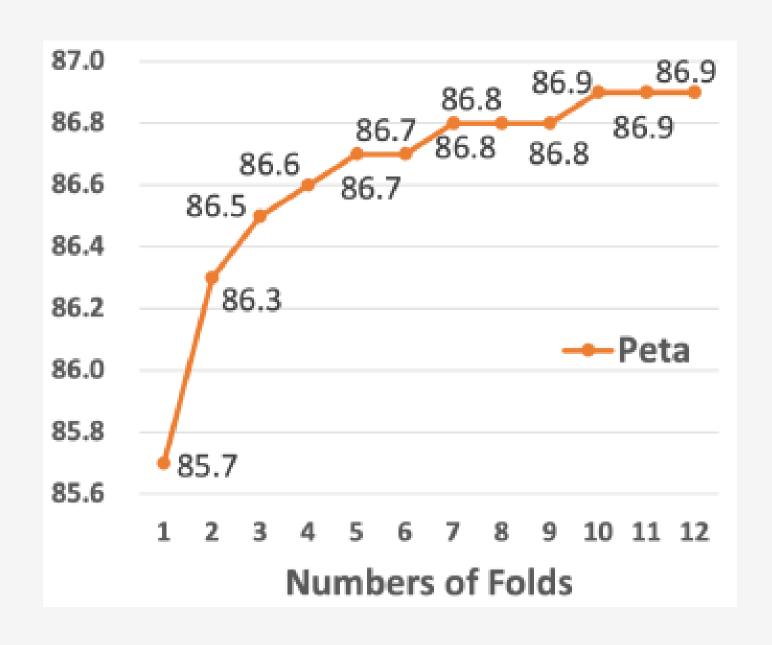
#### Comparing LB-Aug against other SOTA methods on COCO

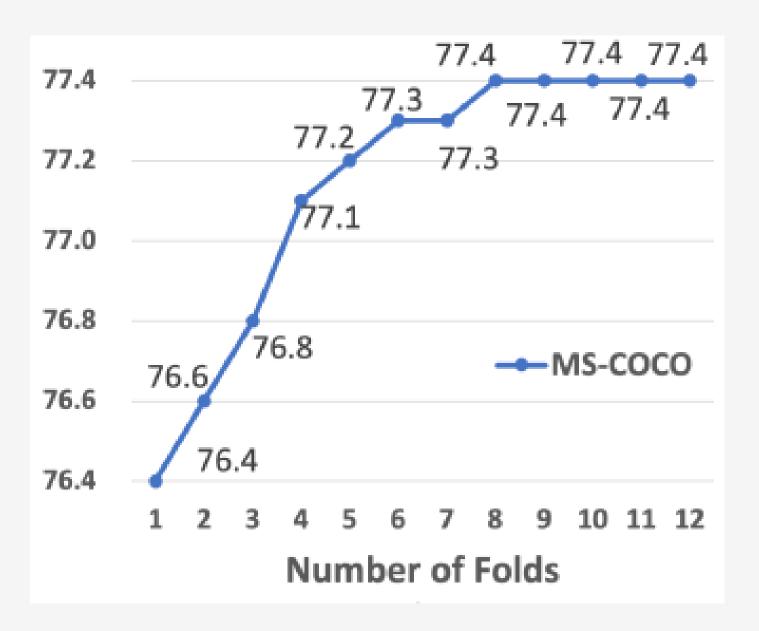
Methods	Backbone	Pretrain	mAP	CF1	OF1
SRN [26]	ResNet101	ImageNet	77.1	71.2	75.8
Multi-Evidence [27]	ResNet101	ImageNet	_	74.9	78.4
CADM [28]	ResNet101	ImageNet	82.3	77.0	79.6
ML-GCN [29]	ResNet101	ImageNet	83.0	78.0	80.3
KSSNet [3]	ResNet101	ImageNet	83.7	77.2	81.5
MS-CMA [30]	ResNet101	_	83.8	78.4	81.0
ASL [31]	TResNet-L	ImageNet	86.6 (86.4*)	81.4 (81.1*)	81.8 (81.6*)
ASL [31] <sup>†</sup>	TResNet-L	ImageNet	88.4 (88.1*)	-(81.6*)	-(82.3*)
$LB$ -Aug $_H^{\ddagger}$	TResNet-L	ImageNet	86.9	81.6	81.9
$LB-Aug_H^{\dagger \ddagger}$	TResNet-L	ImageNet	88.7	82.3	82.7

#### Generalization to other model

Method		ResNet101		Inception-V3			
Mictiod	mA	Accu	F1	mA	Accu	F1	
$LB-Aug_E$	86.6 (+0)	80.0 (-0.1)	86.6 (-0.2)	87.4 (+0.2)	80.9 (+0.1)	87.3 (+0)	
$LB-Aug_H$	86.9 (-0.1)	80.3 (-0.1)	87.0 (-0.1)	87.5 (+0.1)	80.9 (+0)	87.5 (-0.1)	

#### Performance curve on fold numbers K





## 감사합니다