MMfruit - OpenImage 2019 1st solution

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Results Breakdown



- 39 models in total (including exps):
- Data distribution:
 - 26 trained by all classes
 - 3 expert models (low AP)
 - 2 models from COCO
 - 8 models from 0365
- Framework:
 - 27 from pytorch
 - 10 from tensorflow
- 32~512 accelerators for each model
- 2 images on each accelerator

Solutions

- Multiple Large Backbones
- Gradient Decoupling
- Class Sampling & Full Batch
- Augmentation (Segmentation Label)
- Truncated Loss
- Multiscale Testing
- Adj-soft NMS
- Expert Model
- Weakly & Fully Supervised Pipeline
- Auto Ensemble

Multiple Large Backbones

Model Zoo



Decoupling Backbone (Naive implementation)



Gradient Decoupling

Learn the offset for classification and regression separately.

 $C = \mathcal{F}_c(F; heta_c) \;\; C \in \mathbb{R}^{k imes k imes 2}$

CML for classification: $L_c = \left|S_o - S + m_c
ight|_+$

Exp on OpenImage2019

$R = \mathcal{F}_r(F; heta_r)$	$R \in \mathbb{R}^{1 imes 1 imes 2}$
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for regression: $L_r = \left| IoU_o - IoU + m_r \right|_+$

Exp on COCO 2017 (FPN)

Model	DCP	Val	Public LB	Model	A1	A2	A3	CML	DCP	mAP(IOU=0.50:0.95)
ResNet50		64.64	49.79	ResNet50						36.1
ResNet50	V	68.18	52.55	ResNet50	\checkmark					37.3
DCN-ResNext101		68.70	55.05	ResNet50		V				38.0
DCN-ResNext101	V	71.71	58.59	ResNet50						38.5
DCN-SENet154		71.13	57.77	ResNet50					\checkmark	39.7
DCN-SENet154	V	72.19	60.5	ResNet50					\checkmark	40.8

A1: separate classification and regression

A2: Deformable ROIPooling for classification and ROIAlignPooling for regression.

A3: Deformable ROIPooling for classification and Deformable ROIPooling for regression.

Class Sampling & Full Batch

Class-aware sampling & full batch:



Augmentation (Segmentation Label)

Elaborate Augmentation

Select an image and a bounding box







Copy-and-Paste Augmentation



Truncated Loss





Missing (red dashed box): human eye, ear, nose, mouth,

. . .



Missing (red dashed box): wheel, tire, tree, light, ...



[600, 800, 1000, 1333, 1666, 2000]

Adj-soft NMS



 $Input: \mathcal{B} = \{b_1, \cdots, b_N\}, \mathcal{S} = \{s_1, \cdots, s_N\}, T = 0.5$ Begin

$$\mathcal{D} \leftarrow \{\}$$

 $\mathcal{D} \leftarrow Step1(\mathcal{B}, \mathcal{S}, \mathcal{D}, T) \text{ NMS } \}$
 $\mathcal{F} \leftarrow \{\}$
 $\mathcal{F} \leftarrow Step2(\mathcal{B}, \mathcal{S}, \mathcal{D}) \text{ Soft-NMS } \}$
 $return \ \overline{\mathcal{F}}, \overline{\mathcal{S}}$
 end

Model	Adj-soft NMS	Public LB
{4models}		59.40
{4models}	\checkmark	60.35

Expert Model



Weakly & Fully Supervised Pipeline



Auto Ensemble



• Functions set, such as NMS, Adj-soft NMS, and so on



Selecting the operation in each Ovia greedy algorithm

Misc.

• Confusing definitions:





• For more please read our solutions

Thanks, Q & A