Cdiscount's Image Classification

MinWoo Byeon





Cdiscount

 Cdiscount.com is the largest non-food e-commerce company in France.

Overview

• https://www.kaggle.com/c/cdiscount-image-classification-challenge

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Let's Go

Environments

- Python3 (with PyCharm)
- MXNet 0.12.0
- Pascal based 8-GPUs Machine

Dataset

Dataset Distribution

Dataset Cleaning

Models

	Batch Size	GPUs	Memory Usage (per GPU)	Speed (8-GPU, Sync)
ResNext-50	512	8		
ResNext-101-32x4d	512	8	12,461MiB	579.86 / sec
ResNext-101-64x4d	512	8	18,185 MiB	338.29 / sec
SE-ResNext-101-32x4d	512	8	14,057 MiB	411.55 / sec
SE-ResNext-101-64x4d	512	8	19,805 MiB	274.04 / sec
DPNs-92	512	8		
DPNs-98	512	8	16,367 MiB	380.22 / sec
DPNs-98 (224)	384	8	17,487 MiB	269.09 / sec
DPNs-107	256	8	17 707 MiB	246.06 / sec

Weight Initialization

- MSRA PReLu
 - https://arxiv.org/abs/1502.01852
 - Used in Squeeze-and-Excitation Networks

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Accuracy

		Pre-trained	ed Training		Testing (with flip)
	Models	Valid Acc.	Train Acc.	Valid Acc.	Valid Acc.	L.B
M1	ResNext-50					
M2	ResNext-101-32x4d					
М3	ResNext-101-64x4d				0.758480	
M4	DPNs-92					
M5	DPNs-98				0.756047	

Optimizer

- SGD
- NAG
- Adam
- NAdam

Augmentation

• ResNet-34, Batch=512, Opt=Adam, Base-LR=0.0005, Epoch=15

Input Size	Option	Local Train (Top-1)	Local Validation (Top-1)
180 x 180	No Aug.	0.862802	0.688090
180 x 180	Flip	0.787349	0.692491
180 x 180	Flip + HSL	0.774842	0.690422
224 x 224 (NN)	Flip	0.789405	0.695116

Image Size

• ResNet, stride//2 만큼 pad

Input Size	conv	stage1	stage2	stage3	stage4	Local Train (Top-1)	Local Validation (Top-1)
160 x 160	80 x 80	40 x 40	20 x 20	10 x 10	5 x 5		
180 x 180	90 x 90	45 x 45	23 x 23	12 x 12	6 x 6		
192 x 192	96 x 96	48 x 48	24 x 24	12 x 12	6 x 6		
224 x 224	112 x 112	56 x 56	28 x 28	14 x 14	7 x 7		

Label Smoothing

• ResNet-34, Batch=512, Opt=Adam, Base-LR=0.0005, Epoch=15

Smooth alpha	Local Train (Top-1)	Local Validation (Top-1)
0.0	0.785019	0.691917
0.1	0.779468	0.694035
0.2	0.770593	0.692766

GAP + Dropout

• ResNet-34, Batch=512, Opt=NAdam, Base-LR=0.0005, Epoch=15

Dropout Ratio	Local Train (Top-1)	Local Validation (Top-1)
0.0	0.788686	0.693604
0.2	0.743764	0.689954

Multi-View Testing

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MD5 Dictionary

ResNext-101-64x4d (epoch 15, original+flip)

Method	Local Test (geo)	Leader Board
None		0.75671 (arith-maen)
Unique	0.763840	0.76025 (arith-mean)
Majority		0.75308 (geo-mean)
L1		0.76205 (geo-mean)
L2		0.76206 (geo-mean)
Softmax		

Ensemble

		Training			Testing (with flip)	
	Models	Epochs	Train Acc.	Valid Acc.	Valid Acc.	L.B
M2	ResNext-101-32x4d	11				0.74718
М3	ResNext-101-64x4d	15			0.758480	0.75671
M5	DPNs-98	15			0.756047	0.75470
	M3 + M5	-	-	-	0.770675	0.76972

Ensemble Methods

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		Local Validation		Public Lea	ider Board
Model	View	Arithmetic Mean Geometric Mean		Arithmetic Mean	Geometric Mean
3-Ensemble	6			0.77784	0.77918
6-Ensemble	12			0.78268	0.77468

Ensemble Methods

• Ensemble 4 models

Method	Local Validation (Top-1)	Desc
model arith, img arith	0.782746	
model arith, img geo (1 imgs)	0.781875	
model arith, img geo (2 imgs)	0.782525	
model arith, img geo (3 imgs)	0.782848	
model arith, img geo (4 imgs)	0.783498	
model geo, img geo	0.785558	

Ensemble Models

Model	Model Validation (Product Level)	Ensemble Accuracy
dpn107e19	0.767340	0.767340 (+0.0)
seresnext101-c64-15	0.763164	0.778684 (+0.011344)
dpn131e15	0.762358	0.782644 (+0.015304)
resnext101-c64-e17	0.757778	0.780899
densenet161-e15	0.755631	0.782757

MD5 Dictionary

ResNext-101-64x4d (epoch 15, original+flip)

Method	Arithmetic Mean	Geometric Mean
None	0.758480 (LB: 0.75671)	0.761119
Unique	0.761724 (LB: 0.76025)	0.763840
Majority	0.757597	0.754174 (LB: 0.75308)
L1	0.761747	0.763490 (LB: 0.76205)
L2	0.761574	0.763490 (LB: 0.76206)
Softmax	0.753390	0.756910

Large Size Prediction

ILSVRC 2017 Classification Competition

- A. The number of first 1×1 convolutional channels for each bottleneck building block was halved to reduce the computation cost of the network with a minimal decrease in performance
- B. The first 7×7 convolutional layer was replaced with three consecutive 3×3 convolutional layers.
- C. The down-sampling projection 1×1 with stride-2 convolution was replaced with a 3×3 stride-2 convolution to preserve information

ILSVRC 2017 Classification Competition

- D. A dropout layer (with a drop ratio of 0.2) was inserted before the classifier layer to prevent over-fitting
- E. Label-smoothing regularisation was used during training.
- F. The parameters of all BN layers were frozen for the last few training epochs to ensure consistency between training and testing

Sub-category model

- Library Category
 - All Category: 0.511904
 - Library category: 0.530334

Batch Prediction

• ZMQ + OpenCV

Mistakes

- 평균 잘못 구함
- RGB 채널 순서
- object reference

```
>>> import numpy as np
>>> a = np.array([1,2,3])
>>> a
array([1, 2, 3])
>>> b = a
>>> b
array([1, 2, 3])
>>> a
array([1, 2, 3])
>>> b += a
>>> a
array([2, 4, 6])
>>> b
array([2, 4, 6])
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Leaderboard

Time

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Additional Methods

Large Margin Softmax

- https://arxiv.org/pdf/1612.02295.pdf
- https://github.com/luoyetx/mx-lsoftmax

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Mean-Max Pooling

OCR

- Book, Album, ...
- Tesseract OCR
 - https://github.com/tesseract-ocr/tesseract
- Too Bad...

B-CNN

Multi-input CNN

• 12-channel(4 images) input

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References