

# VP-ReID: Vehicle and Person Re-Identification System

Longhui Wei, Xiaobin Liu, Jianing Li, Shiliang Zhang

School of Electronic Engineering and Computer Science, Peking University, Beijing 100871, China

{longhuiwei,liu-xb,ljn-vmc,slzhang,jdl}@pku.edu.cn



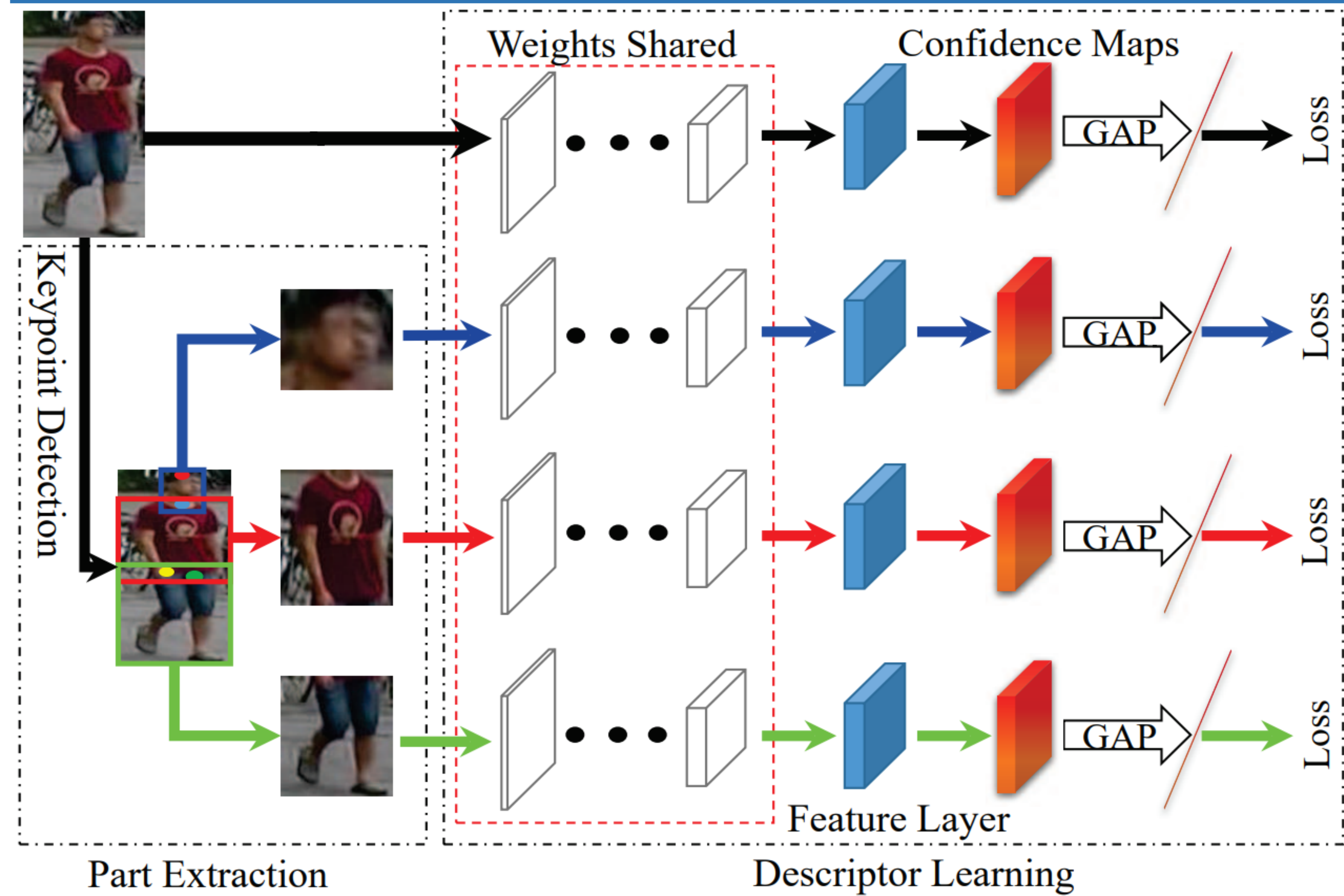
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## Abstract

Person Re-Identification (ReID) and vehicle ReID are the key technology in smart surveillance system. We develop a robust and efficient ReID system, named VP-ReID, to demonstrate our recent research progresses on those two tasks. This system is build based on our recent works including discriminative feature design and efficient off-line indexing. Constructed upon those algorithms, VP-ReID can identify vehicle and person efficiently and accurately from large gallery set.

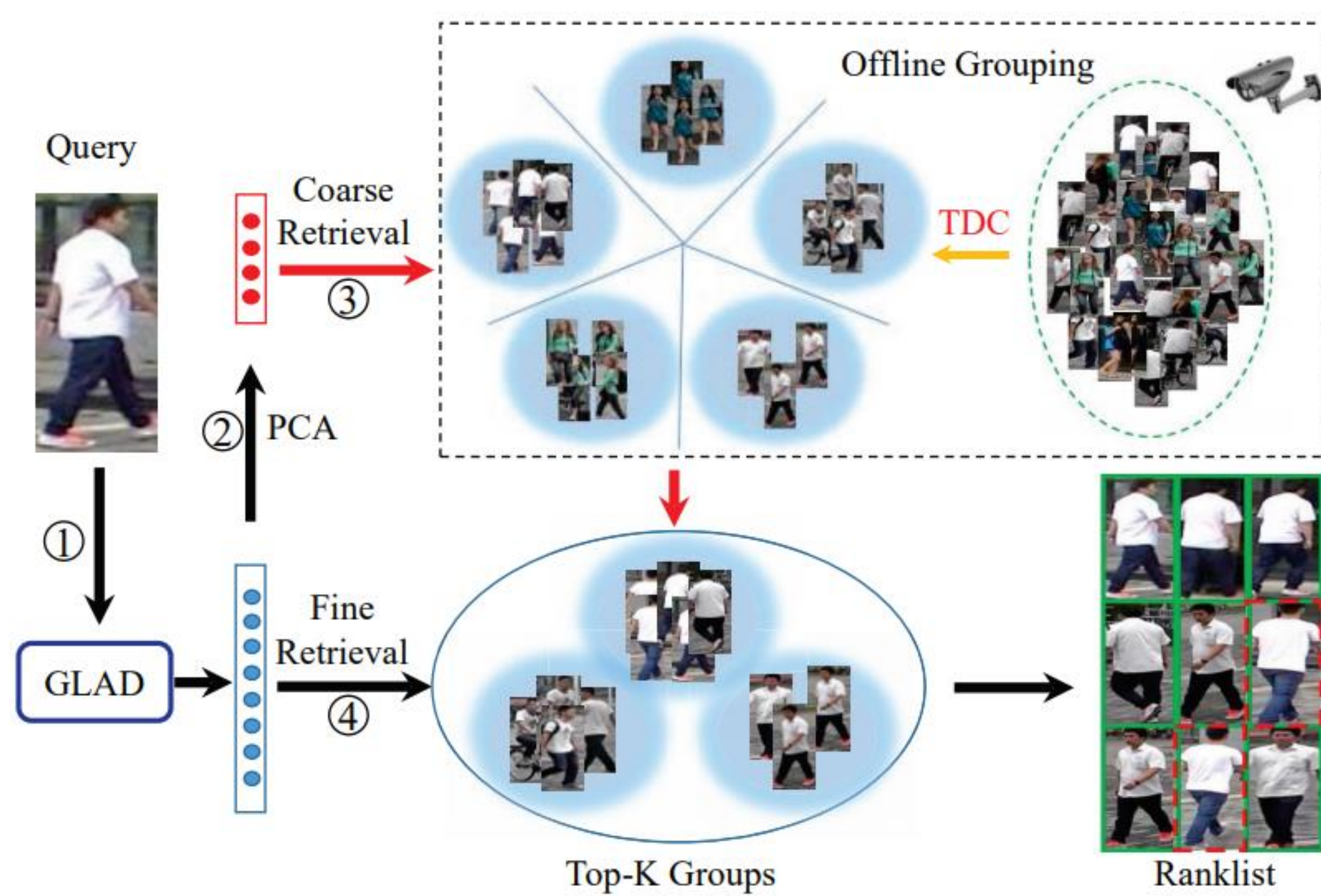
## GLAD Descriptor

- ◆ GLAD explicitly leverages local and global cues to generate a discriminative and robust representation
- ◆ Learn complementary features on both coarse-grained local parts and global regions
- ◆ The matching of local representation can effectively handle the misalignment and pose change issues



## Offline Grouping and Online Retrieval

- ◆ First offline clusters similar images into same group
- ◆ Images in returned groups are retrieved with original GLAD to generate an image rank list.



Re-ID Efficiency on *Market1501*

$\theta$	Group Number	Dim	mAP	Rank-1	Times(ms)
0.0000	19732	4096	73.9	89.9	368
0.0010	13509	4096	73.7	89.9	267
<b>0.0015</b>	<b>8509</b>	<b>4096</b>	<b>73.2</b>	<b>89.9</b>	<b>176</b>
0.0020	2558	4096	71.7	89.8	101
0.0015	8509	512	73.1	89.9	50
0.0015	8509	128	73.0	89.8	31
0.0020	2558	512	71.6	89.7	69
0.0020	2558	128	71.4	89.7	61

## Comparison On *Market1501*

Methods	mAP	rank1
MSCAN [Li, CVPR2017]	57.5	80.3
SVDNet [Sun, CVPR2017]	62.1	82.3
CSA [Zhong, CVPR2018]	68.7	88.1
PSE [Sarfranz, CVPR2018]	69.0	87.7
<b>GLAD</b>	<b>73.9</b>	<b>89.9</b>
GLAD + re-Ranking [Zhong, CVPR2017]	<b>87.1</b>	<b>91.2</b>

## Comparison On *DukeMTMC-ReID*

Methods	mAP	rank1
ACRN [Schumann, CVPR2017]	51.96	72.58
SVDNet [Sun, ICCV2017]	-	76.9
AACN [Xu, CVPR2018]	59.3	76.8
CSA [Zhong, CVPR2018]	57.6	78.3
PSE [Sarfranz, CVPR2018]	62.0	79.8
<b>GLAD</b>	<b>62.2</b>	<b>80.0</b>
GLAD + re-Ranking [Zhong, CVPR2017]	<b>79.3</b>	<b>84.4</b>

# Vehicle Re-Identification:

## Region-Aware deep Model (RAM)

- ◆ We propose a Region-Aware deep Model (RAM) to jointly learn global and regional features.
- ◆ Attribute cues are additionally used to jointly train RAM.
- ◆ Learned features are more discriminative to detailed local cues, and contains attribute cues.

## Experimental Results

1. Performance comparison of features learned by different models on *VeRi*.

Models	mAP	Top-1	Top-5
Baseline	0.550	0.848	0.931
BN	0.581	0.871	0.940
BN+R	0.609	0.887	0.941
<b>RAM</b>	<b>0.615</b>	<b>0.886</b>	<b>0.940</b>

2. Performance comparison of features learned by different models on *VehicleID*.

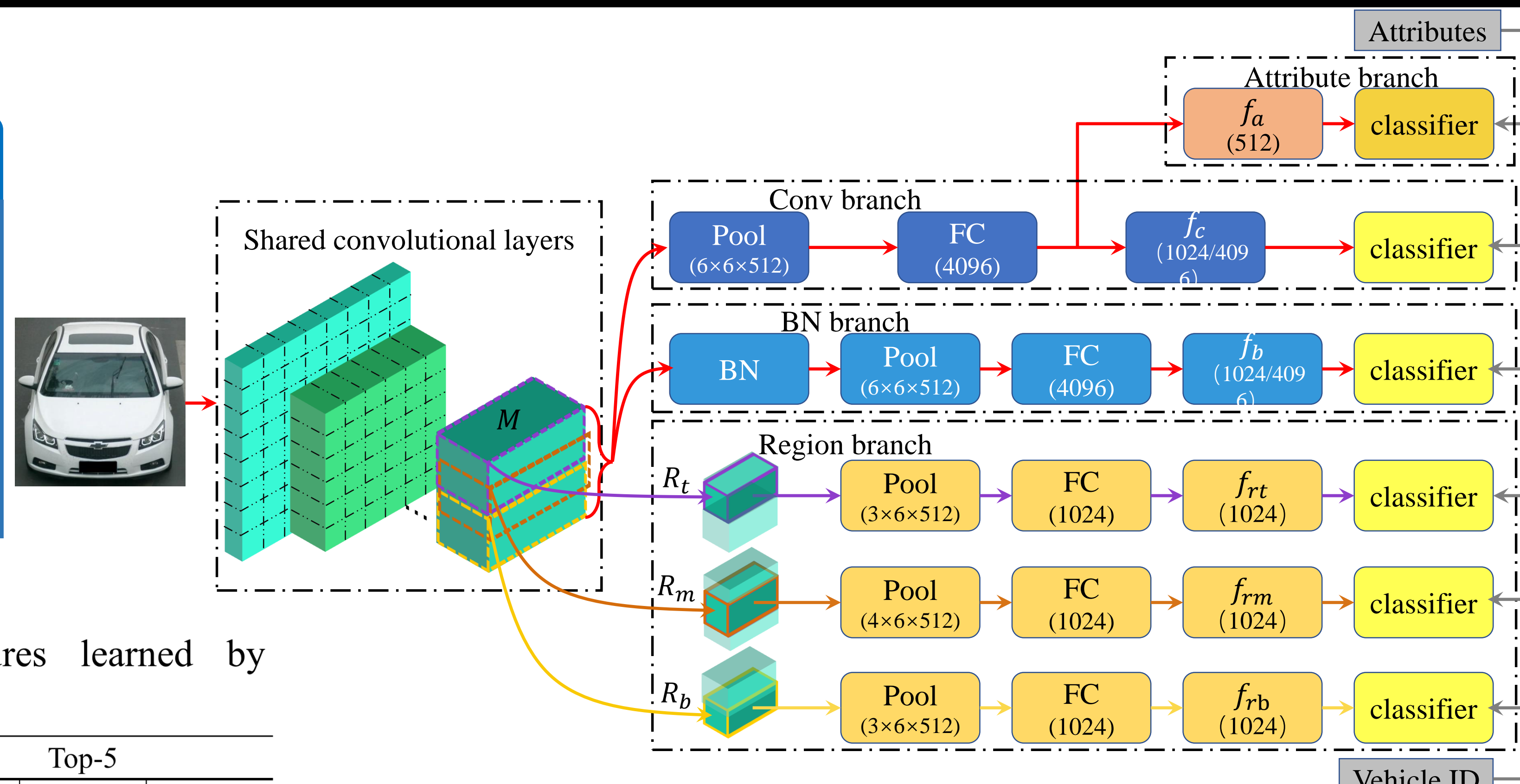
Models	Top-1			Top-5		
	Small	Medium	Large	Small	Medium	Large
Baseline	0.694	0.673	0.632	0.892	0.820	0.795
BN	0.722	0.705	0.666	0.904	0.853	0.832
BN+R	0.747	0.720	0.674	0.908	0.863	0.842
<b>RAM</b>	<b>0.752</b>	<b>0.723</b>	<b>0.677</b>	<b>0.915</b>	<b>0.870</b>	<b>0.845</b>

3. Comparison with recent works on *VeRi*.

Models	mAP	Top-1	Top-5
FACT[Liu, ICME2016]	0.199	0.597	0.753
FPSS[Liu, ECCV2016]	0.278	0.614	0.788
SCPL[Shen, ICCV2017]	0.583	0.835	0.900
OIF[Wang, ICCV2017]	0.480	0.659	0.877
OIF+SF[Wang, ICCV2017]	0.514	0.683	0.897
<b>RAM</b>	<b>0.615</b>	<b>0.886</b>	<b>0.940</b>

4. Comparison with recent works on *VehicleID*.

Models	Top-1			Top-5		
	Small	Medium	Large	Small	Medium	Large
VGGT[Liu, CVPR2016]	0.404	0.354	0.319	0.617	0.546	0.503
CCCL[Liu, CVPR2016]	0.436	0.370	0.329	0.642	0.571	0.533
MDCCL[Liu, CVPR2016]	0.490	0.428	0.382	0.735	0.668	0.616
OIF[Wang, ICCV2017]	-	-	0.670	-	-	0.829
<b>RAM</b>	<b>0.752</b>	<b>0.723</b>	<b>0.677</b>	<b>0.915</b>	<b>0.870</b>	<b>0.845</b>



## Retrieval Result

