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# **Rotation Invariant Aerial Image Retrieval with Group Convolutional Metric Learning**

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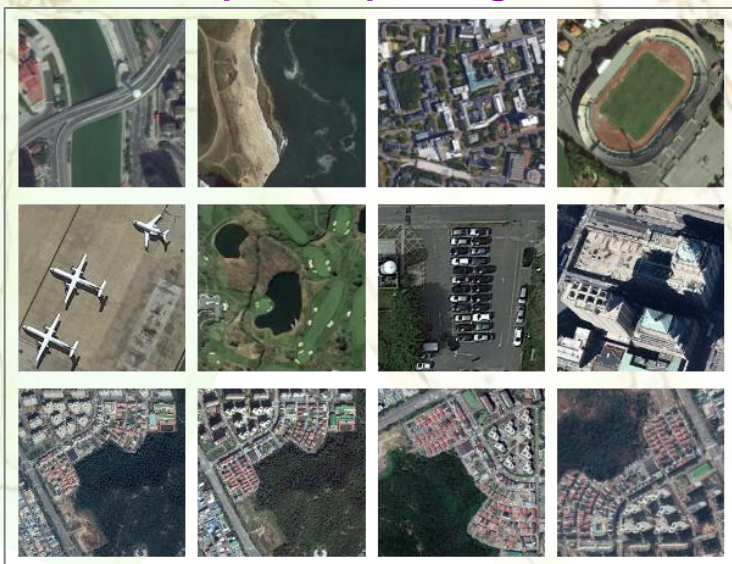
# Introduction

- **Goal**

- Developing a framework to retrieve aerial images with rotational variations

- **Motivation**

- Rotated aerial image retrieval is highly complex task
  - Contains small objects and buildings with variations in scale, shooting range, and height
- Robust retrieval network for rotated aerial images in demand
  - Variations in camera viewpoint depending on aircraft



Example images from AID, NWPU-RESISC45, and Google Earth South Korea dataset



# Methods

## Network modules

### ➤ Group convolutional neural network (G-CNN)

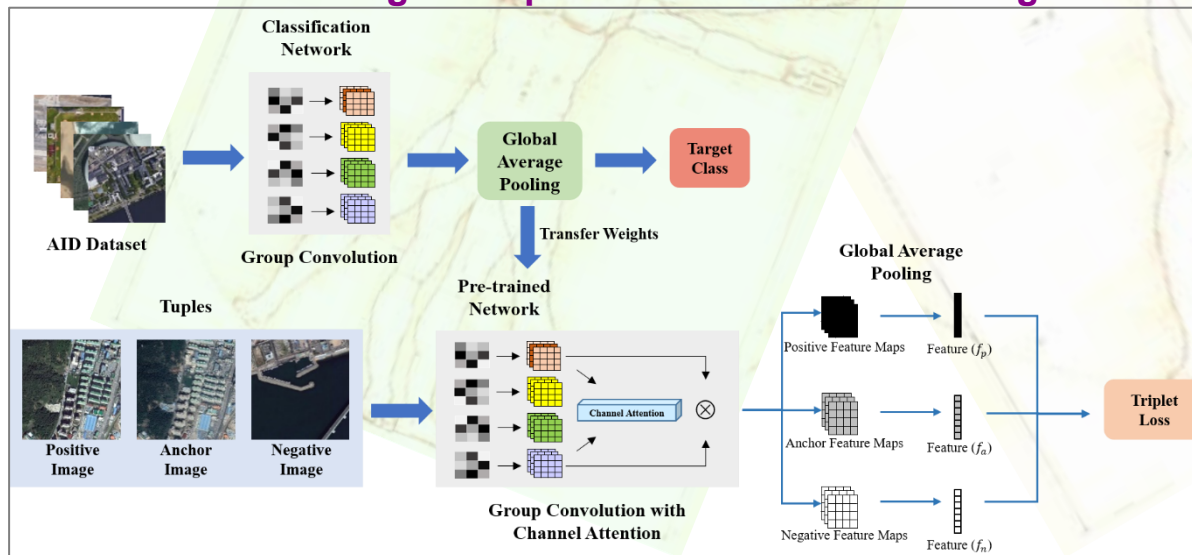
- Utilize rotated filters to pretrain the network for classification task
  - Aerial Image Dataset (AID)
- Fine-tune the pre-trained network with attentive G-CNN and metric learning

### ➤ Channel attention module

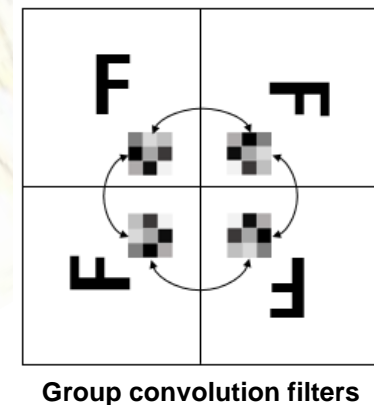
- Emphasize the important features among G-CNN layers

### ➤ Deep metric learning

- Integrate triplet loss function for training with anchor, positive, and negative



Overall architecture of proposed method







# Experimental Results (1/3)

## Performance evaluation on rotated Google Earth South Korea dataset

### ➤ Recall@n

- The percentage of correctly retrieved query up to different  $n$  values

$$Recall @ n = \frac{1}{k} \sum_{i=1}^k \frac{|Ret_i^n \cap Rel_i|}{|Rel_i|}$$

- $Ret^n$  : Top-n retrieved items
- $Rel$  : Total relevant items
- $k$  : Number of query items

(†) indicates additional rotation training

(\*) indicates added channel attention

All methods are pre-trained in AID dataset

Methods	Recall @ n (%)			
	$n = 1$	$n = 5$	$n = 10$	$n = 100$
R-MAC descriptor†	6.5	14.5	24.8	64.0
NetVlad†	7.4	17.4	25.1	68.2
Contrastive loss†	8.1	16.8	24.5	65.0
Triplet loss†	6.9	18.0	24.7	65.6
LDCNN†	8.9	18.4	24.7	66.6
G-CNN (p4m) + Cont. loss	17.8	32.4	38.9	72.0
G-CNN (p4m) + Cont. *	18.5	32.8	39.8	72.0
G-CNN (p4) + Triplet loss	20.1	36.0	43.2	76.9
G-CNN (p4) + Triplet. *	21.2	36.4	43.9	77.2
G-CNN (p4m) + Triplet loss	23.4	44.0	51.7	84.6
<b>G-CNN (p4m) + Triplet loss *</b>	<b>24.5</b>	<b>46.9</b>	<b>52.8</b>	<b>86.3</b>



# Experimental Results (2/3)

## Performance evaluation on rotated NWPUR-RESISC45 dataset

(†) indicates additional rotation training

(\*) indicates added channel attention

All methods are pre-trained in AID dataset

Methods	Recall @ n (%)			
	n = 1	n = 2	n = 4	n = 8
R-MAC descriptor†	23.1	41.4	62.4	83.3
Contrastive loss†	26.6	46.1	66.7	86.4
NetVlad†	27.1	45.3	66.7	85.5
Triplet loss†	35.3	53.1	73.2	88.7
LDCNN†	36.0	54.5	74.0	88.8
G-CNN (p4m) + Cont. loss	39.9	56.6	74.9	88.3
G-CNN (p4m) + Cont. *	41.1	59.8	76.9	89.3
G-CNN (p4) + Triplet loss	43.1	60.8	77.6	89.5
G-CNN (p4) + Triplet. *	43.4	61.3	77.7	89.2
G-CNN (p4m) + Triplet loss	44.8	62.1	77.8	89.5
<b>G-CNN (p4m) + Triplet loss *</b>	<b>45.7</b>	<b>64.3</b>	<b>80.2</b>	89.5



# Experimental Results (3/3)

- Examples of retrieval results in the rotated Google Earth South Korea dataset

